Atmospheric boundary layer height from ground-based remote sensing: a review of capabilities and limitations

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Abstract. The atmospheric boundary layer (ABL) ~~height~~ defines the volume of air ~~within which~~ for the dilution of heat, moisture and ~~pollutants released at the Earth's surface are rapidly diluted. Despite the importance for air quality interpretation~~trace substances. Quantitative knowledge on the temporal and spatial variations of the heights of the ABL and its sublayers is still scarce, despite their importance for a series of applications (including, e.g., air quality, numerical weather prediction, greenhouse gas assessment and renewable energy ~~applications, amongst others, quantitative knowledge on the temporal and spatial variation in ABL height is still scarce. With continuous profiling of the entire ABL vertical extent at high temporal and vertical resolution now increasingly possible due~~ production). Thanks to recent advances in ground-based remote sensing measurement technology and algorithm development, ~~there are also dense measurement networks emerging across Europe and other parts of the world. To effectively monitor the spatial and temporal evolution of the ABL continuously at continent-scale, harmonised operations and data processing are key. Autonomous~~ continuous profiling of the entire ABL vertical extent at high temporal and vertical resolution is increasingly possible. Dense measurement networks of autonomous ground-based remote sensing instruments, such as microwave radiometers, radar wind profilers, Doppler wind lidars or automatic lidars and ceilometers ~~, each offer different capabilities. The overarching objective of this review is to emphasize how these instruments are best exploited with informed network design, algorithm implementation, and data interpretation. A summary of the capability~~ are hence emerging across Europe and other parts of the world. This review summarises the capabilities and limitations of ~~each instrument type is provided together with a review of~~ various instrument types for ABL monitoring and provides an overview on the vast number of retrieval methods developed for ~~ABL height detection~~ the detection of ABL sublayer heights from different atmospheric quantities (temperature, humidity, wind, turbulence, aerosol). It is outlined how the diurnal evolution of the ABL can be monitored effectively with a combination of methods, ~~highlighting where instrument~~ pointing out where instrumental or methodological synergy ~~promise to be particularly valuable. To demonstrate the vast potential of increased ABL monitoring efforts, long-term observational studies are reviewed summarising our current understanding of ABL height variations~~are considered particularly promising. The review ~~emphasizes~~ highlights that harmonised data acquisition ~~and careful~~ across carefully-designed sensor networks as well as tailored data processing are key to obtaining high-quality products, which are essential to capture the spatial and temporal complexity of the lowest part of the atmosphere in which we live and breathe.

1 Introduction

The Atmospheric Boundary Layer (ABL) is the lowest part of the atmosphere ~~, defining the air we breathe, the winds harvested by wind turbines and the formation of fog or low-level clouds, amongst many other processes~~where most of the interactions between the Earth's surface and the atmosphere take place (Seibert et al., 1998). It plays a crucial role for the exchange of momentum, heat, humidity, aerosols, as well as greenhouse and other atmospheric gases ~~between the Earth's surface and the atmosphere~~ (Palmén and Newton, 1969; Garratt, 1994; Stull, 1988). Improved ~~process understanding~~ process-understanding and quantitative knowledge of ~~ABL dynamics is~~ ABL dynamics are hence crucial for a wide range of applications with high societal, economic and health impacts, including ~~air quality assessment (Han et al., 2009), renewable energy generation (Peña et al., 2016)~~the assessment of air quality (e.g., Han et al., 2009; Stirnberg et al., 2021; Sujatha et al., 2016) or greenhouse gases (e.g., Lauvaux et al., 2016), generation of renewable energy (e.g., Peña et al., 2016), numerical weather prediction ~~(NWP; Illingworth~~ (NWP; e.g., Illingworth et al., 2019), sustainable urban planning ~~(Barlow et al., 2017)~~(Barlow et al., 2017, e.g.,), and all aspects of transportation such as aviation, shipping, or road safety ~~(Vajda et al., 2011). Still, the ABL presents the single-most under-sampled part of the atmosphere as its processes are not captured adequately, neither by surface-based station networks nor by the increasingly rich atmospheric profile data gathered by satellites. The latter reaching maximum uncertainty close to the Earth's surface (Zhang et al., 2016; Abril-Gago et al., 2021)~~(e.g., Vajda et al., 2011).

Sampling the ABL vertical profile has historically been mostly achieved using ~~radiosonde~~radiosondes. While these balloon ascends provide indispensable information, their temporal resolution is usually insufficient to capture the full diurnal evolution of the ABL dynamics and the significant horizontal drift of the balloon during the ascent means observations are affected by spatial variations in ABL dynamics which can be challenging for data analysis and interpretation. In recent decades, ground-based remote sensing has started to close this gap, providing high-resolution information, initially with a focus on the lowest kilometre of the atmosphere (see reviews by Wilczak et al., 1997; Emeis et al., 2008). Significant advances in ground-based remote sensing measurement technology and algorithm development now allow for continuous profiling of the entire ABL vertical extent (ranging from a few tens of metres to >3 km, or even higher, depending on geographic settings and synoptic conditions) at high temporal and vertical resolution (Illingworth et al., 2019; Cimini et al., 2020) and automatic detection of ABL ~~layer~~ sublayer heights from different atmospheric quantities (Collaud Coen et al., 2014; Duncan et al., 2021).

With dense ground-based remote sensing networks emerging in Europe and other parts of the world, it is vital to recap capabilities and limitations of the various instruments and analytical approaches to support ~~informed~~ careful network design, algorithm implementation, and sound interpretation of the results. In their recent review Zhang et al. (2020) stress that interpretation of ABL height data should always take into account the specifics of both the retrieval algorithm (e.g. which atmospheric variable is analysed?) and the input data (e.g. characteristics of the sensor used for data acquisition).

The objective of this review is to provide a general overview on the latest ABL profiling techniques while making relevant details easily accessible. The sections hence offer multiple entry points (Figure 1) catering to a range of user backgrounds. The different atmospheric variables routinely analysed to gain insights on the ABL are presented in Section 1.1. Sensor types commonly used for ABL profiling are introduced in Section 2, highlighting their respective capabilities and limitations as well as their deployment in organised sensor networks. The wide range of ABL height retrieval methods is then reviewed in Section 3, linking potential retrieval errors to uncertainties inherent in the observed atmospheric quantity where appropriate. Quantification of layer height uncertainties is challenging, particularly due to the absence of an 'absolute truth' concept that could serve as the reference standard. Section 4 outlines how the various layer height retrievals based on different atmospheric quantities compare throughout the ABL diurnal evolution and depending on atmospheric stability or cloud conditions. This is to support a data users' assessment of how well a certain layer height product may characterise their process of interest. ~~Finally, Section **??** provides an overview of the current knowledge on ABL heights globally, highlighting processes and geographic characteristics that drive spatial and temporal variations.~~

Ground-based profile remote sensing is a powerful tool to enhance our understanding of the atmospheric boundary layer. With careful, harmonised measurement network operations and processing procedures, increasingly detailed information can be collected that is very powerful to support many high-impact applications. The conclusions (Sect. 5) emphasise which aspects of data acquisition, algorithm development, data analysis and applications require additional attention to best advance this area of scientific research, sensor development, ~~scientific research~~ and environmental monitoring operations.

1.1 The atmospheric boundary layer and its sub-layers

The ~~Atmospheric Boundary Layer~~[1] ~~(ABL)~~ [1] is the lowest part of the troposphere where direct interactions with the Earth's surface (land and sea) take place (Seibert et al., 2000). It responds directly ~~(or indirectly)~~ to surface forcing at time scales of less than one hour ~~(day) (Garratt, 1994)~~(Garratt, 1994) while indirect effects (e.g. in the residual layer) can extend to daily time scales. Exchange mechanisms include the transfer of momentum, radiation, heat, moisture, particles and gases. ~~These can be driven by contrasts in surface cover, roughness, topography as well as wind shear (e.g., low-level jet) or cloud dynamics (Garratt, 1994; Seibert et al., 2000).~~ The ABL defines the volume in which ~~gases and aerosol particles emitted at the Earth's surface~~ heat, moisture and trace substances are primarily dispersed ~~. Exchange with~~ following either the release at the surface or some altitude within the ABL or the entrainment from the free troposphere (FT) above. Exchanges with the FT takes place via ~~ejection and entrainment~~ entrainment and ejection processes (Stull, 1988). Horizontal variations in ABL dynamics stem

[1] ~~synonymous with the term *planetary boundary layer* (PBL), also commonly used~~

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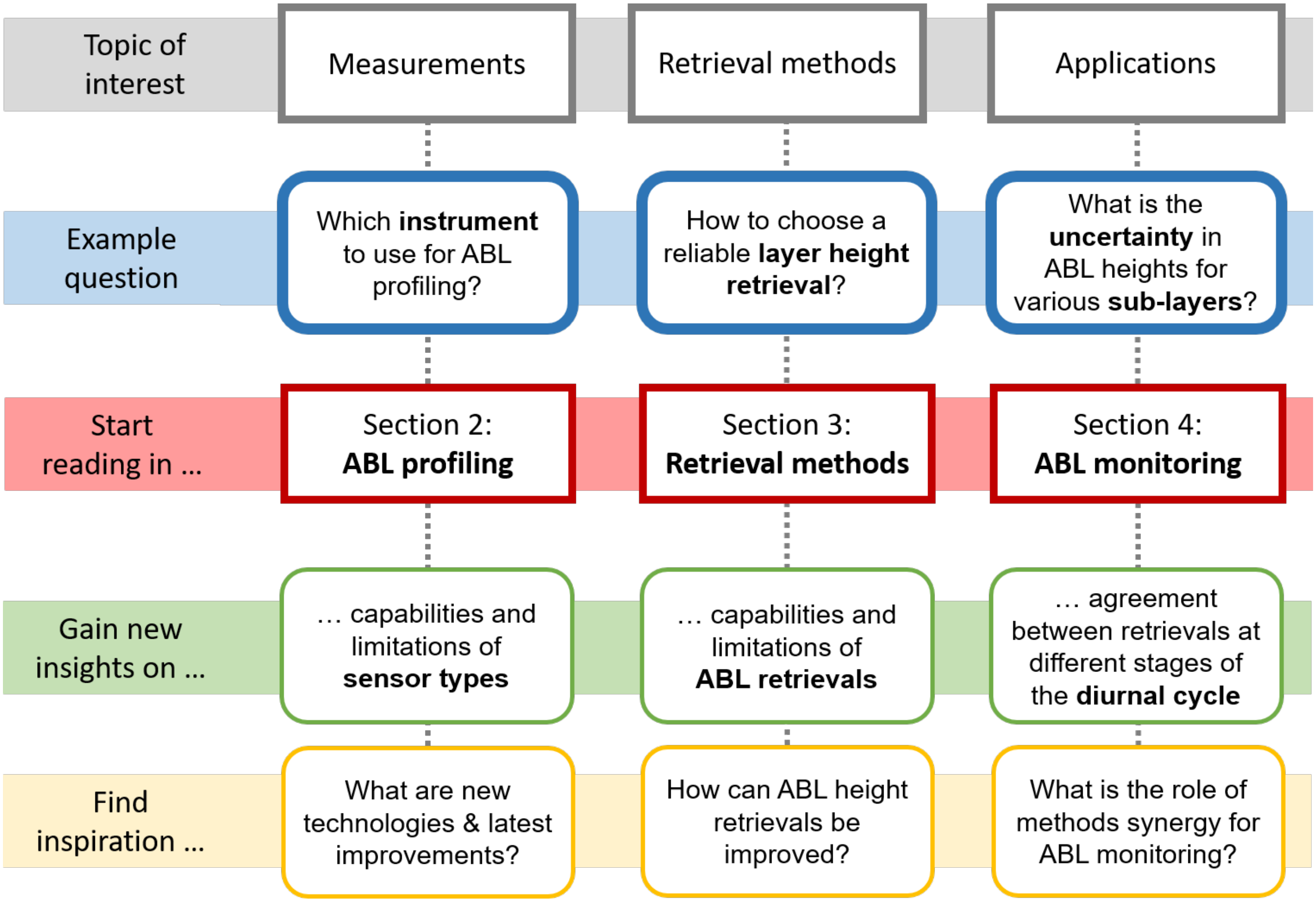

Figure 1. Entry points to this manuscript. The reader is invited to consult the respective section(s) related to their field of interest.

from a combination of synoptic atmospheric conditions (e.g., atmospheric stability, wind shear, cloud dynamics) and surface forcings (driven by contrasts in e.g., surface cover, roughness, topography) (Garratt, 1994; Seibert et al., 2000).

The height of the ABL ($ABLH$) is here considered the height above ground where the surface influence becomes low, i.e. the transition to the FT. Different sub-layers occur within the ABL depending on atmospheric stability. If surface-driven processes dominate over synoptic flow conditions on a warm, cloud-free day, the ABL tends to follow a textbook evolution (Figure 2) with a *convective boundary layer* (CBL) forming in the morning in response to solar heating of the ground and resulting turbulent heat fluxes. The height of the CBL ($CBLH$) increases during the morning and reaches its peak in the early afternoon when it extends over the whole ABL ($ABLH = CBLH$). Around sunset, radiative cooling of the surface induces the growth of a new layer near the ground, the *stable boundary layer* (SBL). At this time of reduced solar input and decaying buoyancy, the CBL breaks down and decouples from the surface, whereby converting into the *residual layer* (RL), now located above the SBL top ($SBLH$). The height of the RL top now coincides with the $ABLH$ ($ABLH = RLH$).

On the following day again, the RL is usually entrained into the newly forming CBL during morning growth. While neutral atmospheric stability usually dominates the RL, it is less frequent near the surface (Collaud Coen et al., 2014) but may still occur when shear production of atmospheric turbulence is strong (Nieuwstadt and Duynkerke, 1996).

In response to surface-atmosphere exchanges, cloud processes or synoptic-scale dynamics, the ABL sub-layers often deviate from this idealised concept. For example, over complex terrain, multiple layers often form in response to different mechanisms driving the ABL dynamics (De Wekker and Kossmann, 2015; Serafin et al., 2018). In cold seasons or over cold surfaces (such as snow and ice), the SBL ~~may~~ can also dominate during daytime ~~and no~~ leading to an absence of the RL ~~may form leading to~~ and consequently $ABLH = SBLH$ during both day and night. In the presence of a low-level jet (LLJ), the jet core (peak wind speed) defines the top of the surface-based shear layer acting as an upper bound for turbulent transport (Banta et al., 2006; Mahrt et al., 1979). The vertical profile of air temperature in the SBL often shows a characteristic surface-based temperature inversion (SBI), whose height ($SBIH$) can be very meaningful in restricting vertical dilution. ~~In the presence of a low-level jet (LLJ), jet core defines the top of the surface-based shear layer acting as an upper bound for turbulent transport (Banta et al., 2006; Mahrt et al., 1979)~~While vertical mixing mainly occurs in the lower levels of the temperature inversion, a combination of potential other processes such as radiative cooling, subsidence or horizontal advection shapes the depth and the magnitude of the SBI.

But also unstable conditions may persist at night where the surface remains relatively warm even after sunset (e.g., urban areas; (Barlow, 2014; Barlow et al., 2015; Pal et al., 2012)). In this case, no SBL is present at night. Still, a shallow mixing layer may form around sunset (decoupled from the RL) as the nocturnal surface buoyancy is now only driven by storage ~~(and potentially anthropogenic)~~ and anthropogenic heat fluxes and is hence weaker than the mixing during daytime. The term *mixing boundary layer* (MBL) ~~is a general term referring~~ generally refers here to the ABL ~~sub-layer~~ sublayer closest to the ground. Its height ($MBLH$) ~~can refer to~~ may indicate either $CBLH$ or $SBLH$, whichever is present at the given moment. The $MBLH$ terminology is applied when no information on atmospheric stability is available to differentiate between SBL and CBL.

~~When buoyancy-driven turbulence is present, exchange~~ Exchange between the CBL and the FT (or the RL) occurs via ~~two processes:~~ the penetration of the CBL thermals into the air aloft and the entrainment of warm and (in the absence of clouds) dry air into the CBL. As horizontal wind speeds are usually lower in the CBL compared to the FT or RL (Figure 2), wind shear at the $CBLH$ further generates mechanical turbulence that contributes to the entrainment. The *entrainment zone* (EZ) refers to this region of interaction around the $CBLH$ and its depth (EZD) is related to the contrasts between the air in the CBL and the above FT (or RL), respectively. The EZ is associated with ~~intermittent turbulence in time~~ temporally intermittent turbulence and a vertical decline in intensity of the turbulence (Gryning and Batchvarova, 1994). ~~EZD is greater when the temperature gradient between~~ The ABL ~~and FT is weak (AMS, 2017). At the top of the neutral RL or the SBL, the~~ transition to the FT is marked by a strongly positive temperature lapse rate, the *capping inversion* (CI). EZD is greater when the temperature difference between ABL and FT is weak (AMS, 2017). The CI often coincides with a sharp vertical decrease in specific humidity and significant vertical wind shear (Figure 2). The ABL-FT exchanges become more increasingly important over heterogeneous surfaces or complex topography (Lehning et al., 1998).

The interaction of clouds and ABL dynamics depends on the cloud type (Harvey et al., 2013). Cumulus clouds (Cu) forming at the CBL top can be understood as generating a deep EZ and thus, the $ABLH$ ~~may be~~ is located above the ~~*cloud base height* (CBH)~~cloud base height, i.e. somewhere within the Cu. Radiative cooling in stratocumulus clouds (Sc) induces top-down mixing from the cloud layer toward the surface during day and night (Hogan et al., 2009; Wood, 2012) so that $ABLH$ ~~may rather coincide~~ rather coincides with the cloud top. If deep convective clouds are present, e.g., cumulonimbus (Cb) before the occurrence of precipitation, the ABL may present higher relative humidity, greater instability, stronger temperature inhomogeneity and less wind shear (Zhang and Klein, 2010) so that it becomes challenging to define the $ABLH$.

Layers of gaseous species or aerosols (e.g., dust, smoke, ash) can be present in the FT, e.g., through long-range transport, volcanic eruptions or pyrocloud convection (Fromm et al., 2010; Lareau and Clements, 2016). The lofted layer may remain decoupled from the local ABL but can also be (partially) entrained (Granados-Muñoz et al., 2012; Bravo-Aranda et al., 2015). ~~Over heterogeneous surfaces or complex topography, the ABL-FT exchanges become more significant (Lehning et al., 1998). Several passive (e.g., flow blocking, flow channeling and lee waves) and active (thermally driven wind systems such as valley and slope winds or plain to mountain transport) mechanisms are involved in the ABL behavior and in vertical transport exchanges over complex terrain (De Wekker and Kossmann, 2015; Serafin et al., 2018). The ABL over mountainous terrain ($MoBL$) often exhibits a complex multilayered structure with successive stable and turbulent layers leading to three-dimensional patterns (Finnigan, 2003; **?**) that are influenced by micro- to mesoscale effects in time scales of up to one day.~~

2 Atmospheric boundary layer profiling

As stated by Beyrich (1997), profile observations should fulfill a series of requirements to adequately support the assessment of ABL dynamics and the detection of layer heights. Namely, they should (i) cover the full extent of the ABL (from the ground to the FT), (ii) have high vertical resolution of about 10-30 m, (iii) high temporal resolution of $\leq$1 h, and (iv) describe either the mixing itself or a result of mixing processes. We add that data with high temporal coverage (e.g., long time series) are necessary to determine variations in ABL dynamics at different temporal scales (synoptic, seasonal, annual, inter-annual) and measurements at multiple geographic locations enable horizontal variations to be assessed. Adequate atmospheric profiles (Sect. 2.1) can be captured by a series of different technologies (Sect. 2.2) that are increasingly operated in coordinated measurement networks (Sect. 2.3).

2.1 Profile variables characterising the atmospheric boundary layer structure

Different quantities provide insights ~~on~~ into ABL dynamics and can be analysed to derive the heights of the various ~~sub-layers~~ sublayers (Sect. 1.1). While *thermodynamic variables* capture atmospheric stability conditions at a given moment, *dynamic variables* describe the mixing processes induced by this stratification and *tracer variables* may portray the result of recent mixing processes (Table 1). Figure ~~**??**~~ 2 indicates how vertical profiles of selected exemplary atmospheric variables ~~change~~ evolve throughout the idealised evolution of the ABL on a cloud-free day~~(Figure 2)~~.

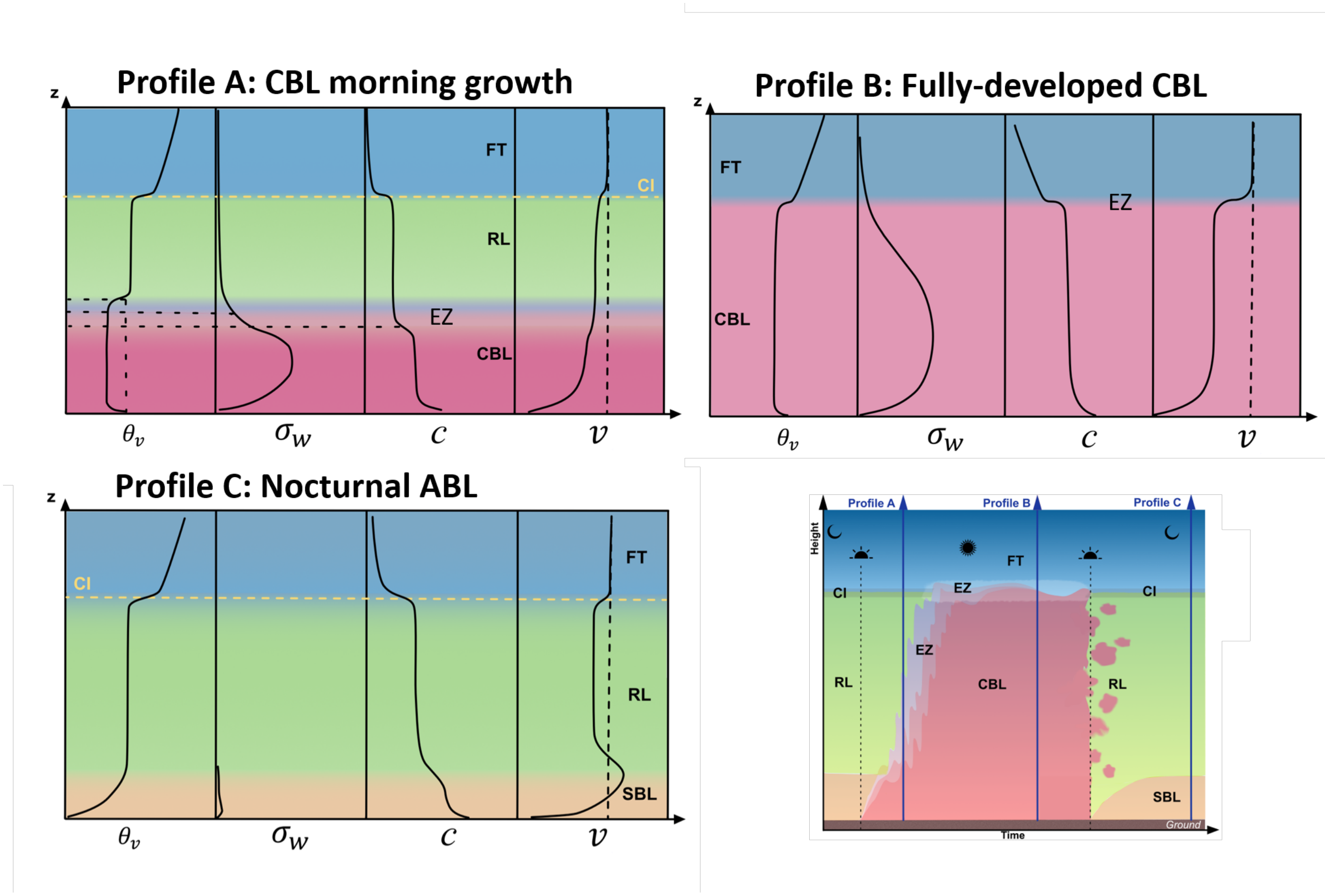

Figure 2. Idealised vertical profiles of exemplary atmospheric variables that are used to characterise thermodynamics (mean virtual potential temperature θ_v), dynamic and turbulent processes (vertical velocity variance σ_w, mean horizontal velocity v), and resulting distributions of atmospheric tracers (mean atmospheric constituent c) during the idealised diurnal evolution of the atmospheric boundary layer (~~ABL~~ABL), which is illustrated in the time-height sketch for an ABL over flat terrain on a cloud-free day. ~~Vertical dashed lines mark times of sunrise and sunset. Idealised vertical~~ Selected profiles ~~of different atmospheric variables~~ are shown at ~~times~~ three distinct moments, with Profile A: the morning growth of the convective boundary layer (CBL; pink shading), Profile B: early afternoon with a fully-developed CBL, and Profile C: nocturnal conditions with a residual layer (~~here marked by vertical~~ RL; green shading) above the stable boundary layer (SBL; orange shading) near the surface. A capping inversion (CI) separates the ABL from the free troposphere (FT; blue ~~arrows~~shading) ~~are shown~~ above. the entrainment zone (EZ) is a region of enhanced exchange between the CBL and the RL or FT, respectively. As the morning growth of the CBL (Profile A) is associated with high temporal variability of temperature, turbulence, and atmospheric constituents in ~~Figure ??~~the EZ, the temperature inversion, the reduction in vertical turbulent activity, and the vertical decrease in atmospheric constituents concentration may not always be located at the same height above ground, which is indicated by slightly changing colours and horizontal dashed lines. Idealised profiles and ABL sub-layer evolution adapted from De Wekker and Kossmann (2015); Beyrich (1997); Stull (1988).

~~Idealised vertical profiles of exemplary atmospheric thermodynamic (mean virtual potential temperature θ_v), dynamic (vertical velocity standard deviation σ_w, mean horizontal velocity v), and tracer (mean atmospheric constituent c) variables at three selected times during the diurnal evolution of the atmospheric boundary layer (ABL, Figure 2) with the free troposphere (FT) above and extent of sublayers indicated by the sidebar: (top) morning growth of the convective boundary layer (CBL) into the residual layer (RL), (middle) peak CBL development, and (bottom) nocturnal stable boundary layer (SBL). Yellow horizontal dashed line and shading mark the height of the ABL. Horizontal pink line in (bottom) indicates the height of the SBL. Horizontal pink lines in (top) indicate height of the CBL depending on the indicator used; note those height contrasts are here exaggerated and the exact relation between heights derived from different indicators is subject to ongoing research. Idealised profiles adapted from De Wekker and Kossmann (2015); Beyrich (1997); Stull (1988).~~

These variables can either be *measurement variables* that are somewhat defined by the observation technology and setup (e.g., radial velocity obtained by a Doppler wind lidar along its laser line-of-sight; Sect. 2.2.3) or *atmospheric variables* that describe a physical process or characteristic of the air rather independently of the observation technique. Some atmospheric variables are output directly by a certain sensor (e.g., air temperature measured with an in-situ thermometer of a radiosonde; Sect. 2.2.1), while others are retrieved during post-processing following methods of various complexity. Certain variables are calculated as a combination of multiple variables (e.g., potential temperature calculated from air temperature and atmospheric pressure, colour ratio determined from backscatter coefficient observed at two different wavelengths) or by applying higher order statistics (e.g., variance of vertical velocity) or both (e.g., turbulent kinetic energy calculated from variances of the three wind velocity components). Other variables require more complex retrieval algorithms, with a series of assumptions (e.g., retrieval of wind speed components from Doppler radial velocity) and even auxiliary information (e.g., retrieving air temperature from microwave radiometer brightness temperature).

Both atmospheric variables and measurement variables can be exploited for ABL height detection (Sect. 3). Those most commonly utilised, can be grouped by their physical relation to ABL dynamics (Table 1).

2.2 Measurement principles

A range of technologies (Table 2) is available to measure the quantities (Sect. 2.1) analysed for layer detection (Sect. 3). Atmospheric profile measurements can be achieved using tower-based or airborne in-situ sensors (Sect. 2.2.1) or with remote sensing techniques that again can be ~~air-borne~~airborne, space-borne or ground-based, respectively. Ground-based remote sensing profilers generally provide data at high temporal and vertical resolution and good sensitivity in the ABL. In the following, sensors are briefly introduced, grouped according to their characteristic output variables into profilers for thermodynamic variables or atmospheric trace gases (Sect. 2.2.2), wind and turbulence profilers (Sect. 2.2.3), as well as aerosol profilers (Sect. 2.2.4). For further technical details, the reader is referred to relevant textbooks (e.g., Emeis, 2010; Foken, 2021).

While some passive radiometer technologies capture thermodynamic profiles (Sect. 2.2.2), most approaches actively emit a signal which is then ~~record~~ recorded after its interaction with the probed ~~atmosphere.~~

~~In addition to the~~ atmospheric volume. Probably the most widely applied approach for ground-based atmospheric remote sensing is using laser technology. Depending on the instrument specifics, lidars can be used to measure profiles of meteorological

Physical meaning	Measurement ~~and atmospheric~~ variables	Atmospheric variables
thermodynamic processes		brightness temperature (T_b), air temperature (T), potential temperature (θ), virtual potential temperature (θ_v), relative humidity (RH)~~specific humidity (q)~~ , water vapour mixing ratio (r)
dynamic and turbulent processes	radial velocity (v_r)	refractive index structure parameter ($C_n{}^2$)~~radial velocity (v_r)**~~ , horizontal wind speed (U), velocity components of the wind vector (u, v, w), variances of the velocity components ($\sigma_u{}^2$, $\sigma_v{}^2$, $\sigma_w{}^2$), turbulent kinetic energy (TKE), eddy dissipation rate (ϵ)
tracers	~~attenuated backscatter coefficient (β_{att})~~ signal-to-noise ratio (SNR), carrier-to-noise ratio (CNR)	mass or number concentration of particles and gases (ρ or c)~~volume depolarisation ratio (δ)particle depolarisation ratio (δ_p)~~, attenuated backscatter coefficient (β_{att}), particle backscatter coefficient (β_p), particle extinction coefficient (α_p)~~particle backscatter coefficient (β_p) colour ratio signal-to-noise ratio (SNR)** carrier-to-noise ratio (CNR)**~~ , volume depolarisation ratio (δ), particle depolarisation ratio (δ_p), colour ratio

Table 1. Atmospheric variables analysed for detection of ABL heights are relevant for thermodynamic and dynamic processes or act as atmospheric tracers. ~~**Measurement variables (e.g., v_r, SNR)~~ Measurement variables provide information on the probed atmosphere but are strongly dependent on sensor characteristics or measurement setup. Depending on the measurement technology, variables are directly observed, retrieved from measurements or calculated. Note: humidity can also be interpreted as an atmospheric tracer but is here grouped with air temperature due to its importance for thermodynamic processes. Based on the variables listed here, other higher-order variables or parameters can be calculated (such as turbulent fluxes or Richardson numbers) that are valuable for characterising the ABL, especially where observations from multiple systems are available for synergy applications.

properties such as wind and turbulence (e.g. Doppler lidars), temperature (e.g. Raman lidar), humidity (e.g. differential absorption lidars), atmospheric gases (e.g. other inelastic lidars), or atmospheric aerosol particle characteristics (e.g. aerosol backscatter lidars). For all lidar systems, the incomplete optical overlap between the field of view of the receiver telescope and the emitted laser beam (Freudenthaler et al., 2018; Simeonov et al., 1999) can significantly increase the uncertainty in the first range gates. The part of the profile affected often extends over several hundred meters, but this varies significantly with instrument design (Haeffelin et al., 2012; Caicedo et al., 2020). And also the maximum range from which the signal is recorded depends on the instrument specifics (e.g. laser power, optics). In general, there is an inverse relation between the near-range and far-range capabilities of lidar systems. While high-power systems have a monitoring range of many kilometers (some reaching the stratosphere), they require an increasingly large telescope which then increases the blind-zone near the sensor. Low-power systems tend to have better performance in the near-range but with a more limited vertical extent. While also the vertical resolution of the recorded profile used to increase with laser power and vertical range extent, manufacturers increasingly apply oversampling procedures to the data products whereby increasing the number of range gates. Thick water clouds fully attenuate the lidar signal, so that the recorded information reduces to noise at some depth inside the cloud. Noise levels further increase due to the background signal induced by solar radiation.

While ground-based techniques ~~that~~ are focus of this review, some ABL information can be gathered by space-borne technologies, including aerosol lidars (e.g., Cloud-Aerosol Lidar and Infrared Pathfinder Satellite Observations (CALIPSO); Jordan et al., 2010; Liu et al., 2015a; Zhang et al., 2016), Doppler wind lidars (e.g., Atmospheric Laser Doppler Instrument (Aeolus-ALADIN); Straume et al., 2020; Flamant et al., 2016), or radio-occultation systems (Global Navigation Satellite System Radio Occultation (GNSS-RO); von Engeln et al., 2005; Ao et al., 2012; Xie et al., 2012; Chan and Wood, 2013; Basha and Ratnam, 2009). Satellite microwave and near-infrared passive observations also allow for the quantification of boundary layer water vapor even beneath uniform marine clouds (Millán et al., 2016). ~~Given their spatial coverage, space-borne data are particularly meaningful for global-scale analyses (Sect. **??**). However, satellite~~ Following the success of COSMIC, the promising COSMIC-2 mission was launched in 2019 to provide radio occultation data at even higher resolution through deeper tropospheric penetration (50% within 200 m of Earth's surface). These observation enable improved detection of the $ABLH$ and superrefraction at the top of the ABL (Ho et al., 2020; Schreiner et al., 2020). Satellite observations are less applicable for the detection of very shallow layers (e.g., Aeolus-ALADIN is not suitable for the monitoring of shallow layer conditions, Abril-Gago et al., 2021) or sublayer heights (such as $SBLH$ and RLH) given the degradation of profiles at low altitudes above the surface (Seidel et al., 2010; Xie et al., 2012) and the relatively coarse horizontal resolution (e.g., $\sim$200 km for GNSS-RO and $\sim$87 km for Aeolus)~~, which~~. The latter introduces additional uncertainty over coastal regions as well as in presence of complex terrain (Ao et al., 2012). Still, satellite-based ABL layer heights are very valuable, as they provide globally consistent estimates (Ho et al., 2015) whose seasonal cycle constitutes an important constraint on the behaviour of global atmospheric models (Chan and Wood, 2013; Liu et al., 2015a).

~~Ground-based remote sensing profilers generally provide data with the highest temporal and vertical resolution and best sensitivity in the ABL. The profiling sensors most~~ In the following sections the emphasis is placed on in-situ platforms and ground-based remote sensing instruments that are to-date commonly used to observe the ABL ~~are here briefly introduced, grouped according to their characteristic output variables into thermodynamic profilers~~ and can be considered the most promising candidates for extensive measurement network operations (Sect. 2.3). These are radiosoundings for in-situ profiling (Sect. 2.2.1; note that significant advances are expected for network operations of uncrewed areal systems), passive radiometers for temperature profiling (Sect. 2.2.2), Doppler wind lidars for profiling of wind and turbulence ~~profilers~~ (Sect. 2.2.3), ~~as well as lidars for the detection of aerosols and gases~~ and finally automatic lidars and ceilometers for aerosol profiling in the ABL (Sect. 2.2.4). ~~For further technical details, the reader is referred to relevant textbooks (e.g., Emeis, 2010)~~

During the discussion of respective sensor capabilities, it is obviously of interest to assess the agreement of observations obtained from different sensors in terms of absolute values. However, it should be kept in mind that layer height retrieval methods (Sect. 3) tend to exploit relative changes (such as vertical gradients) which means aspects such as sensor response time of in-situ measurements, or vertical resolution are generally also critical to consider.

2.2.1 In-situ profiling

In-situ sensors are attached to various kinds of platforms to gather atmospheric profile measurements. Instruments operated at multiple levels on **tall towers** are capable of capturing conditions in the lowest few hundred metres of the atmosphere based on profiles of temperature, humidity, wind, turbulence or atmospheric composition (Bosveld et al., 2020; Ramon et al., 2020; Neisser et al., 2002), often continuously at very high temporal and vertical resolution. A similar range of the atmospheric column can be probed by instruments hosted on **tethered balloons** (Keller et al., 2011; Spirig et al., 2004), however, the latter are still mostly operated manually during dedicated field campaigns only. Such methods are sometimes used in conjunction with radon measurements (Griffiths et al., 2013; Williams et al., 2013).

Other airborne measurements of meteorological variables and atmospheric composition tend to reach higher atmospheric levels, including in-situ sensors attached to radiosonde balloons or on board of airplanes~~or **uncrewed aerial systems (UAS)**. UAS can gather data at very high temporal and vertical resolution often covering the full vertical extend of the ABL, however, they can not (yet) be operated fully autonomously and temporal coverage is often limited. Similarly, data from research aircraft flights (e.g., Guimarães et al., 2019) are scare. The air volume sampled by both UAS and research aircraft flights can be restricted by air traffic control regulations. **Networks of commercial passenger airplanes** gather atmospheric profile information more continuously. The initiatives *Aircraft Meteorological Data Relay*[2] (AMDAR) and *In-service Aircraft for a Global Observing System*[2] (IAGOS) collect e.g. temperature, humidity, wind speed, wind direction, and various atmospheric constituents during their flights whereby gathering vertical profile data near the airports during start and landing. Observation accuracy is generally similar to that of radiosondes (Berkes et al., 2017), however, the vertical resolution is lower and systematic biases have been reported (e.g., AMDAR air temperature bias of up to 0.5-1.0 K; Ballish et al., 2008). Further, the airplane flight paths are associated with a much greater horizontal displacement (∼10 km km^{-1}) than radiosondes (∼ 1 km km^{-1}; Rahn and Mitchel . Naturally, the temporal resolution of IAGOS and AMDAR profiles depends on the frequency of reporting airplanes starting or landing in the region of interest. AMDAR data have been applied successfully to study the ABL in regions with multiple busy airports in close vicinity, such as Los Angeles, USA, (Rahn and Mitchell, 2016), London, UK, (Kotthaus and Grimmond, 2018a) or Paris, France, (Kotthaus et al., 2020), while Petetin et al. (2018) derive generalised ABL profiles for Northern hemisphere mid-latitudes from a climatology of IAGOS profiles.~~

. **Radiosondes** are probably the most common data source used to derive $ABLH$ operationally. In-situ measurements of air temperature ~~, pressure, humidity, and wind speed and direction~~ and humidity are taken by sensors that are being lifted up by ~~an~~ a helium-inflated aerostatic balloon while atmospheric pressure, wind speed and direction are derived along the flight path via satellite tracking (e.g. GPS). The balloon ascent allows profiles to be recorded up to ∼35 km above ground level (a.g.l.) with high and nearly constant vertical resolution at the order of tens of meters. The sounding takes 1.5-2.0 h to reach the maximum

[2]

[2]

altitude before the balloon bursts (usually in the lower stratosphere). Typical uncertainties in radiosonde measurements are $\pm$0.2-0.6 K for air temperature, 6 % for relative humidity, and 0.4-1.0 m s^{-1} for horizontal wind speed (Bian et al., 2011; Dirksen et al., 2014; Renju et al., 2017). Lightweight sondes attached to smaller balloons (Elie Quentin Bessardon et al., 2019) are not always able to profile the entire troposphere, however, they usually ascend to heights above the $ABLH$. As they are technically easier to operate and ~~do not require specific~~ may not require the same level of security clearance they are particularly useful for ABL profiling in populated environments such as cities.

The main advantages of radiosonde data are: (i) observations of temperature, ~~pressure, humidity,~~ humidity, air pressure, wind speed and direction are collected simultaneously using the same measurement system; (ii) coordinated radiosonde ascents are available at a high number of launch sites worldwide (Sect. 2.3); (iii) data are transmitted via international communication networks with very short time delay which makes them well-suited for operational use; and (iv) time series extend for decades, making radiosondes especially valuable for climatology studies~~(Sect. ??)~~. It should be noted however, that only 177 sites worldwide (status 2021) meet the stringent requirements for climate monitoring (CIMO-TECO, 2018; Thorne et al., 2017; WMO, 2010).

The main shortcoming of radiosondes is their low temporal frequency. Most operational sites only launch the balloons twice daily at specified synoptic times (00 UTC, 12 UTC), with some up to four times daily. While these coordinated launches at synoptic times are required to take the extremely valuable *global snapshot* of the atmosphere, they generally limit the representation of the ABL diurnal evolution at a given place. Where the launch times occur e.g. during morning growth and/or evening decay of the CBL, diurnal minima or maxima may not be captured. Even during special field campaigns, 1.5-3.0 h is typically the closest interval between launches. This low temporal resolution hampers the investigation of the diurnal cycle of ABL ~~sub-layer~~ sublayer heights and the comparison of $ABLH$ maxima at different locations. Note that some radiosonde data products of routine ~~ascends~~ ascents limit the vertical information to standard, significant pressure levels for real-time dissemination and archiving. ~~If such data sets are being explored,~~ This often means details of the ABL structure ~~may be obscured.~~ are obscured.

Other specific problems that can result in systematic errors in derived ABL characteristics include humidity sensor uncertainties in cold and dry or cloudy conditions (Seidel et al., 2010; Wang and Wang, 2014) ~~,~~ and significant horizontal displacement of the balloon during the ascent (Schween et al., 2014). This drift means observations are affected by spatial variations in ABL dynamics which can be challenging for data analysis and interpretation. Some stations operate automatic launch systems that can introduce temperature and humidity uncertainties in the lowest altitudes (< 200 m) as sondes are located in climate–controlled chambers before being released into ambient air (Madonna et al., 2020). Site-dependent radar tracking uncertainties (Seibert et al., 2000) that have caused errors in the wind profiles at low ~~levels that can be caused by site-dependent tracking (Seibert et al., 2000), and significant horizontal displacement of the balloon (Schween et al., 2014)~~ altitudes are no longer a concern as GPS tracking is now used instead. Careful removal of discontinuities induced by changes to the operating system helps to harmonise long-term records (Madonna et al., 2022).

Uncrewed aerial systems (UAS) can gather data at very high temporal and vertical resolution often covering the full vertical extend of the ABL, however, they can not (yet) be operated fully autonomously and temporal coverage is often limited.

Similarly, data from research aircraft flights (e.g., Guimarães et al., 2019) are scarce. The air volume sampled by both UAS and research aircraft flights can be restricted by air traffic control regulations. **Networks of commercial passenger airplanes** gather atmospheric profile information more continuously. The initiatives Aircraft Meteorological Data Relay[2] (AMDAR) and In-service Aircraft for a Global Observing System[3] (IAGOS) collect several atmospheric variables (such as temperature, humidity, wind speed, wind direction, or various atmospheric constituents, depending on the measurement system) during their flights whereby gathering vertical ABL profile data near the airports during start and landing. Observation accuracy is generally similar to that of radiosondes (Berkes et al., 2017), however, the vertical resolution is lower and systematic biases have been reported (e.g., AMDAR air temperature bias of up to 0.5-1.0 K; Ballish et al., 2008). Further, the airplane flight paths are associated with a much greater horizontal displacement ($\sim$10 km km^{-1}) than radiosondes ($\sim$ 1 km km^{-1}; Rahn and Mitchel . Naturally, the temporal resolution of IAGOS and AMDAR profiles depends on the frequency of reporting airplanes starting or landing in the region of interest. AMDAR data have been applied successfully to study the ABL in regions with multiple busy airports in close vicinity, such as Los Angeles, USA, (Rahn and Mitchell, 2016), London, UK, (Kotthaus and Grimmond, 2018a) or Paris, France, (Kotthaus et al., 2020), while Petetin et al. (2018) derive generalised ABL profiles for Northern hemisphere mid-latitudes from a climatology of IAGOS profiles.

2.2.2 ~~Thermodynamic profiling~~Profiling of thermodynamic variables and atmospheric gases

Different ground based remote-sensing technologies are available to obtain vertical profiles of ~~air temperatureand humidity: Raman lidarsor~~ thermodynamic variables (temperature, water vapour) and/or other atmospheric gases. These include Raman lidars, differential absorption lidars (DIAL~~; see Sect. 2.2.4~~), radio-acoustic sounding systems (RASS), and radiometers.

~~The standard~~ ~~**RASS**~~**Raman lidar** ~~configuration combines a radar wind profiler with a source of acoustic pulses (e.g. , sodar) transmitted into the vertical beam of the radar (e.g., Emeis, 2010). The frequency of the acoustic pulses are varied in time and Bragg scattering of the radar signal occurs when the wavelength of the acoustic pulse is half the wavelength of the radar. The~~ systems transmit at one or multiple wavelengths and detect the Raman-shifted scattering by molecular excitation at other wavelengths, enabling the determination of the constituent of interest (Table 2), such as the water vapour mixing ratio (Wulfmeyer et al., 2010), the particle extinction coefficient (Ansmann et al., 1992) or air temperature using the rotational Raman technique (Behrendt et al., 2015). Raman lidars widely use Nd:YAG lasers at tens Hz typical repetition rates, with extremely high pulse energy of >1 J at the fundamental wavelength (1064 nm) and up to hundreds of mJ at the second (532 nm) and third (355 nm) harmonics. Depending on the laser repetition rate and pulse energy, temporal resolution ranges from seconds to minutes. Range resolution is defined by the speed of the data acquisition system (e.g., a 100 ns laser pulse length has a 15 m folded scatte , with very high resolution (7.5 m or even higher) possible. The most prominent limitation in the exploitation of Raman lidars is their limited temporal coverage. These systems are generally not operated continuously because Raman channels only provide usable results when the natural background light is low, i.e. at night. In addition, consumables of high-power lidars are expensive, so that most operators limit measurements to times when no low-level liquid-water clouds are present as these

[2]https://public.wmo.int/en/programmes/global-observing-system/amdar-observing-system
[3]https://www.iagos.org/

extinguish the lidar signal at very low altitudes. As a consequence, Raman lidars are rarely used to monitor ABL dynamics and studies focus on atmospheric layers at greater altitudes instead.

A differential absorption lidar (**DIAL**) transmits laser beams at two wavelengths exploiting the differential attenuation (Lammert and Bösenberg, 2006) to derive vertical profiles of water vapour (Behrendt et al., 2007) or trace gases such as CO_2 (Gibert et al., 2008), CH_4 (Robinson et al., 2015), ozone (Banta et al., 1998; Ravetta and Ancellet, 1998), or NO_2 (Piters et al., 2012) . Thanks to recent developments, compact DIAL systems are becoming increasingly available that allow for continuous water vapour profiling of the ABL, using a significantly lower pulse energy compared to the Raman lidar (Newsom et al., 2020).

RASS systems either combine a radar wind profiler with a source of acoustic signals (e.g., sodar) or a sodar system with a source of electromagnetic signals (Emeis, 2010; Foken, 2021). From the Doppler shift of the ~~Bragg-scattered radar signal provides~~ respective returned signal the speed of sound is measured as a function of altitude, from which the profile of virtual temperature can be deduced. The uncertainty in temperature can be < 0.5 K, provided a number of careful corrections are applied (Görsdorf and Lehmann, 2000). Temporal resolution depends on the application with 10 minutes averaging being typical. The vertical resolution of the profile depends on the length of the pulse transmitted~~by the radar~~, with RASS systems usually configured to have a resolution of 30-60 m. ~~Although it is possible to obtain measurements above 1 km, the maximum range is usually~~ As for many ground-based remote sensing instrument types, the capabilities to capture information in the near-range or greater altitude, respectively, depends on the specific RASS system characteristics. While sodar-based RASS or 1-GHz radar wind profilers with RASS capability reach their maximum range at about 500 m~~. However, as the acoustic signal disturbs both humans and animals RASS profiling is now rarely deployed despite its capability~~, measurements well above 1 km can be obtained with RASS systems using a radar wind profiler at about 500 MHz.

Two types of ground-based profiling radiometers measure the downwelling radiance naturally emitted by the atmosphere at selected band channels: microwave radiometers (~~**microwave radiometers (**~~**MWR**~~**)and**~~) and infrared spectrometers (~~**infrared spectrometers (**~~**IRS**~~)~~)). The measured radiance is internally converted to atmospheric brightness temperature (~~T_b~~Table 1). As T_b holds information on atmospheric thermodynamic conditions, further atmospheric variables (e.g., temperature, humidity, ~~water liquid~~ liquid water path and integrated water vapour content) can be derived, using retrieval methods aided by some *a priori* knowledge. The atmospheric variables obtained from MWR and IRS depend on the number and spectral range of the channels utilised by a given sensor.

In the 20-60 GHz frequency (0.5-1.5 cm wavelength) range, the atmospheric thermal radiance is mostly emitted by atmospheric gases (primarily oxygen and water vapour) and hydrometeors (mainly liquid water droplets). MWR operating at several channels in the 20-30 GHz and 50-60 GHz frequency bands observe temperature and humidity profiles, respectively. Vertical resolution of the obtained temperature profiles is higher in the lowest 2 km where most of the information content resides. For humidity profiles the information is spread along the vertical range with generally coarser resolution. Most common MWR profilers provide information on tropospheric temperature and specific humidity and the column-integrated liquid water content (Solheim et al., 1998; Westwater et al., 2004; Rose et al., 2005) at high temporal resolution ($\sim$1 min). When compared to nearby radiosonde ascents, MWR retrievals agree within 0.5-2.0 K root mean square ~~error (RMSE~~deviation (RMSD) for temperature (decreasing from surface ~~up~~upwards) and 0.2-1.5 g m^{-3} for absolute humidity. The mean ~~RMSE~~ RMSD value within the boundary layer is ∼0.8 K for the temperature retrievals (Liljegren et al., 2005; Cimini et al., 2006; Löhnert et al., 2009; Löhnert and Maier, 2012). Bias values between MWR and Raman lidar are within ±0.4 g kg^{-1} (or ±20 %) for water vapor mixing ratio measurements with RMSD < 1 g kg^{-1} (25-55 %) and within 0–1.2 K for temperature measurements with RMSD ∼0.6-1.8 K (at 5 min integration time; Di Girolamo et al., 2020). Bianco et al. (2017) find lower statistical differences against radiosonde data for RASS than for MWR.

IRS exploit high spectral resolution radiances measured in the thermal infrared spectrum to retrieve temperature and water vapor profiles in cloud-free air. The Atmospheric Emitted Radiation Interferometer (AERI) is a Fourier transform IRS operating in the thermal infrared range ~~(3000-520 cm^{-1} wavenumber, 3.3–19 µm wavelength) (Knuteson et al., 2004a, b)~~ (3000-520 cm^{-1} wavenumber, 3.3–19 µm wavelength; Knuteson et al., 2004a, b). It is specifically designed to record downwelling radiance at high spectral resolution (~~$0.5\ cm^{-1}$~~0.5 cm^{-1}). The observed radiance ~~observed~~ is processed to retrieve temperature and water ~~vapor~~ vapour profiles up to cloud base, and in addition cloud properties and trace-gas concentrations (Feltz et al., 2003; Turner and Löhnert, 2014; Turner and Blumberg, 2018), with a temporal resolution of 30 s. When compared to nearby radiosonde ascents, IRS retrievals agree within ∼1 K ~~RMSE~~ RMSD for temperature and ∼0.8 g kg^{-1} for water vapor mixing ratio (e.g., Blumberg et al., 2015; Wulfmeyer et al., 2015; Weckwerth et al., 2016). ~~Synergy of IRS with active remote sensing technologies, such as DIAL or Raman lidar temperature profiling, can help to achieve an improved accuracy of the moisture gradient across the entrainment zone (Smith et al., 2021).~~

MWR and IRS techniques both measure atmospheric natural radiation that is then inverted to estimate thermodynamic variables. Thermodynamic profiles from MWR or IRS have been demonstrated to be useful to estimate $ABLH$ (Cimini et al., 2013) and atmospheric stability indices (Feltz and Mecikalski, 2002; Wagner et al., 2008; Cimini et al., 2015). However, despite their similarities they provide partially complementary information. In general, IRS data have greater information content than MWR, resulting in higher vertical resolution for temperature and humidity profiles, and sensitivity to trace gases and cloud particle size. IRS also provides higher sensitivity to ~~low cloud~~ low-cloud liquid water path, though the signal saturates above ∼40 g m^{-2}. MWR again are only slightly affected by liquid water, which gives them an advantage in capturing profiles even within or above clouds (unlike IRS, which is limited to cloud base). Further can MWR be used within light precipitation (Cimini et al., 2011; Bianco et al., 2017) because the antenna is protected by a radome with hydrophobic coating and a continuous tangential air flow. Still, ~~periods under precipitation are usually excluded from analysis, as~~ the above measures ~~may~~ are generally not be sufficient under moderate to heavy precipitation ~~, thus degrading~~ when the quality of retrieved profiles is degraded and hence usually excluded from analysis.

The most prominent limitation of ground-based radiometric profiling is its low-to-moderate vertical resolution. The information content of ground-based radiometry on the vertical distribution of atmospheric thermodynamics resides in the differential absorption of multi-frequency and multi-angle observations. However, contributions from different layers to the observed T_b (i.e. the weighting functions, Westwater et al., 2004) show significant overlap, leading to substantial redundancy in the observations. Although the retrievals of atmospheric profiles from passive instruments like MWR and IRS are usually provided on fine vertical grids (e.g., ∼50, 100, and 250 m at <500 m, 500–2000 m, and >2000 m, respectively), this spacing should not be confused with the actual vertical resolution, which by definition is the minimum distance at which differences in the vertical profile ~~can be appreciated. More than one method is~~ are resolved. Several methods are used to quantify the vertical resolution of radiometric profiling, including, e.g., the degrees of freedom for signal ~~(DFS)~~(Löhnert et al., 2009), the inter-level covariance ~~, and averaging kernels. Löhnert et al. (2009) showed that for a generic MWR operating in the 20-60 GHz range the DFS, i.e. the number of independent levels that can be retrieved, range from 1 to 4 for both temperature and humidity profiles. In more detail, for temperature profiles the DFS strongly depend on the number of elevation observations (from DFS 2 for zenith only to DFS 4 for some 10 elevation angles) but only slightly on atmospheric conditions, while for humidity the DFS are almost independent of the number of elevation observations but depend noticeably on the water vapor content (from DFS 1-2 at mid-latitude to DFS 2-3 at tropics). In general, DFS are higher for IRS than for MWR, but also more dependent on atmospheric conditions. For temperature (humidity) IRS DFS vary from 4-9 (4-10) in mid-latitudes to 3-5 (3-7) in the tropics.~~

~~Liljegren et al. (2005) used the inter-level covariance (ILC) to quantify the vertical resolution of radiometric profiles. For ground-based MWR profilers, ILC typically decreases almost linearly with height (z): for temperature, ILC ranges from ∼100 m at the surface to ∼6 km at 10 km (ILC ∼0.5·z + 0.1 km); for absolute humidity, ILC increases from ∼400 m at the surface to ∼3.5 km at 10 km (ILC ∼0.3·z + 0.4 km) (Cimini et al., 2006). ILC for temperature profiles can be kept within 1 km throughout the vertical domain by adopting an optimal estimation inversion method initialized with NWP analysis (i.e. 1DVAR, Cimini et al. (2006)). Blumberg et al. (2015) use the vertical resolution definition proposed by ?, i.e. the reciprocal of the diagonal of the averaging kernel matrix, to compare vertical resolutions of MWR and IRS. For temperature profiles they report~~ (Liljegren et al., 2005), and or averaging kernels (Blumberg et al., 2015). Using the latter, temperature profiles show a vertical resolution ~~that varies~~ varying linearly with height as by a factor of ∼2 for MWR and ∼1.4 for IRS, respectively. ~~For example, at 1.5 km the temperature vertical resolution is ∼2 km for IRS while ∼3 km for MWR.~~ The vertical resolution for the water vapor mixing ratio is less regular, but still roughly linearly with height (∼z+1 [km]~~for IRS~~).

~~Di Girolamo et al. (2020) reports results from an inter-comparison effort involving water vapor and temperature sensors, such as one MWR, two ground-based Raman lidars, one airborne water vapor DIAL, as well as airborne in-situ sensors (radiosondes, aircraft). At 5 min integration time, bias values between MWR and Raman lidar are within ±0.4 g kg^{-1} (or ±20 %) for water vapor mixing ratio measurements, while root mean square deviations are smaller than 1 g kg^{-1} (25-55 %). For temperature measurements, bias values are within 0–1.2 K and root mean square deviation values are 0.6-1.8 K. Similarly, Bianco et al. (2017) reports results from an inter-comparison effort involving two MWR and two RASS, as well as radiosondes and instrumented tower. For the layer of the atmosphere covered by the RASS, results show lower statistical differences with respect to radiosondes for RASS than the MWR. Methods to combine MWR and/or IRS with active instruments, such as DIAL and RASS, have also been proposed (e.g., Turner and Löhnert, 2021; Djalalova et al., 2021).~~

To summarise, passive radiometers provide better coverage of temperature and humidity profiles compared to ~~research~~ Raman lidars because they can gather data continuously. But also DIAL systems increasingly provide continuous profiles of water vapour or other gases in the ABL. Vertical resolution is greater for IRS between 0.5-2.0 km and greater for MWR above 4 km; IRS and MWR provide partially complementary information despite their substantial similarities, given the higher vertical information content of IRS in the ABL and the capability of the MWR to gather information within and above clouds and ~~within~~ during light precipitation. Synergy of MWR and/or IRS with active remote sensing technologies such as DIAL or

Raman lidar can improve data quality (e.g., Turner and Löhnert, 2021; Djalalova et al., 2021), e.g. achieving a more accurate representation of the moisture gradient across the entrainment zone (Smith et al., 2021).

2.2.3 Wind and turbulence profiling

Several technologies allow for the vertical profiles of mean wind speed, direction and turbulence to be captured (Foken, 2021), including sodars, radar wind profilers (RWP) and Doppler wind lidars (DWL). Where profiles of both turbulence and temperature fluctuations (e.g., from RASS; Sect. 2.2.2) are observed, profiles of turbulent heat fluxes can be obtained (Engelbart and Bange, 2002; Behrendt et al., 2020).

Sodars send out pulses of sound to probe the atmosphere. The sodar technique is based on fluctuations in the refractive index of the air (Sect. 2.1) and the amplitude of the return signal is related to the ~~structure function for temperature (CT^2; Singal, 1997; Bradley, 2007)~~refractive index structure parameter (${C_n}^2$; Singal, 1997; Bradley, 2007). Based on these, turbulent structures in the ABL can be characterised ~~(Emeis et al., 2008; Kramar et al., 2014)~~(Emeis et al., 2008; Kramar et al., 2014; Beyr . Compared to most other remote sensing profiling systems, sodars have a particular advantage in being capable of sensing close to the instrument, typically within 20 m~~, hence very~~ . Very shallow ABL can also be measured even in challenging polar locations (Kouznetsov, 2009), especially when combined with sonic anemometers (Argentini et al., 2005). ~~Their main~~ This good near-range capability goes along with a rather limited range extent to about 1 km which is linked to the considerable sensitivity of the system to environmental noise. Wind and turbulence derived from sodar observations are severely affected by precipitation as the fall speed of the precipitation disturbs the signal but also water on the antenna tends to increase retrieval uncertainty. Another disadvantage is that the sound signal can often be a disturbance for humans and animals which makes ~~them~~ it difficult to operate sodars continuously in many ~~locations~~environments.

RWP operate on Doppler technology, either in the the very-high frequency (VHF) domain (20-300 MHz) or ultra-high frequency (UHF) domain (0.4-2 GHz) with boundary layer RWP usually around 1 GHz (*L-Band*). UHF RWP are better suited for probing the ABL thanks to their higher vertical resolution and lower cost. An electromagnetic pulse is emitted towards the zenith and ~~two to four~~ 2-4 off-zenith directions (15°-tilted~~, off-zenith directions~~). The angle can be achieved with different antennas or with a single phased-array antenna. In the UHF band, the return signal intensity depends mainly on humidity and temperature gradients in the atmosphere. It is recorded and analyzed in real-time by the system: a succession of coherent averaging and noise filtering steps are followed by a fast Fourier transform (FFT). The frequency spectrum obtained for each range gate is characterized by four moments: noise level, signal power, spectral width and Doppler shift. By combining the Doppler shift of the three beams, mean wind speed and wind direction are calculated at each range gate (Ecklund et al., 1988). Vertical resolution is at the order of 100-400 m, depending on the measurement setup.

The main advantage of RWP is their capability to operate under all weather conditions at moderate cost. They even provide useful information inside cloud or fog layers and when aerosol concentrations are very low, an advantage over most lidar systems. Large errors in RWP profile data are mostly caused by larger objects, such as birds (Lehmann and Teschke, 2008). ~~RWP complement DWL observation capabilities and, provided~~ Provided suitable scan patterns, averaging strategies and quality control, the uncertainties and biases in RWP profiles are comparable to DWL observations. With less than 100 RWP operated worldwide (Sect. 2.3), their limited number is a clear disadvantage when it comes to spatial coverage.

~~**DWL** are active remote sensing systems similar to aerosol backscatter lidars (Sect. 2.2.4).~~ There are two types of ~~DWL~~**DWL**: one uses the molecular backscatter component and applies narrow-band spectral filters to measure the frequency shift while the other type (*heterodyne*~~*Doppler lidar*~~ Doppler lidar) uses the aerosol-particle backscatter component and coherent mixing with a reference beam to detect the ~~Doppler shift~~ slight Doppler shift in frequency between the emitted ~~and backscattered radiation~~pulse and backscattered return. All DWL exploit the Doppler shift along the line-of-sight, or *radial*, to measure the *radial Doppler velocity*. Given their negligible terminal fall velocities, backscattering aerosol particles and cloud droplets are ideal tracers to track the wind motion. Ground-based commercial DWL capable of probing the full depth of the ABL typically use the heterodyne principle. They generally operate at wavelengths between 1.5-2.0 μm, taking advantage of components developed for the telecommunication industry. Note that attenuated backscatter (Sect. 2.1) can also be retrieved from DWL observations, if the instrument telescope function is accounted for (Pentikäinen et al., 2020).

Heterodyne DWL work with continuous-wave technology or by emitting short-laser pulses. ~~For continuous-wave heterodyne systems, the radiation transmission is continuous and the range information is obtained through varying the focus of the telescope.~~ The maximum range for continuous-wave DWL systems is limited to about 250 m as the range-weighting function becomes very broad beyond this distance (Kavaya and Suni, 1991). Pulsed DWL systems emit very short pulses of radiation and the range information is obtained from the round trip time between the transmit pulse and the received echo. ~~The~~ Their maximum unambiguous range ~~is greater than for continuous-wave systems and~~ depends on the pulse repetition frequency ~~. For example, a pulse repetition frequency of about~~ (e.g. 15 kHz pulse repitition frequency corresponds to a maximum ~~unambiguous range of about~~ range of $\sim$10 km.~~Wind velocity information is obtained through measuring the slight shift in frequency between the outgoing pulse and the backscattered return. This Doppler frequency shift can be measured directly, using different types of interferometers such as Fabry-Perot, Mach-Zehnder, or Michelson interferometers. DWL using this so-called *direct-detection* usually operate at visible and UV wavelengths (e.g., 532 or 355 nm) and are optimised towards backscattering from molecules (e.g., **?**). The frequency shift can also be measured using the heterodyne principle, where the return signal is mixed with a stable local oscillator signal of known frequency and converted to an electrical signal by a quadratic detector (**?**). Since the frequency of the local oscillator signal is known, the original frequency shift can be determined from the measured heterodyne or 'beat' frequency. Given their negligible terminal fall velocities, backscattering aerosol particles and cloud droplets are ideal tracers to track the wind motion. All DWL exploit the Doppler shift along the line-of-sight, or *radial*, to measure the *radial Doppler velocity*. Note that attenuated backscatter (Sect. 2.1) can also be retrieved from DWL observations, if the instrument telescope function is accounted for (Pentikäinen et al., 2020)~~) and is greater than for continuous-wave systems. Pulsed DWL are available at different frequencies, with some providing high-resolution data only over a few hundred meters.

Pointing to nadir (zenith), the radial Doppler velocity observed from aerosol or cloud droplets is the vertical air motion w; for larger particles the observed radial Doppler velocity is the sum of the vertical air motion speed w and the fall velocity of the particles. For beams tilted away from zenith, the radial Doppler velocity contains components of both the horizontal wind and the vertical motion. Combining scans from multiple directions permits the horizontal wind component to be derived using trigonometry under the assumption of horizontal homogeneity of the wind field in the observed volume (Banta et al., 2013; Päschke et al., 2015; Teschke and Lehmann, 2017). Combining observations from multiple instruments permits direct retrievals of the three-dimensional wind vector and its fluctuations (Sathe and Mann, 2013). Comparisons with sonic and cup anemometers on towers or masts show that winds can be derived from DWL with sufficient accuracy for wind energy applications (Peña et al., 2008; Pichugina et al., 2012). ~~In~~ Under ideal conditions, the DWL precision is within the uncertainty of the anemometer measurements used as a reference (Gottschall et al., 2012).

If winds are sampled at very high temporal frequency, higher order moments, such as velocity variances (Sect. 2.1) and even skewness, kurtosis, turbulent kinetic energy (TKE) or eddy dissipation rate (ϵ) can be determined (e.g., Cohn, 1995). Being direct measures of turbulence, TKE and ϵ are best suited for fair site and instrument inter-comparisons and can be obtained from various scan strategies, including vertical stare, multi-beam, conical scanning (Banakh and Smalikho, 1997; Banakh et al., 2010; Sathe et al., 2015; Bonin et al., 2017; Smalikho and Banakh, 2017; Yang et al., 2020), or a combination of scan types (Bonin et al., 2018). TKE can also be obtained by scanning at the specific elevation angle of 35.5° (Eberhard et al., 1989). These methods usually include measurements of the wind profile to provide the horizontal length scales required (O'Connor et al., 2010). ~~Combining observations from multiple instruments permits direct retrievals of the three-dimensional wind vector and its fluctuations (Sathe and Mann, 2013). DWL retrievals of the~~ Instrument noise can play a role when measuring turbulence statistics if high-frequency variations are introduced into the signal (Tucker et al., 2009; Lenschow et al., 2000). Gravity waves and other larger-scale atmospheric motions can hamper the simple interpretation of velocity fluctuations as a proxy for turbulent motion. Methods are under development to diagnose and account for such situations (Banakh and Smalikho, 2016; Bonin et al., 2018). In general, it is crucial to assess the implications of noise filtering, sampling frequency, integration time, and measurement volume on turbulence observations (Bonin et al., 2017; Pichugina et al., 2008). Also physical quantities describing atmospheric turbulence derived from DWL observations have been successfully evaluated against data gathered by sonic anemometers on masts (Bonin et al., 2016; Bodini et al., 2018; Bonin et al., 2018), tethered-balloons (Frehlich et al., 2008; O'Connor et al., 2010), radiosondes (Tucker et al., 2009), or a combination of these (Wildmann et al., 2019). For a review on pulsed DWL including descriptions of the various scan strategies, the reader is referred to e.g., Liu et al. (2019b).

The intrinsic uncertainty in the measured Doppler radial velocity is directly related to the DWL ~~SNR~~ carrier-to-noise ratio (Rye and Hardesty, 1993; O'Connor et al., 2010). As the latter depends on both the lidar system and the aerosol load of the atmosphere, uncertainty estimates should take into account the sampling strategy and potential instrument-specific corrections (Manninen et al., 2016; Vakkari et al., 2019). Increased uncertainties have been reported in pristine conditions such as the Arctic (e.g., Hirsikko et al., 2014). As for all lidar systems~~(Sect. 2.2.4)~~, the incomplete optical overlap of DWL usually means data in the lowest range gates need to be treated with caution and are often discarded from analysis. For DWL that are able to gather profiles up to several kilometres range and are hence able to capture $ABLH$ peaks during deep afternoon convection this *blind zone* may hinder the assessment of very shallow ABL sublayers. ~~Other~~ Here *mini*DWL systems ~~can observe wind profiles also in~~ that focus on the very near range at high vertical resolution ~~but then have the trade-off of a maximum range limited to several hundred metres (generally <1 km)~~offer valuable sensor synergy. The issue of incomplete optical overlap in the near range can also be somewhat overcome by scanning DWL systems ~~that are able to combine multiple scan patterns. By~~

~~adding low–level~~ because a combination of low-level scanning strategies to higher elevation scan patterns ~~, the vertical extent of the sampled wind profile can cover practically~~ means wind profiles are sampled throughout the entire ABL extent (Banta et al., 2006; Pichugina and Banta, 2010; Vakkari et al., 2015).

~~Instrument noise can play a role when measuring turbulence statistics if high-frequency variations are introduced into the signal (Tucker et al., 2009; Lenschow et al., 2000). Gravity waves and other larger-scale atmospheric motions can hamper the simple interpretation of velocity fluctuations as a proxy for turbulent motion. Methods are under development to diagnose and account for such situations (Banakh and Smalikho, 2016; Bonin et al., 2018). In general, it is crucial to assess the implications of noise filtering, sampling frequency, integration time, and measurement volume on turbulence observations (Bonin et al., 2017; Pichugina~~

~~Precipitation can have a similar impact,~~ DWL can operate under all weather conditions at high temporal (<1 min) and vertical (<100 m) resolution, however, thick water clouds usually fully attenuate the signal so little information can be obtained above the cloud base. Precipitation can cause significant uncertainties in the wind and turbulence retrievals since rapid variation in terminal fall velocities for different sizes of large precipitation particles ~~, such as drizzleor rain dropsand most ice particles, will manifest itself~~ (drizzle, rain drops, ice particles) manifest themselves as vertical velocity fluctuations that ~~look similar to turbulence. In other words, significant terminal fall velocity superimposed on the air motion of the measurement volume will impart~~ resemble turbulence. This imparts biases if not accounted for. Methodologies using the associated variations in the signal backscattered from precipitation particles are being developed to identify such cases. Also, rainfall on the antenna reduces measurement accuracy.

~~DWL can operate under all weather conditions at high temporal (<1 min) and vertical (<100 m) resolution.~~ The resolution, spatial extent, and accuracy of the retrieved wind information depends on the instrument model~~and~~ , the scan strategy, and the state of the atmosphere. As for aerosol lidars (Sect. 2.2.4), the strength of the backscattered signal increases aerosol load and relative noise levels can be high were little aerosol is present. A major advantage of DWL with scanning capabilities is that a series of different measurement setups can be alternated to gather optimised sampling strategies for several advanced data products simultaneously. This proves valuable not only for the detection of ABL sublayer heights but also for in-depth characterisation of ABL dynamics (Sect. 4).

2.2.4 Aerosol ~~and gas~~ profiling

~~Lidar systems are active remote sensing instruments that emit short and intense laser pulses which are then scattered by aerosol particles, droplets, and molecules. A fraction of this radiation is backscattered towards the instrument and collected by a telescope. A set of optical devices lead the radiation signal to the optical detectors, converting it into an electrical signal. The round trip time of each emitted laser pulse determines the distance between the lidar and altitude at which the radiation has been backscattered, allowing the generation of vertical profiles of backscattered signal. The backscattered lidar signal is proportional to the concentration and size of particle scatterers. Thus, the recorded signal increases with the amount of atmospheric scatterers in form of aerosol particles and cloud/rain droplets. Thick water clouds fully attenuate the lidar signal, so that the recorded backscatter reduces to noise at some depth inside the cloud. Noise levels further increase due to the background signal induced~~

~~by solar radiation.~~ Ground-based lidar systems available for the profiling of aerosols ~~and gases~~ differ greatly in laser power and wavelengths utilised (Foken, 2021). It can be generally differentiated between high-power lidar systems and the comparatively low-power automatic lidars and ceilometers (ALC).

~~High-power lidars include aerosol research lidars, Raman lidars, and DIAL. Aerosol research and Raman lidars widely use Nd:YAG lasers at tens Hz typical repetition rates, with extremely high pulse energy reaching of >1 J at the fundamental wavelength (1064 nm) and up to hundreds of mJ at the second (532 nm) and third (355 nm) harmonics. Depending on the laser repetition rate and pulse energy, temporal resolution ranges from seconds to minutes. Range resolution is defined by the speed of the data acquisition system ((e.g., a 100 ns laser pulse length has a 15 m folded scattering length, Weitkamp, 2005), with very high resolution (7.5 m or even higher)possible.~~ ~~**Raman lidar**~~ ~~systems transmit at one or multiple wavelengths and detect the Raman-shifted scattering by molecular excitation at other wavelengths, enabling the determination of the constituent of interest, e.g., water vapour mixing ratio (Wulfmeyer et al., 2010), the particle extinction coefficient Ansmann et al. (1992) and the temperature profile (Sect. 2.2.2)using the rotational Raman technique (Behrendt et al., 2015).~~ ~~**DIAL**~~ ~~transmit laser beams at two different wavelengths exploiting the differential attenuation (Lammert and Bösenberg, 2006) to derive vertical profiles of water vapour (Behrendt et al., 2007) or trace gases such as CO_2 (Gibert et al., 2008), CH_4 (Robinson et al., 2015), ozone (Banta et al., 1998), or NO_2 (Piters et al., 2012)~~The latter is a collective term that refers to both ceilometers which traditionally focused on cloud base height estimation and those backscatter lidars primarily designed to continuously provide aerosol profile information (such as micro pulse lidars; MPL). Capabilities and limitations of high-power aerosol lidars have been outlined for the Raman lidar (see description in Sect 2.2.2), a research-grade lidar which is able to sample water vapour and at times temperature, in addition to aerosol properties (Table 2).

ALC are compact, simple backscatter lidars which operate at wavelengths mostly in the infrared or visible spectral region ~~,~~ (e.g. ~~,~~532 nm, 808 nm, ~910 nm, 1064 nm are common wavelengths). ALC record the attenuated backscatter (Sect. 2.1) signal, which often needs to be absolutely calibrated during post-processing (Wiegner and Geiß, 2012; Hopkin et al., 2019). While most ALC are monochromatic, few models with multiple ~~wavelength~~ wavelengths do exist and first sensors with depolarisation capabilities start to emerge. ~~Typical ALC are the micro pulse lidars (MPL) and ceilometers, originally designed as cloud base height (CBH) recorders.~~ CBH is the standard output variable for all ALC in addition to the attenuated backscatter profiles. Retrievals of ABL sublayer heights are also increasingly incorporated ~~into the ALC firmware versions~~by the manufacturers.

The most striking disadvantage of ALC compared to high-power lidars is ~~the comparatively lower~~ their comparatively low SNR. Given the latter not only depends on atmospheric composition but is largely determined by the laser power and optics of the lidar system (Heese et al., 2010), data from ~~DIAL and Raman~~ high-power lidars are often able to capture more details of the atmosphere vertical structure and high-quality information can be obtained over a greater vertical extent. ALC performance is reduced in pristine environments where aerosol load is low or at elevated heights above the sensor. But also among ALC the SNR capabilities vary greatly ~~(Caicedo et al., 2020; Kotthaus et al., 2016, 2020)~~ (Caicedo et al., 2020; Kotthaus et al., 2016) due to the wide range of models available from various manufacturers. ~~ALC performance may be especially limited in pristine environments where aerosol load is low or at elevated heights above the sensor (e.g., deep ABL development)~~It can be generally differentiated between ALC that provide high-SNR observations and those with rather low-SNR (Kotthaus et al., 2020). While data from ~~high-power lidars and~~ high-SNR ALC can usually be analyzed at the recorded temporal resolution, averaging was found to improve the SNR of low-SNR ALC (e.g., Markowicz et al., 2008; Stachlewska et al., 2012; Lee et al., 2019; Mues et al., 2017; Min et al., 2020; Caicedo et al., 2020; Tsaknakis et al., 2011). It should be noted that some instrument-related artifacts have been detected that may be associated with specific hardware or firmware versions (Kotthaus et al., 2016; Kotthaus and Grimmond, 2018a).

~~For all lidar systems, the incomplete optical overlap between the field of view of the receiver telescope and the emitted laser beam (Freudenthaler et al., 2018; Simeonov et al., 1999) can significantly increase the uncertainty in the lowest several hundred meters. The extent of the profile affected, varies with instrument design (Haeffelin et al., 2012; Caicedo et al., 2020) .~~ High-power lidars have a significant blind zone while ALC usually reach full optical overlap at lower levels, giving them an advantage in monitoring shallow ABL ~~layers~~sublayers. Although most ALC manufacturers supply optical overlap correction functions (at times specific to the individual sensor) more complex correction models can be necessary to dynamically account for variations in the overlap function (e.g., dependent on the instrument internal temperature; Hervo et al., 2016; Geiß et al., 2017).

~~The most prominent limitation in the exploitation of Raman and research lidars is their limited temporal coverage as they are generally not operated continuously as Raman channels can usually only be exploited when the natural background light is low, i.e. at night. In addition, as consumables of high-power lidars are expensive, most operators of limit measurements to times when no low-level liquid-water clouds are present which would completely extinguish the lidar signal at very low altitude.~~

~~As ALC can be operated continuously , and~~ ALC are usually operated continuously as they work autonomously under all weather conditions with very low maintenance~~, their~~ . Their data have a much greater temporal coverage than those collected by ~~research lidars which often focus on specific periods of interest. Finally, the~~ high-power lidars that are mostly limited to specific research infrastructures (Sect. 2.3) and usually do not operate continuously (as mentioned for the Raman lidar; Sect. 2.2.2), although the number of systems with 24 h-operation is increasing. A clear strength of ALC is their unprecedented spatial distribution (Sect. 2.3). Aerosol profile information obtained from lidar systems can be analysed using aerosol-based ~~ABL height~~ retrievals (Sect. 3.3) to obtain heights of the ABL and its sublayers.

2.3 Profiling Sensor Networks

Profile data of the atmospheric boundary layer (Table 1) gain value when gathered by coordinated and harmonised measurement networks as these add information on variations in the horizontal spatial domain. High-quality ABL network data not only provide unprecedented details for process studies but also show great potential for the advancement of NWP via data assimilation (Illingworth et al., 2019; Martinet et al., 2020; Tangborn et al., 2021). Mobile platforms equipped with multiple instruments can be a powerful addition during intensive observation periods (Wagner et al., 2019).

While radiosonde stations have been organised in coordinated networks for decades, collaborative measurement networks of RWP, DWL, MWR and ALC are now also emerging (Figure 3) ~~. Mobile platforms equipped with multiple instruments can be a powerful addition during intensive observation periods (Wagner et al., 2019)~~as off-the-shelf commercial instruments can now be deployed for unattended, continuous operations, providing atmospheric profile observations in nearly all-weather

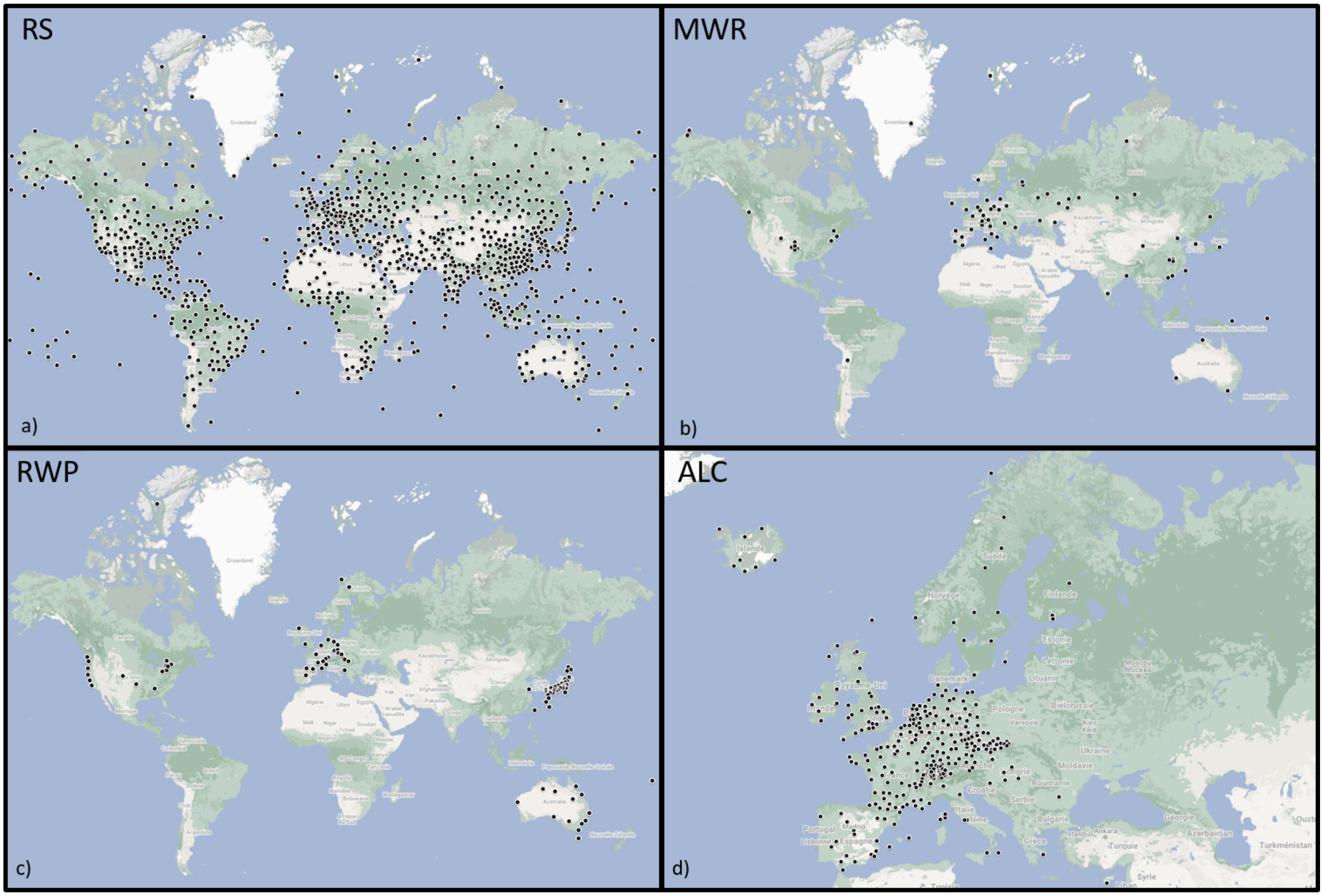

Figure 3. ~~Operational~~ Selected operational networks of ~~selected profile~~ profiling stations ~~in~~ (status December 2021): a) global distribution of radiosonde stations (RS) [WMO], b) global distribution of microwave radiometers (MWR) [MWRnet,MTP-5, RPG], c) global distribution of radar wind profilers (RWP) [JMA, NOAA, E-PROFILE], and d) European distribution of Automatic ~~Lidar~~ Lidars and Ceilometers (ALC) [E-PROFILE]. Note that this is by no means a comprehensive representation of all profiling instruments being operated. Additional networks do exist but meta data, such as station locations are not always be easily accessible. Backgroud map © Google Maps 2022.

conditions (Sect. 2.2). DIAL and Raman lidars are mostly organised in research networks, such as ACTRIS/EARLINET[4] (Pappalardo et al., 2014), NDAAC[5], or PollyNet (Baars et al., 2016), providing observations of the full troposphere and even lower stratosphere. However, as these sensors are less autonomous compared to MWR, RWP, DWL, or ALC, spatial coverage tends to be lower for these networks. Other ground-based remote sensing profiling technologies (e.g. sodar) are to-date operated less continuously and fewer stations for the reasons outlined above (Sect. 2.2).

[4] https://www.earlinet.org
[5] https://www.ndsc.ncep.noaa.gov/

Worldwide there are ∼1300 **radiosonde** launch sites (Figure 3a, WMO, 2017), with ~~100-200~~ ∼800 stations making observations ~~once or twice per day~~at least once but mostly twice daily. A subset of upper-air stations (∼170) comprises the global climate observing system (GCOS) upper-air Network (GUAN, WMO, 2014). The GUAN Monitoring Centre is hosted at the European Centre for Medium-range Weather Forecasts (ECMWF). Analysis of GUAN data is optimised by the U.S. National Oceanic and Atmospheric Administration (NOAA) National Climatic Data Center (NCDC). NOAA/NCDC archives all GUAN data and makes them available through the Integrated Global Radiosonde Archive~~(IGRA~~[6][6] (IGRA). A subset of GUAN has been selected to establish the GCOS reference upper-air network ~~(GRUAN, WMO, 2013)~~(GRUAN; WMO, 2013) , providing radiosonde data from reference-quality stations with traceable uncertainty estimates (Bodeker et al., 2016). Higher vertical resolution radiosonde data, but spatially and temporally more limited, are provided by the Stratospheric-tropospheric Processes And their Role in Climate data center~~(SPARC~~[7] [7] (SPARC) through the U.S. High Vertical Resolution Radiosonde Data~~(HVRRD~~[8][8] (HVRRD).

~~Off-the-shelf commercial **MWR** and DWLare now robust instruments that can be deployed for unattended, continuous operations, providing atmospheric profile observations in nearly all-weather conditions. However, for both technologies~~For both MWR and DWL, networking at national and international level is still in its infancy (Hirsikko et al., 2014; Thobois et al., 2018), meaning MWR and DWL data could be exploited more effectively in the future. The U.S. ARM program[9] runs a network of several MWR (Cadeddu et al., 2013) and also IRS, though still at a limited number of stations. A first attempt at MWR network operation in Europe was the LUAMI (Lindenberg Upper-Air Method Intercomparison) campaign funded by the German ~~Weather~~ Meteorological Service (DWD) to demonstrate the capabilities of MWR profiler systems for use in operational meteorology. A test network of eight MWR profilers supplied quality-checked data in near real-time to a network hub (Güldner, 2013, and references therein). Several European COST[10] actions taking place over the last fifteen years have worked towards the establishment of an international network of MWR~~(~~: MWRnet[11] ~~). MWRnet~~ is a bottom-up network of users, currently grouping more than 100 MWR of different types worldwide (Figure 3b), with 25 profilers located in Europe. MWRnet activities demonstrate the potential of MWR observations for data assimilation (Caumont et al., 2016) and the maturity of these sensors for network deployment (Illingworth et al., 2019). As a consequence, the European national meteorological services network (EUMETNET) accepted the business case for a European MWR network as part of the ~~operational~~ Composite Observing System (EUCOS) service E-PROFILE[12] which will be implemented until 2023 (Rüfenacht et al., 2021).

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[6]~~https://www.ncdc.noaa.gov/data-access/weather-balloon/integrated-global-radiosonde-archive~~
[6]https://www.ncdc.noaa.gov/data-access/weather-balloon/integrated-global-radiosonde-archive
[7]
[7]http://www.sparc-climate.org/
[8]
[8]http://www.sparc-climate.org/data-center/data-access/us-radiosonde/
[9]www.arm.gov/capabilities/instruments/mwrp
[10]Cooperation in Science and Technology; https://www.cost.eu/
[11]http://cetemps.aquila.infn.it/mwrnet/
[12]https://e-profile.eu/

~~The worldwide national~~ National **RWP** networks are operated worldwide (Figure 3c) ~~are operated~~ utilising various frequencies, i.e. in Australia (14 systems), China (128, Liu et al., 2020), Japan (33, JMA[13]), Canada (7), United States (9, NOAA[14]), and in several European countries (32)~~, covering various frequencies~~. EUMETNET E-PROFILE [15] coordinates the RWP network operations in Europe, Canada and Australia. However, many RWP are not yet integrated in such coordinated networks but are rather operated individually by national hydrological and meteorological services (NHMS), airports, private companies, or research institutions (Ruffieux, 2014).

In Europe, many of the **DWL** dedicated to meteorological applications are located at stations that also serve the Aerosol, Cloud and Trace gases Research Infrastructure~~(ACTRIS[15])community~~[15] (ACTRIS). The US ARM program operates a network of several DWL alongside their MWR and cloud radars (Mather and Voyles, 2013). Operational DWL are incorporated in the urban meteorological observation system (UMS-Seoul) designed and installed in Seoul, South Korea (Park et al., 2017), the 3DREAMS network in Hong Kong, China (Yim, 2020), and DWL are a major component of the New York State Mesonet~~(Thobois et al., 2018). There are now significant numbers~~[16] (Thobois et al., 2018; Brotzge et al., 2020; Shrestha et al., 2021). There is now a significant number of DWL deployed in commercial networks for wind energy applications mostly dedicated to observe winds at turbine level (around 50-150 m altitude) rather than the full extent of the ABL, and the data may be commercially sensitive.

ALC are the most widely used instruments in ground-based profile remote sensing networks. There are several network initiatives coordinating ALC measurements globally, such as the NASA-led Micro-Pulse Lidar Network~~(MPLNET[17]), the Photochemical Assessment Monitoring Stations (PAMS[18]) network of the~~ (MPLnet; Welton et al., 2018), the U.S. Environmental Protection Agency (EPA) network for Photochemical Assessment Monitoring Stations[18] (PAMS; Caicedo et al., 2020), or the ~~AD-Net in Asia (Shimizu et al., 2016). Currently, two ACTRIS thematic centers~~ Asian Dust and aerosol lidar observation network (ADnet; Shimizu et al., 2016), amongst others. With more than 370 units (status 2021) transmitting data in near real-time (Figure 3), EUMETNET E-PROFILE combines the majority of ALC networks established across Europe (Figure 3d). E-PROFILE ALC data partly coincide with operations of ACTRIS, which has two topical centres (Center for Aerosol Remote Sensing (CARS); Center for Cloud Remote Sensing (CCRES)) ~~are~~ developing services to enhance the quality of ALC measurements ~~. And~~ and also the European research infrastructure Integrated Carbon Observing System~~(ICOS[19]) is increasingly operating ALCs to monitor the ABL height at their stations. E-PROFILE[20], part of the EUMETNET Composite~~

[13]https://www.jma.go.jp/jma/en/Activities/windpro/windpro.html

[14]https://psl.noaa.gov/data/obs/datadisplay/

[15]

[15]~~https://www.actris.eu/~~

[15]https://www.actris.eu/

[16]https://www2.nysmesonet.org/

[17]https://mplnet.gsfc.nasa.gov/

[18]

[18]https://www.epa.gov/amtic/photochemical-assessment-monitoring-stations-pams#sites

[19]https://www.icos-cp.eu/

[20]~~https://www.eumetnet.eu/activities/observations-programme/current-activities/e-profile/~~

~~Observing System (EUCOS), combines the majority of ALC networks established in Europe. In 2021, more than 370 units were transmitting data in near real-time (Figure 3).~~

~~DIAL and Raman lidars are mostly organised in research networks, such as ACTRIS/EARLINET[20], NDAAC[20], or PollyNet (Baars et al., 2016), providing observations of the full troposphere and even lower stratosphere. However, as these sensors are less autonomous compared to MWR, RWP, DWL, or ALC, spatial coverage tends to be lower for these networks.~~

~~Several European COST actions have helped advancing ground-based profile remote sensing of the ABL and the development of layer height retrieval methods, including Action 710 (Harmonisation of the pre-processing of meteorological data for atmospheric disper~~ ~~, EG-CLIMET (European Ground-based observations of essential variables for CLImate and METeorology; Illingworth et al., 2015)~~ ~~, and TOPROF (Towards Operational ground based PROFiling with ceilometers, Doppler lidars and microwave radiometers for improving~~ ~~. Following the progress made in this field over recent decades, the action PROBE[20] (PROfiling the atmospheric Boundary layer at European~~ ~~now focuses on the harmonisation of operational procedures which is necessary to ensure also higher level products are comparable across Europe and even globally.~~

~~These EU COST actions are paramount for the exchange of knowledge and best practices between networks such as E-PROFILE, ACTRIS and ICOS. The networks not only collect and archive the observations but also aim to harmonise sensor settings and standardise file formats. In close collaboration with NHMS, academia and instrument manufacturers, standard operating procedures are being formulated and implemented, house-keeping data are closely monitored, detailed correction procedures are applied (e.g., Hervo et al., 2016) and advanced data products are developed, including the detection of~~ (ICOS), which is increasingly monitoring ABL ~~heights~~sublayer heights for the support of greenhouse gas assessments.

3 ABL height retrievals

Layer boundaries both within and at the top of the ABL (Sect. 1.1) constitute zones of transition between air of different characteristics. The various physical quantities (Sect. 2.1) derived from profile measurements (Sect. 2.2) each capture some aspects of the ABL development determining these layer heights (Figure ~~??~~2). The most common methods developed to retrieve the ABL sublayer heights from profiles of temperature and humidity (Sect. 3.1), wind and turbulence (Sect. 3.2), or aerosol characteristics (Sect. 3.3) ~~, respectively,~~ are outlined in this section. Certain approaches (such as the bulk-Richardson method here described in Sect. 3.1) are in fact not only associated with only one of those categories as they exploit a combination of atmospheric variables. While some measurement systems capture multiple variables simultaneously (e.g. radiosondes), the synergy between measurements from different ground-based remote sensing profilers (e.g. combining the temperature profile from MWR and wind profile from DWL) is a promising approach as it allows for multi-variable parameters to be calculated.

Limitations and uncertainties are discussed and where possible linked to ~~limitations~~ the characteristics of the sensors used for data collection. Two prominent effects reducing the capability of many active ground-based remote sensing instruments are a) a potential blind zone that reduces the capability of observing shallow layers in the near range and b) insufficient ~~SNR~~

[20]~~https://www.earlinet.org~~

[20]~~https://www.ndsc.ncep.noaa.gov/~~

[20]~~http://www.probe-cost.eu/~~

~~in~~ signal-strength at higher altitudes. Profilers with a certain blind zone (many ~~aerosol and Doppler wind lidars~~ lidars or radar wind profilers) do not provide information in the first range ~~rates~~ gates near the sensor which means, when the signal is send upwards (e.g., DWL vertical stare or high elevation angles), the first reliable measurement level may be located above ~~the $MBLH$ when layers are very shallow . In this~~ a shallow $MBLH$. In such a case, the derived heights ~~only provide~~ should be interpreted as an 'upper limit' of the true ~~layer height~~$MBLH$. Similarly, observations obtained under low SNR conditions (e.g., due to low aerosol load) may not capture the full extent of the ABL (Liu and Liang, 2010) in which case derived layer heights should be considered a lower limit (Bonin et al., 2018; Krishnamurthy et al., 2021).

It is generally challenging to objectively quantify the performance of a method used for layer height detection, mainly because there is no absolute reference for ABL heights against which the derived product could be verified. Instead, evaluation is usually based on inter-comparisons, both between methods using the same quantity and between results obtained from different atmospheric variables. During interpretation it is hence key to consider that discrepancies not only reflect the errors of the respective height retrieval methods and the uncertainties in the atmospheric profiles analysed but may further be affected by a series of methodological aspects~~:~~.

– ~~a potential mismatch in~~ A potential mismatch can be introduced by the representation of the ~~observed atmospheric quantity~~ analysed profile linked to data acquisition or processing (e.g., profile vertical ~~resolution~~and temporal resolution (averaging), horizontal displacement of the sensor)~~,~~.

– ~~potential difference in the signatures of atmospheric processes captured by the respective methods~~ Atmospheric processes portrayed by the observations may differ (e.g., when comparing thermodynamic layer estimates to aerosol-based layer estimates).

– ~~all~~ All layer *heights* in reality relate to a transition *zone* between two atmospheric layers, so that the specific signature in the atmospheric profile associated with the respective layer height is relevant ~~(e.g., is $CBLH$ located at the bottom, middle or top of~~ ~~.~~

– ~~during times of pronounced temporal variations in layer heights (e.g., morning growth of CBL), temporal resolution and averaging can naturally affect the agreement between methods~~(e.g., is $CBLH$ located at the bottom, middle or top of EZ?; Helmis
.

Due to the lack of a better alternative, thermodynamic layer heights (Sect. 3.1) derived from radiosonde profiles (Sect. 2.2.1) are most commonly used as a reference (Seibert et al., 2000). However, comparing balloon ascends and ground-based remote sensing data can be prone to some systematic discrepancies connected to horizontal and temporal variations in ABL dynamics.

– The horizontal drift of the balloon during the ascent means ~~radiosondes may observe~~ vertical profiles derived from radiosondes may be influenced by spatial variations in ~~ABL dynamics that are not captured by the ground-based remote sensing instrument at a fixed location. This may impact~~ ABL dynamics and do not necessarily represent the ABL structure just above the launch site. This impacts the comparison especially where ABL dynamics respond to surface heterogeneities (e.g., Tang et al., 2016; Peng et al., 2017)~~, but~~. But also the synoptic flow plays a role given radiosonde balloons are drawn into regions of convergence so that their profiles are more likely to trace convective activities (Schween et al., 2014).

- Spatial displacement between balloon ascends and the ground-based profile can be further altered if the remote sensing instrument is operating on a moving platform (e.g., ship-based observation; Tucker et al., 2009).

- At the EZ, convective plumes can cause variations of $ABLH$ at the order of several hundred metres (∼150-250 m; Hennemuth and Lammert, 2006; Granados-Muñoz et al., 2012) within a few minutes. While some ground-based profiling sensors operate at very high resolution and can hence capture such temporal variations, the radiosondes only monitor the layer boundary at one given instance.

- The agreement between layer heights detected by methods based on different atmospheric quantities varies with atmospheric conditions (such as stability, cloud dynamics, etc.; Sect. 4.4). As these usually change through the course of a day, the timing of radiosonde ascents relative to the diurnal cycle of the ABL dynamics can affect the comparison statistics. ~~It should be noted that~~

- ~~*standard sounding data*~~*Standard* ~~,~~ sounding data (i.e. radiosonde profiles reduced to significant pressure levels~~(~~; Sect. 2.2.1) ~~,~~ yield higher $ABLH$ than data at high vertical resolution which can introduce structural uncertainties of a few hundred meters in long-term statistics.

- Systematic performance errors of the radiosonde humidity sensors (Sect. 2.2.1) lead to reduced accuracy of humidity-based detection methods (Sect. 3.1) in the presence of clouds.

All these aspects should be considered when interpreting limitations and uncertainties of the various methods. In general, uncertainties in layer height detection vary with time of day and differ between the layer targeted. Uncertainty increases when multiple ABL ~~sub-layers are present as layer boundaries not only have to be detected but a second,~~ sublayers are present given not only the detection of a layer boundary is required but rather a second step, the so-called *layer attribution*~~step ,~~ is required. Particularly at times with significant temporal ~~variation~~ variations in ABL dynamics (e.g., morning growth and evening decay of the CBL, formation of a low-level jet, advection of air masses, ~~morning growth and evening decay of the CBL,~~ formation of clouds or fog), multiple layer boundaries need to be interpreted with care.

Beyrich and Leps (2012) developed a scheme that utilises the agreement between different methods (in their case thermodynamic and wind-based detection applied to radiosonde profiles) to quantify the uncertainty of the layer heights at a given moment and assign quality flags accordingly. This is a promising approach that could be further extended where data from multiple systems are available simultaneously that allow for a range of detection methods based on different atmospheric variables to be applied in synergy.

3.1 Methods based on temperature and humidity

Detection methods for ABL heights based on temperature and/or humidity profiles rely on thermodynamic effects. They allow for the identification of daytime and nighttime layer heights, namely $CBLH$, $SBIH$, $SBLH$, and RLH (Sect. 1.1; Seibert et al., 2000; Seidel et al., 2010, 2012, and references therein). While some methods are directly applied to the profiles of air temperature, others utilise the potential temperature that considers atmospheric stability or the virtual potential temperature which accounts for atmospheric humidity effects in addition (Figure ~~??~~2). Computation of θ (θ_v) requires atmospheric pressure (and humidity) which are at times obtained from external data sources (e.g., other sensors, reanalysis). Some methods directly explore profiles of relative humidity or specific humidity (Beyrich and Leps, 2012). Alternatively to air or potential temperature, the brightness temperature (Sect. 2.1) observed by radiometer profilers (Sect. 2.2.2), can be used as an input for layer height retrievals, as this physical quantity holds information on both temperature and humidity (similarly to the virtual potential temperature).

Temperature and humidity methods can be applied to profile data from in-situ measurements (Sect. 2.2.1), radiometers, DIAL or Raman lidars (Sect. 2.2.2) but are also very commonly implemented in numerical modelling when ABL heights are provided as a diagnostic variable determined from the model output (e.g., Cohen et al., 2015).

3.1.1 Methods

The two most commonly applied temperature-based approaches for the detection of $CBLH$ are the parcel method (Holzworth, 1964) and the ~~bulk-Ridchardson method (Vogelezang and Holtslag, 1996)~~bulk-Richardson method (Hanna, 1969; Vogelezang and Holtslag . The parcel method defines $CBLH$ as the height to which an air parcel with ambient surface air temperature can rise adiabatically from the ground by convection and is obtained by following the dry adiabat from the surface up to its intersection with the temperature profile. ~~The~~ While the parcel method is only applicable ~~in~~ under unstable atmospheric conditions~~. The bulk Richardson~~ , the bulk-Richardson method takes into account the implications of wind shear contribution to turbulence generation and is hence applicable in all stability regimes. The bulk-Richardson number Ri_b represents the ratio of turbulence induced by thermal buoyancy and wind shear, respectively~~. It can essentially be interpreted as an advanced parcel method that considers the shear contribution. Profiles~~ , and profiles of both temperature and horizontal wind ~~velocity~~ are required to calculate Ri_b. ~~The bulk Richardson method can be applied to detect the~~ It is essentially a synergy approach that combines thermodynamic and dynamic effects and could as well be grouped into dynamic retrieval methods (Sect. 3.2). Both $CBLH$ ~~and indicates when the atmospheric stratification is stable but does not provide a layer height estimate in the latter case. These layer heights are detected as the heights~~ and $SBLH$ can be determined as the altitude where Ri_b exceeds a critical threshold. Typical ~~thresholds for the bulk Richardson method are~~ values of this thresholds are around 0.10-0.40 (Sørensen et al., 1996) ~~and~~ or 0.25-0.50 (Seibert et al., 2000) with the value 0.25 used to estimate ~~$ABLH$ provided in~~ layer heights provided in the ERA-Interim re-analysis data (von Engeln and Teixeira, 2013). The ~~choice of the~~ exact threshold value has a relatively modest impact on the layer detection accuracy ~~(Seidel et al., 2012; Guo et al., 2016; Beyrich and Leps, 2012; Cimini et al., 2013)~~ if remaining < 0.5 (Seidel et al., 2012; Guo et al., 2016; Beyrich and Leps, 2012; Cimini et al., 2013). As the ~~bulk Richardson~~ bulk-Richardson method and the parcel method are identical if the threshold value is set to 0, ~~the bulk Richardson method, by definition, gives a higher layer result than the parcel method~~layer estimates from the former are greater by definition. This increment was found to be about 20 m on average for the $CBLH$ (Collaud Coen et al., 2014). ~~Given~~ As moisture lightens the air and allows it to ~~convectively rise~~ rise convectively to greater altitudes, ~~the parcel and the bulk-Richardson methods based on~~

using θ_v instead of θ (Sect. 2.1) ~~result~~ in both methods results in slightly greater layer heights ~~(3–8% Collaud Coen et al., 2014)~~ (by 3–8%; Collaud Coen et al., 2014).

Both the parcel and ~~bulk Richardson~~ bulk-Richardson method highly depend on the accuracy of the ambient air temperature at the surface. A temperature excess corresponding to the strength of convective thermals can be added to θ_v at the surface under unstable conditions (Holtslag and Nieuwstadt, 1986; Seibert et al., 2000). This excess temperature is usually applied when the surface air temperature is measured at a height exceeding the standard 2 m, as in e.g., radiosoundings or NWP model data (Stohl et al., 2005).

In addition to the commonly used ~~Parcel and bulk Richardson~~ parcel and bulk-Richardson methods, several ~~others~~ other thermodynamic methods are available to detect ~~ABL heights~~ heights of ABL sublayers, including:

- $SBIH$ and $SBLH$ under stable conditions
  - As a clear indicator of a stable boundary layer, ~~the height of the SBI~~ $SBIH$ is diagnosed from air temperature profiles (Bradley et al., 1993; Seidel et al., 2010).
  - At the transition between the SBL and the neutral residual layer, $SBLH$ is marked by a vertical gradient of θ equal to zero, that corresponds to the theoretical lapse rate (Collaud Coen et al., 2014) or equal to a critical lapse rate determined by the ~~maximal~~ maximum variance of the gradient (Min et al., 2020).
  - Liu and Liang (2010) refine $SBLH$ detection by choosing the first height above ground that either shows a minimum in the potential temperature gradient or local maximum in horizontal wind speed if a LLJ is present. The method uses surface classification (land, ocean, ice) to determine critical thresholds.

  It should be noted that the accuracy of $SBLH$ (and to some extend $SBIH$) detection highly depends on the vertical resolution of the analysed temperature profile. In contrast to the parcel method or the Ri_b method, no vertical interpolation or smoothing of the profile data ~~should~~ can be performed.

- $CBLH$ under unstable conditions
  - The *Heffter method* determines $CBLH$ as the minimum height where the vertical gradient of θ exceeds 0.005 K m^{-1} while θ changes by more than 2 K across the inversion layer (Heffter, 1980).
  - The minimum height where θ reaches a certain increment compared to its ABL minimum can mark the $CBLH$ (Nielsen-Gammon et al., 2008).
  - The maximum negative vertical gradient of refractivity (Sect. 2.1) or humidity was found to mark the $CBLH$ (Seidel et al., 2010).
  - Schmid and Niyogi (2012) improve the detection of $CBLH$ by allocating heights where a change in the vertical θ_v gradient coincides with a dew point temperature inversion.

- $MBLH$ independent of atmospheric stability

- ~~The vanishing gradient in the air temperature profile marks~~ To derive the $MBLH$ from temperature profiles, both the base of an elevated temperature inversion ~~that serves as a cap to mixing below and can be considered the~~ (Seidel et al., 2010) and the height of the maximum positive θ gradient (Stull, 1988; Seidel et al., 2010) have been applied. As vertical mixing can already be reduced for a certain region below a positive vertical gradient in air temperature, the latter criterion can be a more accurate indicator for $MBLH$ ~~(Seidel et al., 2010)~~. However, the profile data of atmospheric pressure required to calculate θ may not always be available.

- Cimini et al. (2013) apply a multivariate statistical regression method trained with real observations to derive $MBLH$ directly from T_b. This method exploits all the information in the MWR observations and is independent of uncorrelated retrieval errors in the temperature and humidity profiles (Sect. 2.2.2) as both systematic and random errors are inherently accounted for.

- To ensure continuous layer detection, different temperature-based methods can be combined depending on atmospheric stability as the most applicable method may vary during the course of the day and between land cover types. Maybe the most common method synergy is the combination of the parcel method ($CBLH$) and the $SBIH$ at night.

- Some disagreement between temperature- and humidity-based methods stems from the presence of clouds, which create a complex vertical ABL structure (Sect. 4.6). While humidity-based methods tend to respond to the layer boundary at the cloud top (large negative humidity gradient), the maximum gradient of θ usually occurs in the middle of the EZ above the cloud. To overcome this issue, $ABLH$ can be assigned to the level where all of the above variables exhibit pronounced variations simultaneously, rather than looking for the strongest change in one specific profile variable only.

3.1.2 Capabilities and limitations

Long-term, multi-site comparisons reveal some systematic differences between the various temperature- and humidity-based methods (e.g., Seidel et al., 2010; Beyrich and Leps, 2012). The Heffter method often overestimates $MBLH$ and the definition of thresholds was found challenging (see discussion in Caicedo et al., 2020, and references therein). Sinclair et al. (2021) find agreement and sign of systematic biases depend on atmospheric stability. The T_b regression method (Cimini et al., 2013) provides height estimates that are mostly consistent with the bulk-Richardson approach. Methods based on finding extreme vertical gradients are in better agreement with each other than those based on locating elevated temperature inversions (Seidel et al., 2010).

It is generally concluded that uncertainties in layer detection are closely linked to uncertainties in the atmospheric profiles analysed (e.g., errors in surface wind speed, vertical interpolation and vertical resolution; Seidel et al., 2012). Given such errors are much more pronounced (10-80 %) for low layer heights (<1-2 km), relative uncertainties of the layer detection can be large (>50 %) for shallow layers but usually remain below 20 % for layer heights >1 km (Seidel et al., 2012; Aryee et al., 2020; Guo et al., 2016). Uncertainties are usually greatest during the evening decay (Sect. 4.4) of the CBL as this is a time of significant change. Methods agree better when applied to radiosonde profiles at midday compared to midnight conditions (Beyrich and Leps, 2012)~~. This~~, which further highlights that the $CBLH$ layer boundary is ~~better defined~~ often well-defined while detection of $SBLH$ is more ambiguous.

~~As the parcel method is more likely to capture shallow layer heights, Seidel et al. (2010) conclude that diurnal and seasonal variations based on this method generally tend to have a greater amplitude and can be considered more consistent than those derived from other approaches. This is in agreement with the MWR analysis of Collaud Coen et al. (2014) who found~~ Both the parcel method ~~to be more robust (compared to the bulk Richardson method or the analysis of surface-based temperature inversions or humidity gradients) and hence better suited for automatic real-time detection of the $MBLH$, as it provides good results under a wide range of meteorological conditions. The parcel method is~~ and the bulk-Richardson method are sensitive to surface-level data (Sect. 3.1.1). For example, a change in surface temperature by ±0.5 K leads to uncertainties at the order of ±50-150 m for the maximum $CBLH$ in the early afternoon ~~(Collaud Coen et al., 2014)~~ at a mid-latitude continental site (Collaud Coen et al., 2014). Careful quality control of the measurements is hence required to ensure physically reasonable coupling of the surface air temperature value to the first values of the temperature profile (Beyrich and Leps, 2012). Horizontal and/or vertical separation between the site of the surface measurements and the radiosonde launch site can cause artificially large vertical gradients in the combined temperature profile which may result in significant average differences in the derived $ABLH$ of up to several hundred metres (Seidel et al., 2010). In such ~~cases~~a case, it is preferable to initialise the ~~parcel method~~ layer detection with the first reported upper air level instead of surface observations.

From an analysis using MWR data, Collaud Coen et al. (2014) found the parcel method to be more robust and hence better suited for automatic real-time detection of the $CBLH$ compared to the bulk-Richardson method because the latter requires more input data. In addition to the temperature (and pressure) data, wind profile observations are needed which introduces additional measurement uncertainties and missing values from a second system. Such issues are slightly reduced when the methods are applied to e.g. radiosonde data, as here both wind and temperature are gathered by the same measurement system. Due to the simplicity of the parcel method it is more likely to capture shallow layer heights. Seidel et al. (2010) conclude that diurnal and seasonal variations based on this method generally tend to have a greater amplitude and can be considered more consistent than those derived from other approaches. This is in agreement with the analysis by Collaud Coen et al. (2014).

The presence of clouds increases uncertainty in $CBLH$ retrievals for all methods (Sect. 4.6), so that temperature-based methods applied to radiosondes and MWR profile data show better agreement during clear-sky days (Cimini et al., 2013). When the parcel method is applied to temperature profile data obtained from radiosondes and MWR, the latter tend to significantly under-estimate the $MBLH$ (Cimini et al., 2013). Up to now, no quantitative, comparison analysis has been performed regarding $MBLH$ estimates from different MWR types, although there have been field campaigns where multiple commercial MWR were operated side-by-side, such as the Joint CALibration experiment ~~(JCAL, Pospichal et al., 2016)~~ (JCAL; Pospichal et al., 2016) or the recent Field Experiment on submesoscale spatio-temporal variability in Lindenberg~~(FESSTVaL~~[20][20] (FESSTVaL).

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[20]https://fesstval.de/

The method using T_b regression analysis (Cimini et al., 2013) relies on "independent training data". Given there is no absolute reference when it comes to ABL heights, the choice of training data and potential systematic differences in the physical representation of ABL dynamics by the observed quantity analysed ~~(Sect. 1.1)~~ and the respective detection approach ~~(Sect. 4) may~~ can affect the performance of this method. Still, even when trained with aerosol-derived layer estimates (Sect. 3.3), the T_b regression method shows better agreement with layer estimates from the bulk-Richardson method applied to radiosonde profiles compared to the parcel method and θ_v-gradient method applied directly to ~~the MWR~~ temperature and humidity data from the MWR.

3.2 Methods based on wind or turbulence

Methods exploiting wind profile observations to detect ABL heights can generally be grouped into those using components of the mean wind and those based on turbulence indicators. The objective of these methods is to identify the height of the turbulent layer connected to the surface. The mixing is either caused by buoyancy-driven turbulence ~~(in case of the CBL),~~ or shear-driven turbulence~~(in case of the SBL)~~, or a combination of the two. ~~Intermittant~~ Intermittent turbulence in the residual layer can affect the performance of layer detection algorithms (Pichugina and Banta, 2010) but wind and turbulence methods are usually not applied to detect RLH.

3.2.1 Methods

Using mean wind profiles (Figure ~~??~~2), the most commonly applied layer detection approach is the bulk-Richardson method (see Sect. 3.1). Looking at the relation between thermally-induced buoyancy and shear-induced turbulence, this synergy method is applicable under all stability conditions (i.e. for detection of both $SBLH$ and $CBLH$). Alternatively, $SBLH$ can be identified as the height of a local maximum in horizontal wind speed or a local minimum in vertical wind speed or wind shear, respectively (Balsley et al., 2006; Banta et al., 2006; Pichugina and Banta, 2010; Lemone et al., 2014). Johansson and Bergström (2005) find significant changes in the mean ascent rate of radiosonde balloons indicate the transition from turbulent to non-turbulent regimes, whereby exploiting a mean quantity to diagnose turbulence indicators indirectly (Lemone et al., 2014).

Due to advances in high-resolution ground-based profiling, direct measures of atmospheric turbulence can be determined quantitatively with increasing accuracy (Sect. 2.2.3).

Turbulence can be diagnosed from the refractive index structure parameter (Sect. 2.1) observed by sodar and RWP (Sect. 2.2.3). The peak in the vertical profile of the refractive index structure parameter caused by small-scale buoyancy fluctuations across the entrainment zone has been found to coincide with the $MBLH$ (White, 1993; Angevine et al., 1994; Wilczak et al., 1997). Given these fluctuations are associated with relatively high SNR in sodar and RWP observations, some methods assign $CBLH$ to a local peak in RWP SNR (Liu et al., 2019a; Collaud Coen et al., 2014). RLH can also be detected by analyzing profiles of C_n^2, but only for specific (mostly cloud-free) weather conditions. $ABLH$ is diagnosed from space-borne GNSS-RO observations as the strongest negative gradient in refractivity (Ao et al., 2012; Chan and Wood, 2013), that is associated with the strong moisture and temperature gradients usually present at the top of the ABL (Xie et al., 2012).

Estimates of atmospheric turbulence can also be obtained from temporal and/or spatial fluctuations in high-resolution wind profiling data. The most commonly exploited turbulence variables derived from high-frequency wind components are the variance of vertical velocity, variance in horizontal velocity, turbulent kinetic energy and the eddy dissipation rate (Sect. 2.1). To ensure layer detection relies on the measurement of turbulence intensity, it is important to remove non-turbulent fluctuations from the wind field components (Bonin et al., 2018). Applying a high-pass filter was found to be a simple but effective means to sufficiently reduce the influence of sub-mesoscale motions, drainage flows, and ~~low atmospheric~~ gravity waves (Bonin et al., 2017, 2018), with frequencies on the order of minutes to tens of minutes (Finnigan et al., 1984). Berg et al. (2017) chose to detect layer heights based on the normalised vertical velocity variance to reduce the impact of coherent vertical motions above the ABL.

During convective atmospheric conditions, the vertical velocity variance from vertically pointing profile observations is the most direct measure of the instantaneous mixing within the CBL. The $CBLH$ is commonly assigned to the height above ground where the vertical velocity variance (Figure 2) falls below a set threshold~~(Figure ??)~~, with both absolute ~~(0.04-0.16 m^2 s^{-2}; Tucker et al., 2009; de Arruda Moreira et al., 2018; Barlow et al., 2011; Huang et al., 2017; Vakkari et al., 2015)~~ (0.04-0.16 m^2 s^{-2}; Tucker et al., 2009; de Arruda Moreira et al., 2018; Barlow et al., 2011; Huang et al., 2017; Vakkari et al., 2015; Theeu and relative values ~~(e.g., 10 % of profile maximum, Barlow et al., 2011)~~ (e.g., 10 % of profile maximum; Barlow et al., 2011) implemented successfully. Given the gradual decay of turbulence in the afternoon and evening CBL, Schween et al. (2014) find $CBLH$ detection to be particularly sensitive to the threshold value during ~~these periods~~this period. The choice of threshold value can depend on the ABL structure (Tucker et al., 2009; Huang et al., 2017) and the ~~scanning strategy-dependent~~ scanning-strategy dependent noise levels (Bonin et al., 2018). When shear-driven turbulence dominates, horizontal velocity variance becomes a better indicator for the layer boundaries. The vertical profile of horizontal velocity variance depends on atmospheric stability, with a near-surface peak under slightly stable conditions, a rather constant vertical distribution under medium stable conditions, and a maximum aloft near the core of the LLJ under strongly stable conditions (Banta et al., 2006). Tucker et al. (2009) use the same threshold values as for the vertical velocity variance to determine shallow layer heights from horizontal velocity variance.

Turbulent kinetic energy or eddy dissipation rate are other quantitative measures of turbulence useful for layer detection. Lemone et al. (2014) find a relative value of 5% of the profile maximum TKE most suitable for the detection of $SBLH$ while LeMone et al. (2013) apply fixed values (0.101 and 0.200 m^2 s^{-2}) to determine $CBLH$ from model data. Vakkari et al. (2015) assign $CBLH$ to the height where the eddy dissipation rate falls below 10^{-4} m^2 s^{-3} while Frehlich et al. (2006) examine the strongest negative gradient of this quantity. For specific scan patterns, the variance of radial velocity (i.e. the native variable obtained form Doppler wind lidar measurements) is directly related to TKE (Sect. 2.2.3) and is hence also exploited for layer detection. Pichugina and Banta (2010) determine the $SBLH$ as the height of the first significant local minimum in the vertical profile of the radial velocity variance.

Under well-mixed, convective conditions, i.e. when $CBLH$ coincides with $ABLH$, turbulence-based detection methods can be supported by applying SNR requirements (Moreira et al., 2015; Pearson et al., 2010; Singh et al., 2016). Lothon et al. (2006) evaluate the total velocity variance from DWL observations for layer detection, which inherently includes the SNR information as the recorded signal responds to both atmospheric variations and instrument-related noise (Tucker et al., 2009). ~~Some methods assign $CBLH$ to a local peak in RWP SNR (Liu et al., 2019a; Collaud Coen et al., 2014).~~

To cover the full range of ~~$ABLH$~~ $MBLH$ at a given measurement location, wind measurements from multiple data sources can be combined ~~(e.g., Doppler sodar and RWP; Beyrich, 1997; Angevine et al., 2003)~~(e.g., sodar and RWP; Beyrich and Görsdorf, 1995; . The great advantage of scanning DWL systems is that a series of wind and turbulence variables can be obtained within a rather short time interval by a single sensor (Sect. 2.2.3). For example, vertical stare measurements can be alternated with range-height indicator (RHI) scans (Tucker et al., 2009) to capture convection or plan position indicator (PPI) scans at low elevation angles (Vakkari et al., 2015) to capture shallow layers. To facilitate the composition of layer information from various atmospheric variables (mean wind fields, different turbulence indicators, SNR), fuzzy logic algorithms (Bianco and Wilczak, 2002; Bianco et al., 2008; Allabakash et al., 2017) or ~~random forest~~ machine learning (Krishnamurthy et al., 2021) are increasingly implemented. Recent advanced approaches (Bonin et al., 2018; Krishnamurthy et al., 2021) combine a diverse combination of atmospheric variables which enables reliable layer detection under nearly all atmospheric conditions. To enhance agreement with aerosol-derived layer heights (Sect. 3.3), Bonin et al. (2018) give less weight to the vertical velocity variance during layer height retrieval whereby moving the focus towards those indicators that portray the resulting 'mixed conditions' instead of the mixing process itself. Where the ABL responds to significant surface heterogeneities (Banks et al., 2015; Haid et al., 2020; Vakkari et al., 2015), site-specific design of DWL scanning strategies is recommended to best capture the ~~local ABL spatial and temporal variability~~ variability in $MBLH$.

3.2.2 Capabilities and limitations

The altitude range of the atmospheric profile captured by the measurements and the accuracy of wind and turbulence data from ground-based remote sensing systems depends on instrument capabilities and measurement setup (Sect. 2.2.3). The ability to detect shallow layers generally depends on how large the blind zone of the sensor is, while observing the full depth of deeper convective conditions is dependent on SNR. Turbulence-based $MBLH$ estimation is particularly applicable in daytime convective conditions (Bianco et al., 2008; Collaud Coen et al., 2014). Decaying turbulence in the residual layer can be a source of added uncertainty (Lemone et al., 2014).

Rainfall can be a significant source of uncertainty for automatic layer detection from profiles of wind and turbulence. It is possible to diagnose from the vertical velocity profile based on the terminal fall speed of rain drops ~~(e.g., using column averaged vertical velocit~~ (e.g., using column averaged vertical velocity $< -1\ \mathrm{m\,s^{-1}}$; Bonin et al., 2018), but such filters will not detect precipitation conditions with lower fall speeds (such as drizzle or snow) and can miss precipitation that evaporates before reaching the surface (virga).

3.3 Methods based on attenuated backscatter

~~Given the distribution of aerosol particles and moisture is in parts a result of mixing processes, attenuated backscatter profiles~~ The distribution of aerosols and moisture usually results from a complex combination of processes, including emission, formation, accumulation, deposition, transport (advection), and also mixing. Profiles of attenuated backscatter (Sect. 2.1)

~~represent~~ hence trace some aspects of the recent history of ABL dynamics. Layer boundaries can be detected if aerosol properties differ between the atmospheric layers examined. The most pronounced layer edge is usually the $ABLH$ because aerosol concentrations and humidity tend to be significantly higher in the ABL than in the FT (Figure 2). But also within the ABL, mixing dynamics can lead to contrasting aerosol properties between different layers~~(Figure ??)~~. During night and early morning, the lowest layer is considered the $MBLH$ ($SBLH$ under stable conditions) while the layer above defines RLH (Sect. 1.1). During unstable daytime conditions, the aerosol-based $MBLH$ forms in response to recent CBL mixing processes. Vertical and temporal changes recorded at high resolution allow for EZ characteristics to be examined. Decoupled, elevated aerosol layers above the ABL can be identified if they possess distinct aerosol characteristics. Lidars that capture ~~depolarization information (Sect. 2.2.4~~additional information (such as e.g., depolarisation, colour ration; Sect. 2.1) can provide ~~additional~~ valuable insights that allow for boundary layer aerosols to be better distinguished from lofted layers (e.g., Bravo-Aranda et al., 2017).

The physical quantity of attenuated backscatter is most commonly observed by ALC or aerosol research lidars (Sect. 2.2.4) but can also be derived from DWL (Sect. 2.2.3). Layer heights derived from different systems can be combined to overcome instrument-related limitations, e.g. results from more powerful lidars can be combined with those from ALC observations (e.g., Wang et al., 2020). As the majority of layer detection algorithms does not rely on absolute values in attenuated backscatter but rather assess relative variations of this quantity in time and height, the range-corrected signal is often used as an alternative input. ~~SNR (or CNR) highly depends on the atmospheric aerosol load so that aerosol-based methods have also been applied to these noise profile data directly (Compton et al., 2013)~~Also profiles of SNR (from e.g. ALC, DWL or RWP; Compton et al., 2013; Molod e can be used as input under the assumption that this variable is directly related to the distribution of scatterers in the atmosphere. While the discussion here focuses on ground-based profilers, it should be noted that aerosol-based techniques can also be used for the analysis of airborne lidar profiles (e.g., Scarino et al., 2014), satellite data ~~(CALIOP, Zhang et al., 2016)~~(CALIOP; Zhang et al., 2016 , or output from numerical simulations.

3.3.1 Methods

Aerosol-based retrievals of ABL heights detect layer boundaries based on regions of significant vertical (and at times temporal) change in attenuated backscatter. Where multiple layers are present within the ABL, the role of the respective layers needs to be examined carefully. Hence, two steps are required to determine the ABL heights from aerosol backscatter observations: (1) *detection of layer boundaries* within and at the top of the ABL, and (2) *layer attribution* to distinguish between simultaneous layers (e.g., $MBLH$ and RLH). Methods that predominantly address the task of layer detection are here considered *first-generation* aerosol-based retrievals, while those with a special focus on the more challenging aspect of layer attribution are grouped into *second-generation* aerosol-based retrievals.

To detect heights (or regions) of potential layer boundaries from attenuated backscatter profiles, a range of indicators has proven useful ~~(see reviews by Emeis et al., 2008; Dang et al., 2019)~~(see reviews by e.g., Emeis et al., 2008; Dang et al., 2019) . These include negative vertical gradients and inflection points (e.g., Sicard et al., 2006; Emeis et al., 2008; Münkel, 2007; Schäfer et al., 2004; Lee et al., 2019), 2D-edge detection (e.g., Canny, 1986; Parikh and Parikh, 2002; Haeffelin et al., 2012), wavelet covariance transform (WCT; e.g., Cohn and Angevine, 2000; Morille et al., 2007; Baars et al., 2008; de Haij et al., 2006; Gan et al., 2010; Granados-Muñoz et al., 2012; Lewis et al., 2013; Caicedo et al., 2020), the cubic root gradient which takes into account the influence of gravity waves (Yang et al., 2017), and spatio-temporal variance (e.g., Menut et al., 1999; Martucci et al., 2007; Lammert and Bösenberg, 2006; Piironen and Eloranta, 1995; Hooper and Eloranta, 1986). For example, $ABLH$ is derived from CALIPSO satellite observations as the maximum in vertical and horizontal variance (Jordan et al., 2010).

Many layer detection methods show varying reliability at different stages of the diurnal ABL evolution. $ABLH$ is usually marked by the strongest negative gradient in attenuated backscatter~~and the~~. The mixing between moist ABL air and dry FT air across the EZ results in an area of strong spatio-temporal variance (e.g., Menut et al., 1999). But also entrainment of RL air into the CBL during morning growth can cause distinct variance signatures (e.g., Lammert and Bösenberg, 2006). While some methods (such as the wavelet approach) are less affected by noise, the simple method of vertical gradient detection can be advantageous in capturing layers at low ranges (Di Giuseppe et al., 2012).

Several approaches have been developed to accomplish the second task of layer attribution. First-generation retrieval methods apply attribution criteria either on the respective indicator (e.g., strongest negative vertical gradient) and/or simply based on height (e.g., first significant negative gradient above ground) to assign the layer of interest. Second-generation retrieval algorithms again can broadly be grouped into the following categories:

- Methods based on general layer characteristics: These methods group ~~observations along the observed~~ aerosol data along the vertical profile into categories. To differentiate between the ABL with high aerosol load and the FT with low aerosol signal, an idealised profile can be ~~fit~~ fitted to the observations (e.g., Steyn et al., 1999; Eresmaa et al., 2006, 2012; Li et al., 2017; Peng et al., 2017). Further do recent artificial intelligence (AI) approaches analyse the profile across the whole layer, incl. extended Kalman filters (Lange et al., 2013; de Arruda Moreira et al., 2018; Kokkalis et al., 2020), *K-means* cluster analysis applied to either the attenuated backscatter profile (KABL; Toledo et al., 2014; Rieutord et al., 2021; Min et al., 2020) or layer candidates derived from first-generation methods (ISABL; Min et al., 2020), or a supervised *AdaBoost* algorithm (ADABL; Rieutord et al., 2021).

- Combination of identification techniques: Given that the various layer-detection techniques can be sensitive to slightly different layer boundaries, a combination of indicators can help to distinguish between layers. For example, Martucci et al. (2010a) combine the height of maximum negative gradient and height of maximum variance to detect $MBLH$ both during day and night together with lofted, decoupled aerosol layers. STRAT-2D (Morille et al., 2007; Haeffelin et al., 2012) uses the variance field to determine which wavelet-detected layer boundary is likely associated with $MBLH$ by analysing the location of the EZ. pathfinderTURB (Poltera et al., 2017) and STRATfinder (using advantages of STRAT+ and pathfinderTURB; Kotthaus et al., 2020) combine gradient and variance field diagnostics before tracing $MBLH$. COBOLT (Geiß et al., 2017) uses a combination of gradients, variance statistics and WCT, varying with solar angle to identify the $MBLH$. Applying Gradient Boosted Regression Trees, de Arruda Moreira et al. (2022) combine an $MBLH$ first-estimate derived using the gradient method with several meteorological variables to retrieve $MBLH$

values that are comparable to those derived from a microwave radiometer, discriminating between $CBLH$ and $SBLH$. The profile-fit approach is combined with WCT by Sawyer and Li (2013) and with the negative gradient detection in the proprietary Vaisala BLview software (Münkel, 2016), respectively. It should be noted that the BLview algorithm provides several layer candidates so that some post-processing is required to select the appropriate $MBLH$. This has been achieved using, e.g., the provided quality flags (Geiß et al., 2017), gradient thresholds (Haman et al., 2012), manual screening (Caicedo et al., 2017), or a combination of time-tracking and height criteria ~~(Mues et al., 2017). Applying Gradient Boosted Regression Trees, (de Arruda Moreira et al., 2022) use a first estimation of the~~ $\cancel{ABLH}$ ~~derived from the gradient method to a ceilometer signal and several meteorological variables to retrieve ABLH values comparable to those derived from a microwave radiometer, discriminating between~~ $\cancel{CBLH}$ ~~and~~ $\cancel{SBLH}$(Mues et al., 2017; Lotteraner and Piringer,
.

- A priori assumptions based on ancillary observations: Where climatology statistics are available from independent measurements (e.g., radiosondes) limits can be prescribed that may vary by season. For example, in pathfinderTURB and STRATfinder absolute limits for $MBLH$ and morning transition growth rate are specified by the user. Some studies set time-specific coefficients (morning, afternoon, and night, respectively) for the WCT ~~Gan et al. (2011); Caicedo et al. (2020)~~ (Gan et al., 2011; Caicedo et al., 2020). STRAT+ (based on STRAT-2D; Pal et al., 2013) ~~,~~ uses radiosonde profiles and turbulent surface sensible heat flux measurements to derive stability information ~~to~~ that aid interpretation of the variance field.

- A priori assumptions from model results: Simple models describing general ABL dynamics or output from NWP models representing varying synoptic conditions have been used to guide layer attribution. For example, Di Giuseppe et al. (2012) use a bulk model (Tennekes, 1973) based on surface sensible heat flux data to define times of morning and evening transition.

- Temporal layer tracking: Temporal consistency is a powerful criterion for layer attribution (Angelini et al., 2009) as it reduces physically unreasonable height fluctuations and growth rates. For example, the temporal-height-tracking method ~~(THT, Martucci et al., 2010a)~~ (THT; Martucci et al., 2010a) uses the $MBLH$ estimate at a previous time step to define the search window for the subsequent detection~~; a~~. A similar approach is implemented in COBOLT (Geiß, 2016) and by Wang et al. (2012) or Caicedo et al. (2020). The MIPA algorithm (Vivone et al., 2021) uses morphological and object-based image processing techniques to improve temporal consistency of the detected $MBLH$. A recent family of algorithms (pathfinder, pathfinderTURB, STRATfinder) apply a graph theory approach to trace the path of $MBLH$ through the day. In CABAM (Kotthaus and Grimmond, 2018a), points of significant negative gradients are connected to layers which are then traced through the day following growth and decay criteria in a dynamic decision tree. The extended Kalman filter applied by Kokkalis et al. (2020) uses information from past profile analysis to inform $ABLH$ detection, which generally improves temporal consistency but can also lead to errors when air with different aerosol load is advected~~(Kokkalis et al., 2020)~~. Also the supervised ADABL algorithm (Rieutord et al., 2021) considers temporal consistency.

– Additional lidar profiles: ~~Some research lidars and novel ALC (Sect. 2.2.4) provide additional profile information other than the attenuated backscatter that can be exploited to differentiate layer characteristics.~~ The POLARIS algorithm analyses the depolarisation ratio in connection with the WCT approach (Bravo-Aranda et al., 2017). The MDS method (Liu et al., 2018) determines $ABLH$ by adding information on the particle size by calculating the degree of difference in aerosol characteristics between observations from two adjacent lidar range gates as a combination of aerosol backscatter and the color ratio (Sect. 2.1). As research lidars and some novel ALC (Sect. 2.2.4) increasingly provide additional profile information on top of the attenuated backscatter, promising developments are expected from algorithms that exploit more than one aerosol variable to differentiate layer characteristics.

While most of these retrieval algorithms output one layer height (usually either $MBLH$ or $ABLH$), simultaneous identification of several layers is possible provided the layer attribution step does account for it (Kotthaus et al., 2020; Milroy et al., 2012; Caicedo et al., 2020; Toledo et al., 2017). Elevated aerosol layers can be traced in addition to ABL heights. For example, Poltera et al. (2017) detect a continuous aerosol layer above the ABL using pathfinerTURB while Pandolfi et al. (2013) track a decoupled aerosol layer above the ABL at a coastal site and Gan et al. (2010) derive RLH and elevated aerosol layers using the WCT approach.

In the absence of clouds, aerosol-derived $ABLH$ is usually located somewhere in the centre of the EZ, where vertical negative gradients and spatio-temporal variance are strongest due to the exchange of aerosols and moisture between the ABL and FT (Menut et al., 1999). Where observations at very high temporal resolution in the order of minutes are available and SNR is sufficient, temporal variations in $ABLH$ permit the estimation of ~~EZ thickness (Cohn and Angevine, 2000)~~ EZD (e.g., Cohn and Angevine, 2000) and entrainment velocities (Träumner et al., 2011). Martucci et al. (2010a) performed spectral analysis on high-resolution $ABLH$ observations to characterize entrainment processes under different atmospheric stability conditions. Alternatively, the transition zone concept, based on the difference between high attenuated backscatter values in the ABL and low values in the FT (Steyn et al., 1999) can be used to determine the ~~EZ thickness~~EZD. Statistical concepts capturing temporal, spatial and small-scale turbulence variations can differ significantly from the transition zone estimates given the latter is mostly limited to small-scale turbulence effects (Träumner et al., 2011).

3.3.2 Capabilities and limitations

Both the detection and the attribution step of aerosol-based layer retrievals highly depend on the quality of the attenuated backscatter profiles analyzed (de Haij et al., 2006; Milroy et al., 2012). $MBLH$ and $ABLH$ retrievals based on attenuated backscatter have seen significant improvements due to recent advances in ALC measurement technology (Illingworth et al., 2019; Cimini et and detailed correction procedures (Hervo et al., 2016; Kotthaus et al., 2016) as both improve data quality and availability. An important variable describing the information content of a measurement at a certain time and range is the SNR (Sect. 2.1). Any combination of high instrument-related noise, low aerosol (and moisture) load, or very deep convection reduces the SNR which can lead to both under- and over-estimation of peak $ABLH$ (Kotthaus et al., 2020). Applying an SNR filter can improve layer detection (Poltera et al., 2017; Min et al., 2020), however, care must be taken in pristine environments (Boy et al., 2019), where atmospheric scatterers are scarce and the recorded signal may not necessarily exceed instrument and background noise significantly. While SNR-limitations mostly lead to uncertainties in the detection of layer boundaries at elevated heights above ground, the detection of shallow layers (nocturnal $MBLH$ and $SBLH$) can be affected by the incomplete optical overlap and near-range artefacts (Schween et al., 2014; Kotthaus et al., 2020; Caicedo et al., 2020). ~~$MBLH$ and $ABLH$ retrievals based on attenuated backscatter have seen significant improvements due to recent advances in ALC measurement technology (Illingworth et al., 2019; Cimini et al., 2020) and detailed correction procedures (Hervo et al., 2016; Kotthaus et al., 2016) as both improve data quality and availability.~~

In addition to instrument-related uncertainties, discrepancies in layer results arise from the choice of retrieval algorithm. Haeffelin et al. (2012) compared five $MBLH$ detection techniques applied to observations from three different ALC at two contrasting measurement sites. While layer detection methods (first derivative, WCT, and two-dimensional derivative) often agree, the greatest uncertainty in final products was associated with the step of layer attribution, even when considering simple categories only. Second-generation algorithms (Sect. 3.3.1) hence put a special focus on the interpretation of the ABL sublayers. Comparing different second-generation methods, it appears that those including criteria on temporal consistency of layer estimates tend to perform slightly better (de Bruine et al., 2017; Knepp et al., 2017).

Agreement between retrieval methods varies with the complexity of the ABL structure. Provided sufficient SNR, results from different methods and sensors tend to agree best in the afternoon during peak convective activity (Milroy et al., 2012) when the CBL extends over the whole ABL leaving essentially no sublayers to confuse the algorithms (Toledo et al., 2017). Layer attribution is challenged when several aerosol layers are present simultaneously, such as during night and early morning. If MBL and RL aerosols have similar characteristics, the $MBLH$ may not be characterised by a particularly strong gradient (Granados-Muñoz et al., 2012). Highest uncertainty generally occurs during the evening transition (Geiß et al., 2017) when new aerosol gradients start to form gradually and decaying turbulence may not be traced successfully by backscatter variance methods in case of homogeneous aerosol distributions (Poltera et al., 2017). In addition to ABL-internal sublayers, elevated aerosol layers ~~advected over the ABL add complexity~~ add complexity and hence layer retrieval uncertainty. Any aerosol-based method is challenged when temporal variations of gradients (or variances) are dominated by advection of aerosols (e.g., due to strong sea breezes or dust transport; Tang et al., 2016; Bravo-Aranda et al., 2017; de Arruda Moreira et al., 2020; Diémoz et al., 2019b; Caicedo et al., 2019). Advances in measurement technology (such as depolarisation information becoming increasingly available; Sect. 2.2.4) and continued algorithm development (including AI methodologies; Rieutord et al., 2021) are expected to further improve layer attribution efforts.

The lidar signal is strongly attenuated by liquid clouds, so that the signal is often completely extinguished at ~~about 300 m~~ a few hundred meters above cloud base ~~(O'Connor et al., 2004)~~(depending on the system characteristics; O'Connor et al., 2004) . As a consequence, the profile of attenuated backscatter yields little information above the altitude of such thick water clouds. Where clouds form within the ABL, aerosol-derived layer heights can severely underestimate $CBLH$ (Hennemuth and Lammert, 2006). Advanced detection algorithms hence increasingly take into account the presence of boundary layer clouds (Poltera et al., 2017; Caicedo et al., 2020). Where the cloud base height variable provided by the ALC is used, it should be noted that this product can show systematic differences between internal algorithms from various manufacturers (Martucci et al., 2010b;

Pattantyús-Ábrahám et al., 2017) and even between models from the same brand (Liu et al., 2015b). Most methods naturally struggle with reliable layer detection and attribution during precipitation or the passage of synoptic fronts (de Bruine et al., 2017; Yang et al., 2017) as the ABL is generally poorly defined during those times when surface-atmosphere interactions play a minor role compared to larger-scale processes.

Some studies assign a quality flag to the derived layer heights based on the magnitude of the attenuated backscatter vertical gradient (e.g., de Haij et al., 2006; Ketterer et al., 2014). For $ABLH$, such indicators can be suitable given the strong contrasts in aerosol content between the ABL and the FT. However, the strength of the vertical gradient does not necessarily reflect the uncertainty in $MBLH$ detection at night or during morning growth as the contrast between the layer coupled to the surface and the RL is often weaker than to the FT. Careful quality control during post-processing (e.g., based on physically reasonable temporal variations in layer heights) can help focus inter-comparison or evaluation efforts (Kotthaus et al., 2020; Caicedo et al., 2020).

Aerosol-based retrievals for ABL heights are most commonly evaluated against thermodynamic retrievals (Sect. 3.1) applied to radiosonde profile data (Sect. 2.2.1). When comparing aerosol-derived layers to thermodynamic results, the ~~conceptual differences between the two approaches should~~ differences of the underlying physical processes need to be taken into account during interpretation (Hennemuth and Lammert, 2006)~~regarding the physical processes that are being assessed (Sect. 1.1)~~. Naturally, comparison statistics vary with the retrieval method applied on the radiosonde data (Haman et al., 2012). Compared to thermodynamic estimates, aerosol-derived $CBLH$ can have a negative bias (e.g., de Haij et al., 2006; Bravo-Aranda et al., 2017; de Bruine et al., 2017; Liu and Liang, 2010) as the atmospheric quantities of temperature and aerosol show different EZ characteristics (see discussion in Sect. 4.2). Best agreement between the temperature- and aerosol-based layer height detection is again found in the early afternoon, when the CBL extends over the whole ABL (de Bruine et al., 2017). Given the impact of advection on ABL complexity, agreement between the different approaches can vary with synoptic conditions or local circulations induced by surface cover heterogeneities (Pandolfi et al., 2013; Hennemuth and Lammert, 2006). Liu et al. (2018) find agreement between several aerosol-based $ABLH$ results and the bulk-Richardson method applied to daytime radiosonde profiles clearly improves with increasing atmospheric instability.

At night, thermodynamic detection of the height of the CI from radiosondes or AMDAR profiles (Sect. 2.2.1) were found to coincide well with aerosol-derived RLH (Martucci et al., 2007; Kotthaus et al., 2020; Milroy et al., 2012). The few studies showing direct $SBLH$ comparisons between aerosol-derived and thermodynamic results generally suggest a good agreement between the layer estimates, with small biases reported in either direction (Pal et al., 2013; Haman et al., 2012; Tang et al., 2016). Still, substantial systematic biases may occur, with aerosol-derived $MBLH$ remaining below $SBIH$ (Marsik et al., 1995)~~or results from the Heffter method (by about 250 m on average; Lotteraner and Piringer, 2016)~~. Mismatches are explained by the contrasting physical processes that are being traced, such as when radiative cooling leads to the formation of a surface-based temperature inversion which is not necessarily associated with any contrasts in aerosol characteristics (Milroy et al., 2012).

Due to the lack of suitable reference data and the physical difference between aerosol-based and thermodynamic layer detection, a few studies applied manual or semi-automatic layer detection for the evaluation of aerosol-based retrievals (de Bruine et al., 2017; Poltera et al., 2017; Kotthaus and Grimmond, 2018a). Although manual detection can be a very valuable tool, it is labour-intensive and not necessarily objective (Poltera et al., 2017).

4 Monitoring ABL heights

~~Given~~ As ground-based remote sensing profilers have different capabilities (Sect. 2.2) and algorithm uncertainties depend on a variety of atmospheric characteristics, the performance of the various ~~ABL height (Sect. 1.1)~~ sublayer height retrieval methods (Sect. 3) changes throughout the diurnal evolution of the ABL (de Arruda Moreira et al., 2018). The following section summarises the most important strengths and weaknesses of the methods when monitoring the height of the boundary layer at night or during stable conditions (Sect. 4.1), morning growth (Sect. 4.2), peak CBL development (Sect. 4.3), and evening decay (Sect. 4.4). Further, capabilities are discussed that are relevant for the characterisation of the entrainment zone (Sect. 4.5) and the cloud-topped ABL (Sect. 4.6).

The few available synergy applications are highlighted to indicate possible future pathways of ground-based remote sensing implementation. Where observations from multiple ~~ground-based atmospheric~~ profilers are available simultaneously, analyses suggest a synergistic interpretation of results from different methods could lead to an enhanced description of the ABL, including the detection of ~~(sub-)layer~~ sublayer heights (Saeed et al., 2016) and the description of the processes shaping the ~~ABL~~ ABL development (Manninen et al., 2018). ~~The few available synergy applications are highlighted to indicate possible future pathways of ground-based remote sensing implementation~~In addition, combining observations from several sensor systems provides crucial information for the assessment and quantification of layer detection uncertainties (e.g., Cohen et al., 2015) and could be exploited to assign quality flags for the derived layer heights (as has been done for various retrievals from radiosonde profiles by B

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It should be noted that studies directly inter-comparing ABL height retrievals based on different atmospheric quantities are still rare, especially those covering extended time periods. Given their impact on ~~$ABLH$~~ layer height uncertainty, measurement setup (such as MWR calibration, DWL focal setting and scan strategy, aerosol lidar optical overlap, amongst others; Sect. 2.2), data processing, and quality control (Sect. 3) should all be carefully evaluated when comparing results from various methods.

4.1 Nocturnal and/or stable boundary layer heights

At night, the $MBLH$ is rather shallow, with stable conditions being more likely. For the detection of shallow layers, ~~the near-range capabilities of a~~ high vertical resolution of the profile data is beneficial. In-situ data obtained on e.g., tall towers, tethered balloons or using UAS usually provide very detailed information, however, some lack temporal coverage. For radiosondes launched with automatic systems, measurement uncertainties in the lowest part of the profile can pose challenges for the assessment of very shallow layer heights or stability conditions.

For ground-based remote sensing ~~profiler~~ instruments, the near-range capabilities are critical (Sect. 2.2). Layer heights can only be detected if they exceed a potential blind zone of the instrument and they are not obscured by sensor-related uncertainties. MWR are very suitable for shallow layer height detection given their sensitivity is maximal near the sensor

(Sect. 2.2.2). High-power research lidars often do not provide information in the lowest few hundred meters (Sect. 2.2.4), meaning that ALC can be more suitable for the detection of shallow layers from aerosol profiles. ~~A similar advantage is found for DWL in comparison to RWP (Sect. 2.2.3).~~ Improved monitoring of the lowest few hundred metres of the atmosphere at high vertical resolution can be achieved by operating active remote sensing profilers at a low elevation angle (e.g., ALC; Poltera et al., 2017). Scanning DWL (Sect. 2.2.3) can alternate scans at low and high elevation angles, whereby the associated wind and turbulence retrievals have to assume spatial homogeneity of the atmosphere across the sampled volume.

~~Wind, turbulence, temperature , and humidity profiles provide further valuable insights to assess atmospheric stability (Sect.3.1)and the processes leading to the formation of the various layers. In addition to $SBLH$, $SBIH$ can be determined from temperature profiles. Wind and turbulence observations are valuable to detect the nocturnal phenomenon of~~ Where both temperature and wind information are available (e.g. from in-situ observations, or through instrument synergy), $SBLH$ can be derived from the profile of the bulk Richardson number. While results of this method are particularly sensitive to the first-level air temperature observations in the in-situ profile, uncertainty may increase in general when combining results obtained from two sensor types (e.g. DWL and MWR). $SBLH$ can also be approximated from various features in the temperature profile, that also allows for the detection of $SBIH$. In the presence of a LLJ (Sect. 3.2)~~and to~~, wind and turbulence profile data not only allow for the retrieval of $SBLH$ but also help interpret its effects on mixing and advection of moisture, heat and pollutants (Hu et al., 2013; Reitebuch et al., 2000; Bennett et al., 2010). The relation of $SBIH$ to the position of the LLJ changes over time, with $SBIH$ increasing over the course of the night often to exceed the height of the LLJ maximum at some point (Mahrt et al., 1979). In general, discrepancy between temperature-based methods and those analysing vertical wind profiles can be profound (Beyrich and Leps, 2012). Aerosol-based methods can track shallow $MBLH$ if quality observations in the near-range are available and careful layer attribution is performed to reduce confusion with the residual layer height.

Some systematic differences in nocturnal $MBLH$ are reported between results from the various methods available (Sect. 3), with discrepancies between layers detected based on the same or different atmospheric quantities, respectively, at the same order of magnitude. On average, uncertainty in $SBLH$ detection is estimated around 30-40% (Steeneveld et al., 2007). Since turbulence in the SBL is usually not uniform (Beyrich, 1997), the diagnosed layer heights can differ systematically from thermodynamic or aerosol-based methods. While Schween et al. (2014) find turbulence-based nocturnal $MBLH$ exceeds aerosol-based layer heights by about 300 m on average during stormy winters in rural Germany, average nocturnal $MBLH$ differences between turbulence and aerosol-based methods in London, UK, are mostly at the order of their day-to-day variability (Barlow et al., 2011; Kotthaus et al., 2018). More studies are needed to assess the impact of ABL dynamics and atmospheric stability on the relative agreement of the various methods for nocturnal layer height detection.

At night (and early morning), the detection and layer attribution of the $MBLH$ ($SBLH$) can be challenged by the presence of the RL. Layer detection becomes more uncertain if atmospheric characteristics are similar within the MBL and the RL above. As aerosol-based methods are particularly challenged by the presence of a RL, most second generation algorithms (Sect. 3.3) aim to specifically address this source of error. Further, aerosol characteristics (e.g., size distributions) and intermittent turbulence can cause signatures in attenuated backscatter and turbulence fields, respectively, that may appear as additional layer boundaries within the RL. For the turbulence analysis, layer detection can be improved by distinguishing between surface-driven processes in the MBL and the decoupled mixing above (Sect. 4.7). Temperature-based layer heights (Sect. 3.1) derived from MWR profiles (Sect. 2.2.2) are less likely to mistake elevated layer boundaries for $MBLH$ as these profilers are more reliable in the near range and respond less to RL signatures.

The RL is a remnant of the previous day's CBL (Sect. 1.1), so that aerosols and moisture remaining in this elevated layer above the $MBLH$ present very suitable atmospheric tracers. While some turbulent exchange between the RL and the FT can be picked up (Fochesatto et al., 2001), the RLH can be tracked most reliably using thermodynamic retrievals (Sect. 3.1) of the CI applied to airborne in-situ sensors or by aerosol-based methods (Sect. 3.3) because the contrasts at the $ABLH$ are usually striking. As MWR profiles are generally less sensitive to contrasts near the RLH (Sect. 2.2.2) algorithms are usually not applied to radiometer profiles for the detection of this RLH. Uncertainty in RLH detection can be increased by various atmospheric processes, such as low aerosol load (reducing SNR for lidar systems; Sect. 2.2.4), ~~advection of air masses with different aerosol or humidity content or~~ the presence of shear layers (generated by e.g., orography, LLJ) or the advection of air with different aerosol or humidity content.

Instrument synergy has been identified as a promising means to better characterise the nocturnal boundary layer. Exploiting a combination of attenuated backscatter and turbulence variables derived from DWL profiles, Manninen et al. (2018) present a synergy approach to characterise the RL as the non-surface-connected region of the ABL where the turbulence activity is intermittent or absent. Collaud Coen et al. (2014) and de Arruda Moreira et al. (2020)

Several studies highlight that synergy analysis of MWR and aerosol ~~lidars~~ lidar data is particularly promising for nocturnal layer assessment given the respective strengths in observing $SBLH$ and RLH features (Collaud Coen et al., 2014; de Arruda Moreira et al

. Saeed et al. (2016) use information on temperature inversion heights derived from MWR profile data to constrain an aerosol-based $SBLH$ retrieval.

4.2 Morning growth

The time of $CBLH$ morning growth is characterised by substantial temporal variations (Halios and Barlow, 2017), especially where ~~solar energy input is significant (Sect. ??). Radiosonde ascends are rare between sunrise and solar noon~~ the surface energy balance exhibits a pronounced diurnal cycle. For the detection of the $CBLH$ from radiosonde profiles, again the bulk Richardson method is available but also its simplified version based on temperature profile data alone, i.e. the Parcel method, is here applicable (amongst others; Sect. 3.1).

Compared to the limited temporal coverage of radiosonde ascents (Sect. 2.2.1)~~. Given the different stages of CBL development (Halios and Barlow, 2017)~~, the continuous monitoring enabled by remote sensing profilers is a clear advantage for the ~~characterization~~ assessment of this period ~~compared to balloon ascends~~characterised by significant seasonal but also day-to-day variability. Approaches based on high-frequency variations of wind (Sect.3.2) or aerosol (Sect. 3.3) often reveal pronounced signals near the $CBLH$ during this time of day. Given their ability to capture turbulence even in the near range (Sect.2.2), sodar systems are particularly valuable for the monitoring of the growth (onset) of the $CBLH$ (Beyrich, 1995). Turbulence-based $CBLH$ from RWP usually requires longer integration times (20-60 min) compared to DWL or ALC that both range in the order of minutes.

As for nocturnal conditions (Sect. 4.1), the presence of a RL can affect the detection of $CBLH$ during morning growth as entrainment of RL air (instead of air from the FT) can act to reduce the contrasts of measured quantities near the $CBLH$. Among aerosol-based approaches, detection methods that account for the potential presence of a RL in addition to $CBLH$ are more reliable (Sect. 3.3). ~~Analysis of RWP data allows for the~~ The detection of $CBLH$ ~~morning growth, once~~ from RWP data during morning growth relies on the careful differentiation between the turbulent signature ~~from~~ generated by the entrainment of RL air into the CBL ~~is stronger than that at~~ and variations near the RLH (Bianco et al., 2022).

Based on selected case studies, turbulence- and aerosol-based $CBLH$ during morning growth are often very similar provided appropriate layer attribution is performed (Cohn and Angevine, 2000; Collaud Coen et al., 2014). However, several studies also report a temporal delay of aerosol-derived $CBLH$ morning growth both relative to temperature-derived $CBLH$ (Wang et al., 2012; Kotthaus et al., 2020) and turbulence-derived $CBLH$ ~~growth~~ results (Wiegner et al., 2006; Barlow et al., 2011; Kotthaus et al., 2018), with time lags of up to two hours. Presumably, it can require some time before aerosols emitted at the surface and transported upwards by turbulent mixing establish a clear layer boundary relative to the RL. In addition, entrainment of RL air with lower humidity and aerosol load may delay morning growth of aerosol-based $MBLH$ (Gibert et al., 2007). Some studies found turbulence-derived $CBLH$ to not only start rising earlier but also to grow faster than layer heights from aerosol-based methods (Barlow et al., 2011; Schween et al., 2014). However, this may be partly linked to the response of the respective detection algorithms to the presence of clouds (Wiegner et al., 2006), as Kotthaus et al. (2018) found similar growth rates when looking at cloud-free conditions only. No clear picture has yet emerged on a potential time lag between the growth of temperature- and turbulence-derived layer heights (de Arruda Moreira et al., 2018, 2019). Due to advances in algorithm development (Sect. 3), multi-sensor analysis has the potential to better quantify the relation of layer heights based on thermodynamic, dynamic and aerosol-based retrievals, respectively, which is expected to provide valuable new insights into ~~morning~~ the understanding of ABL dynamics during CBL development.

4.3 Daytime Convective Boundary Layer

Most methods ~~(Sect. 3) show very good performance~~ for $MBLH$ detection perform best during daytime, especially ~~when~~ once the $CBLH$ coincides with $ABLH$ (Sect. 1.1). Provided sufficient SNR and careful data processing, $CBLH$ from all retrieval methods can agree within a few hundred metres or even less (Granados-Muñoz et al., 2012; Renju et al., 2017). If radiosonde ascends at noon are compared to ground-based remote sensing profile data, the CBL may not yet reach its full extent so that layer attribution (i.e. confusion of $CBLH$ and RLH) can be a general source of uncertainty. Sensors restricted by low SNR (such as, e.g., sodar, RASS, ~~early ALCmodels~~some ALC; Sect. 2.2), ~~might not always reliably observe the fully developed~~ often fail to detect the fully-developed CBL in the afternoon, especially where boundary layer development is deep~~and/or~~, aerosol load is low, or environmental noise levels are high (Boy et al., 2019).

Schmid and Niyogi (2012) highlight that a thick EZ, likely to occur during deep afternoon convection ($>$ 3 km), can result in a weaker delineation at the $ABLH$, increasing uncertainty in layer detection for all methods. While turbulence-based algorithms are challenged in the presence of strong shear layers above the ABL (Marsik et al., 1995), elevated aerosol layers increase the likelihood of false layer attribution for aerosol-based techniques (Granados-Muñoz et al., 2012; Tang et al., 2016). Layers holding advected aerosol with characteristics differing from local emissions (e.g., long-range transport of desert dust) may further alter the air temperature profile (Guerrero-Rascado et al., 2009), whereby potentially inducing errors in the applied thermodynamic retrieval (Granados-Muñoz et al., 2012). Synoptic circulation or orography-induced flow patterns that are influencing cloud conditions or the advection of decoupled layers have hence been found to affect comparison statistics (Granados-Muñoz et al., 2012; Pandolfi et al., 2013; Tang et al., 2016; Pearson et al., 2010; de Arruda Moreira et al., 2019).

Daytime maxima of the layer estimates from temperature-, wind- or turbulence-, and aerosol-based methods are most similar in clear-sky conditions (Barlow et al., 2011; de Arruda Moreira et al., 2018). In accordance with the delayed morning growth of aerosol layers (Sect. 4.2), ~~studies often~~ some studies find $CBLH$ from aerosol-based methods to peak up to 2 h later than layer heights diagnosed from turbulence (Barlow et al., 2011; Kotthaus et al., 2018) or temperature profiles (Renju et al., 2017). No clear relation has yet been established between peak daytime $CBLH$ from aerosol-based retrieals and either turbulence or thermodynamic methods, as negative (Barlow et al., 2011; Kotthaus et al., 2018), positive (Schween et al., 2014; Granados-Muñoz et al., 2012; de Arruda Moreira et al., 2019), as well as no bias (Collaud Coen et al., 2014; Wang et al., 2012) are reported. Similarly, both positive, negative, and negligible deviations were found when comparing turbulence-derived $CBLH$ based on DWL data and temperature-based results obtained from MWR profiles (de Arruda Moreira et al., 2018, 2019). Further research is needed to assess which factors (including algorithm uncertainty, sampling strategy and atmospheric dynamics) best explain potential biases, accounting not only for cloud dynamics and the presence of elevated aerosol layers but also the presence of thermals overshooting the inversion at the ABL top which are inducing vast temporal variability (Renju et al., 2017; de Bruine et al., 2017) or atmospheric stability and moisture transport which can affect $CBLH$ growth rates (Helbig et al., 2021).

4.4 Evening decay

The daytime CBL transitions into the nocturnal boundary layer around sunset (Sect. 1.1). The decay in surface-driven buoyancy is directly monitored by the turbulence profiles obtained by sodars, DWL, or RWP (Sect. 2.2.3). Thermodynamic and turbulence-based $CBLH$ are in general agreement, showing a gradual decrease in the afternoon up to about sunset (Wang et al., 2012; Collaud Coen et al., 2014; Renju et al., 2017). Schween et al. (2014) illustrate the breakdown of turbulent exchange in the afternoon based on DWL profile data while Manninen et al. (2018) highlight elevated turbulence can occur during the evening transition as the RL decouples. Layer detection from RWP observations was found more uncertain in the presence of elevated shear layers (Ketterer et al., 2014).

In response to the vanishing buoyancy, aerosols start to settle in the afternoon whereby slowly forming new layer boundaries. As aerosol-based $MBLH$ is a record of the history of recent turbulence activity (Träumner et al., 2011), layer attribution is especially challenged during the time of evening CBL decay (Sect. 3.3.2). Where aerosol emissions at the surface are high (e.g., in cities), new shallow aerosol layers tend to become visible around sunset as the reduced vertical dilution increases low-level concentrations. If no clear aerosol gradient forms close to the surface around sunset (Sect. 4.1), the RLH may at times be misinterpreted as $MBLH$ in which case evening discrepancies between layer results from different methods can be at the order of magnitude of the $ABLH$. Reliable data in the near range, careful processing algorithms and high surface aerosol emission rates increase the likelihood of this transition time to be captured ~~accurately~~ (accurately) by ALC (Sect. 3.3.2).

4.5 Entrainment zone

Characterising the entrainment zone (Sect. 1.1) around the $CBLH$ can greatly benefit interpretation of ABL dynamics, local climate conditions and air quality. The ~~EZ thickness~~ EZD can be estimated ~~either based on gradients of observed quantities between the CBL and FT (RL) or by exploiting temporal~~ from temporal (or spatial) variations in $CBLH$ (Sect. 3) ~~. de Bruine et al. (2017) report differences between alternating thermals in the EZ~~ that are a direct measure of the entrainment process, provided observations are collected at very high temporal resolution. Aerosol profiles with sufficient SNR are rather suitable for the assessment of EZD. For example, de Bruine et al. (2017) report fluctuations of 100-500 m ~~, fluctuating~~ around an average, rather constant $ABLH$.

~~When aerosol-based $CBLH$ is derived at very high resolution (seconds-minutes), its temporal variation captures clear FT air being mixed into the ABL (Sect. 3.3). Turbulence-based~~ Cohn and Angevine (2000) find EZD based on fluctuations of $ABLH$ from aerosol-derived methods to be more consistent than those from RWP observations, as the latter are associated with greater noise levels for the sensor used. Where turbulence-based $CBLH$ from DWL ~~may require longer~~ requires rather long integration times (Sect. 3.2), ~~which can be~~ this is a disadvantage for the estimation of ~~EZ thisckness (Träumner et al., 2011) based on $ABLH$ variability. Cohn and Angevine (2000) find EZ thickness based on fluctuations of $ABLH$ from RWP to be greater than those based on variations of aerosol-derived $ABLH$, as the former is associated with greater noise levels for the sensor used~~EZD (Träumner et al., 2011). For shallow CBL, also sodar observations have been successfully exploited to determine the EZD (Beyrich and Gryning, 1998).

Alternatively, the EZD can be approximated based on gradients of mean observed quantities between the CBL and FT (RL), derived from e.g. radiosonde data. The rather low range resolution of the MWR near the $ABLH$ (Sect. 2.2.2) can cause considerable uncertainty when studying the EZ based on remotely-sensed temperature profiles (Wang et al., 2012).

4.6 Cloud-topped boundary layer

While many studies focus on the analysis of clear-sky conditions, cloud-topped mixed layers are starting to receive increasing attention. Given the diverse capabilities and limitations of the remote sensing profilers for the observation of clouds (Sect. 3), the disagreement between layer estimates generally increases with cloud complexity (e.g., Cohn and Angevine, 2000; Collaud Coen et al., 2014; Cimini et al., 2013; Emeis et al., 2009). The strong attenuation of the lidar signal by water clouds causes a distinct signature in ALC and DWL profiles. Developed to record cloud base ~~altitude~~height, ALC inherently have built-in algorithms reporting this quantity. It should however be noted that cloud base height detection methods vary between manufacturers (Sect. 3.3) so that generalised algorithms may need to be applied during post-processing when consistent products are required across a diverse sensor network. High-power lidar systems are usually not operated under cloudy conditions (Sect. 2.2.2, 2.2.4). MWR and RWP can penetrate the layer of cloud droplets. RWP observations have been used to determine the cloud top but frequent false detection was linked to elevated layers of high humidity or turbulence (Collaud Coen et al., 2014).

Doppler cloud radars (DCR) can be used to characterize the vertical extent of boundary-layer clouds, such as shallow convective clouds, stratiform clouds (stratus or stratocumulus), and even fog. In the case of adiabatic fog, the fog layer (typically 50-400 m deep) is destabilized due to strong radiative cooling at the top coinciding with strong temperature inversion. Mixing then occurs between the fog top and the surface. Wærsted et al. (2017) combine measurements from MWR and DCR to retrieve the temperature profile in adiabatic fog layers, hence characterizing precisely the depth of the mixing.

Automatic ABL height retrieval algorithms are increasingly incorporating the presence of clouds into the layer detection and attribution process (Poltera et al., 2017; Caicedo et al., 2020). Both cloud cover and cloud type can be critical ~~(Sawyer and Li, 2013)~~(Sawyer and Li, 2013; Kotthaus and Grimmond, 2018b), given convective clouds are associated with surface-driven turbulence and stratiform clouds initiate mixing by cloud dynamics (Hogan et al., 2009). Turbulence characteristics present useful information to differentiate between surface- and cloud-driven turbulence (Harvey et al., 2015). ~~For shallow Cu clouds, the CBH~~ The cloud base height may be used as a reasonable proxy for $CBLH$ for shallow Cu clouds, where the convective nature of the cloud can be assessed by surface heat flux measurements (Schween et al., 2014; Wiegner et al., 2006) or derived from remote sensing data (Manninen et al., 2018). Depending on the retrieval applied to determine the ~~CBH~~cloud base height, potential biases may be introduced. For deeper Cu, the relation between cloud base and $CBLH$ is more ambiguous. In autumn and winter, the cloud base ~~altitude~~ height is often related to the dissipation of Sc clouds or fog processes (Schween et al., 2014). The relation of ~~CBH~~ cloud base and $MBLH$ is subject to ongoing research.

4.7 Atmospheric stability and ABL classification

ABL dynamics ~~respond to~~ form in response to a complex combination of processes, including e.g., surface forcing, the synoptic flow (Shi et al., 2019), ~~local-scale circulations induced by land cover contrasts (Moigne et al., 2013) or topography (Rotach and Zardi, 2007), surface forcing or~~ elevated sources of turbulence associated with clouds or winds (e.g., LLJ), local-scale circulations induced by land cover contrasts (Moigne et al., 2013) or topography (Rotach and Zardi, 2007). To understand the relative importance of ~~all~~ these drivers in defining ABL sublayer heights, automatic classification methods are increasingly developed. ~~Probably the most~~

A common ABL classification is the delineation between cloudy and cloud-free conditions, which can be accomplished using surface radiation data, the cloud information reported by an ALC (Sect. 2.2.4) or by exploiting any remotely sensed profile signal sensitive to clouds (Sect. 2.2). To account for differences in ABL heights associated with cloud dynamics, ALC data have also been used to automatically distinguish simple cloud types (Kotthaus and Grimmond, 2018a). Pal et al. (2013) classify ABL regimes using cloud cover (cloudy vs clear-sky) and atmospheric stability (from surface observations) to distinguish between days dominated by local surface-driven buoyancy and those dominated by larger scale events. Using airborne profile measurements, Mahrt (1991) find ABL humidity exchanges are generally either associated with an *entrainment-drying* regime ~~with~~ characterised by a vertical divergence of the moisture flux or a *moistening* boundary layer dominated by surface evaporation fluxes.

~~As turbulence is not only driven by surface buoyancy but can also be generated aloft (~~Atmospheric stability indicators that account for both thermal buoyancy and wind shear (such as the Richardson number), can be obtained from suitable in-situ observations (Sect. 2.2.1) or a synergy of wind and temperature profiles observed by multiple systems. Based on temperature profiles alone, lapse rates and other indices of thermal atmospheric stability can be derived (Feltz and Mecikalski, 2002; Wagner et al., 2008 and used to classify the ABL regime (Liu and Liang, 2010).

But also from wind- and turbulence alone, valuable characteristics can be determined, e.g. ~~, by LLJ or cloud processes) additional details on the ABL dynamics can be obtained from higher order moments of turbulence observations~~to differentiate between buoyancy- or shear-driven turbulence (Tucker et al., 2009), surface or elevated turbulence sources (Tonttila et al., 2015; Manninen , and elevated turbulence sources associated with the flow (e.g., LLJ; Tuononen et al., 2017) or cloud dynamics (Marke et al., 2018; Harvey . The sign of the vertical velocity skewness provides information on the source of turbulence, with positive values typical for surface-driven buoyancy in clear-sky CBL and negative values associated with cloud-topped boundary layers dominated by 'downwards convection' driven by radiative cooling at the cloud top (Hogan et al., 2009). The vertical velocity skewness further helps to determine whether clouds are coupled to the surface (e.g., shallow cumulus clouds) or decoupled from the surface (e.g., some nocturnal Sc; marine Sc).

~~Combining the various physical quantities that describe atmospheric turbulence (Sect. 2.1) with wind shear information enables detailed classification of the ABL, including the identification of the MBL stability regime (Banta et al., 2006), and the differentiation between (i) turbulence driven by buoyancy or shear (Tucker et al., 2009), surface or elevated turbulence sources (Tonttila et al., 2015; Manninen et al., 2018; Huang et al., 2020; Harvey et al., 2015), elevated turbulence sources associated with the flow (e.g. LLJ, Tuononen et al., 2017; Marke et al., 2018) or cloud dynamics (Harvey et al., 2013; Manninen et al., 2018)~~ The SBL turbulence regime from velocity variance indicators was found consistent with stability indicators derived from AERI temperature profiles (Bonin et al., 2015) and useful for the characterisation of LLJ evolution (Bonin et al., 2019).

Building on the profile-based classification approach from Harvey et al. (2013), Manninen et al. (2018) developed a pixel-based ABL classification scheme that exploits several atmospheric quantities derived from DWL observations. An example of a clear-sky case (Fig. 4.1) illustrates the complexity in ABL dynamics ~~(Manninen et al., 2018)~~ with diurnal variations clearly detectable from the profile data (Manninen et al., 2018). Unstable atmospheric conditions drive the $CBLH$ morning growth in two stages (Fig. 4.1a,b), i.e. the slow increase of near surface convective conditions followed by a rapid growth phase (Halios and Barlow, 2017). The gradual decay of convective activity in the evening transition is clearly detected (Fig. 4.1a,b). A LLJ forms at some point after sunset (Fig. 4.1c,d).

~~Time-height plots of atmospheric boundary layer classication using the Manninen et al. (2018) scheme showing (a) whether mixing is connected to the surface or cloud driven and (b) the turbulent mixing source, together with time-height plots of (c) wind direction and (d) wind speed on 28 August 2017 at Jülich, Germany. The black lines on the two lower panels show low-level jet (LLJ) altitude (Tuononen et al., 2017).~~

5 ~~ABL climatology~~

~~Advances in measurement technology (Sect. 2.2), the operation of coordinated measurement networks (Sect. 2.3), and retrievals of layer heights (Sect. 3) increasingly allows for ABL climatology studies to be performed. To give an overview on the~~

~~insights gained from long-term observations (here with a duration of at least one year), characteristics of ABL heights are summarised at the global scale (Sect. ??) and specifically over land (Sect. ??) and for marine environments (Sect. ??). Finally, it is highlighted where long-term trends in ABL heights start to emerge. While most studies considered here are using a combination of airborne in-situ or ground-based remote sensing observations, some satellite measurementsand modelling studies are included to highlight the detected similarities and differences in ABL characteristics.~~

4.1 ~~Atmospheric boundary layer heights at the global scale~~

~~Figure ?? provides a global picture of the midday $ABLH$ obtained from ECMWF ERA-Interim reanalysis (Figure ??a; von Engeln and Tei~~ ~~, satellite-based CALIPSO observations (Figure ??b; ?), and an ensemble mean of global circulation models (Figure ??c; ?) . $ABLH$ peak values are usually bound to convective conditions driven by high surface temperatures that occur in response to strong solar irradiance, low surface albedo and low soil moisture as found~~ Based on profiles of the aerosol scattering ratio derived from ALC measurements, the complex ABL dynamics in an Alpine valley were classified which proved valuable for the assessment and understanding of local air quality conditions (Diémoz et al., 2019a, b). Studies that group aerosol or trace gas profile observations according to e.g. ~~, over dry subtropical land regions such as the Saharan desert in summer (??Guo et al., 2016; ?; Aryee et al., 2020). Over the subtropical deserts in Northern Africa and Australia, monthly average $ABLH$ >3 km a.g.l. have been reported (Ao et al., 2012) with daytime maxima exceeding 5 km. Large marine $ABLH$ values (>1.8 km) are found in the intertropical convergence zone, for example over the South Atlantic east of Brazil, the Gulf of Aden and over the western Indian Ocean. A particular phenomenon with very high $ABLH$ (up to 5 km a.g.l. corresponding to 9 km a.s.l.) occurs over the Tibetan Plateau in the winter season (??), when a deep neutral layer prevails so that~~ turbulence, stability, or LLJ regimes (e.g., Su et al., 2020; Klein et al., 2017, 2019; Dieudonné et al., 2013), characteristic features are starting to emerge that can be very valuable for the assessment of near-surface air pollution concentrations. The gained processes-understanding of such synergy applications that combine multiple atmospheric variables, means ABL classifications are increasingly able to account for the complex interactions of the ~~$ABLH$ approaches the upper troposphere. The lowest $ABLH$ are found in winter over polar regions when strong inversions near the ground dominate (von Engeln and Teixeira, 2013) . As a general rule, lower $ABLH$ but with a large seasonality are found over high latitudes, whereas a weaker seasonality is found between 30° and 50° of latitude (Chan and Wood, 2013).~~

~~Large-scale synoptic flow and land-sea-atmosphere interactions lead to contrasting $MBLH$ characteristics between eastern and western coastlines. In the Northern hemisphere, western coastlines of continents are predominantly covered by stratocumulus clouds, whereas the eastern coasts experience a clear seasonality. Here, $MBLH$ is relatively high during winter due to cold air outbreaks and frontal systems and low during summer due to frequent fog. As a result, average $MBLH$ are lower at eastern coastlines (0.5-1.0 km) than for western coastlines (∼2 km) in the Northern hemisphere (von Engeln and Teixeira, 2013)~~ processes that drive ABL dynamics and atmospheric composition.

~~Very long observational records now allow for the examination of inter-annual variations in seasonal and diurnal variability (e.g., ?Wang et al., 2020). Such studies demonstrate that regional to global scale phenomena such as heat waves or the El Niño Southern Oscillation (ENSO) have an imprint on ABL heights.~~

4.0.1 ~~The urban boundary layer~~

~~The structure of the urban boundary layer is increasingly considered for the interpretation and modelling of air quality and greenhouse gas emissions (e.g., Klein et al., 2017; Lauvaux et al., 2016; Shi et al., 2019; **?**; de Arruda Moreira et al., 2020; **?**). Observational evidence reveals that ventilation (i.e. a combination of horizontal advection and vertical dilution)is a particularly important driver for near-surface pollutant concentrations when buoyancy is weak (e.g., Lee et al., 2019; Stirnberg et al., 2021; Sujatha et al~~

5 Conclusions

Despite the importance for a range of applications, quantitative knowledge on the temporal and spatial variations in atmospheric boundary layer (ABL) height is still scarce. While synchronised radiosonde data provide an immensely valuable assessment of conditions at the global scale now for decades, their comparatively low temporal resolution (amongst other limitations) means they often do not capture diurnal and seasonal variations of ABL dynamics at a given location. Thanks to advances in ground-based remote sensing, high-quality profile observations spanning the ABL extent at very high temporal resolutions are now increasingly collected by operational measurement networks. As these data start to resolve spatio-temporal variations in ABL dynamics even in complex environments such as mountainous terrain or cities, they provide valuable contributions to international research projects in these domains (e.

~~Short-term measurement campaigns (for example in Paris or London; Pal et al., 2012; Barlow et al., 2015) have shown that the urban CBL is often deeper than over the suburban and even more over the rural surroundings (see review by Barlow, 2014) . The enhanced buoyancy (and reduced moisture) over the urban surface has even been linked to higher cloud base altitudes of boundary layer clouds (Theeuwes et al., 2019). The well-studied urban heat island effect also influences $MBLH$ at night, with layer heights over the city centre systematically higher (up to 200-300 m) and more heterogeneous (Pal et al., 2012). Urban areas can maintain slightly unstable stratification well into (or even throughout) the night, suggesting nocturnal low-level jets are initiated over rural areas instead (Barlow et al., 2015).~~

~~Long-term climatological observations of the urban boundary layer have been performed in several cities, including Athens, Greece (Kokkalis et al., 2020), Beijing, China (Tang et al., 2016; **?**), Granada, Spain (Granados-Muñoz et al., 2012; de Arruda Moreira et a , Hong Kong, China (Huang et al., 2020), Houston, Texas, USA Haman et al. (2012), Leipzig, Germany (Baars et al., 2008), London, UK (Kotthaus and Grimmond, 2018b), Paris, France (**?**), Shanghai, China (Peng et al., 2017), Vancouver, Canada (**?**) , Vienna, Austria (Lotteraner and Piringer, 2016), Yuen Long, China (**?**), and Warsaw, Poland (Wang et al., 2020). However, only a few studies include a rural reference site. Lotteraner and Piringer (2016) analyse one year of $MBLH$ observations in Vienna at a central urban and a rural site and found an increment of $CBLH$ over the city with a mean midday difference <100 m. This difference was explained not only by enhanced vertical mixing over the city, but also by local topographic features that further support a more rapid urban morning growth rate. Dense observational networks are starting to emerge in urban settings, responding to to the need of a more detailed understanding of the urban boundary layer in a wide variety~~

~~geographical settings and across the diverse and heterogeneous characteristics of the urban surface and anthropogenic activities.~~

5.0.1 ~~The boundary layer over complex topography~~

~~Although the $MoBL$ is affected by severe spatial and temporal complexity, some general characteristics and mechanisms have been quantified. As for the ABL development over other land regions, the heights of the various $MoBL$ sublayers and the resulting exchanges with the FT are highest (lowest) during local summer (winter) time. This has been observed for CBL over the valley floor (**?**), the aerosol layer (AL) (Poltera et al., 2017), plain-to-mountain venting (**?**), as well as the altitude of the valley winds (**?**). High-altitude sites are hence often sampling FT air during winter (**?**). Still, the ABL influence at a given mountain measurement site needs to be assessed carefully because the station altitude is not necessarily the main criterion (**?**) . In a valley with dense urbanization, pollution episodes are most severe in cold conditions when the $CBLH$ remains lower than the ridge altitude (**?**).~~

~~The occurrence of up-down valley winds (adabatic and katabatic winds) is directly bound to the times of sunrise and sunset (**?**) and are originate at the valley floor (**?**). Slopes exposed to morning sunshine cause earlier slope wind development, an erosion of the nocturnal thermal inversion and the set-in of a cross-valley circulation (**?**). $CBLH$ maxima in the valley floor are higher and occur earlier than those on the slope of the valley (**?**). The $MoBL$ tends to follow the terrain in the morning and to flatten to a level of 0.5-2.0 km over range altitude during the afternoon (De Wekker and Kossmann, 2015; Pal et al., 2012; **?**) . As long-term ABL measurements remain scarce over complex topography, many characteristics of the $MoBL$ are still poorly understood. More ABL profile observations representing the diversity in mountainous topographies, soil properties and mesoscale influences are required to reach a more global comprehension of the MoBL complexity.~~

5.1 ~~Marine environments~~

5.0.1 ~~Atmospheric boundary layer over the ocean~~

~~The marine ABL exhibits similar dynamics as the boundary layer over land (Sect. **??**), however,~~ g. the ~~lower variability in sea surface temperature, marine air temperature, and surface sensible heat fluxes cause structural differences in layer heights. While the daytime maxima in $MBLH$ are smaller compared to over land (Figure **??**b; **?**)), the daily average marine $MBLH$ is deeper overvall (Figure **??**a,c; von Engeln and Teixeira, 2013; **?**)) due to its reduced diurnal amplitude (**?**Liu and Liang, 2010). Some studies find ocean $MBLH$ peak values around solar noon (Liu and Liang, 2010), but the marine $MBLH$ diurnal cycle is often rather indiscernible, which can be explained by a persistent capping inversion (**?**von Engeln and Teixeira, 2013; Chan and Wood, 2013; **?**; **?** . Over land, significant amounts of heat and mass are exchanged between the ABL and FT over the course of the day (Stull, 1988). Over the ocean again, the overall lower exchanges are mostly accomplished by cloud dynamics of the predominant Sc- or Cu- topped ABL (**??**). CBL conditions are more frequent over the oceans with unstable stratification even persisting at night. The neutral RL regime is more rare over the ocean during daytime, with its frequency reduced by 20 %–33 %~~

~~compared to the average land boundary layer (Liu and Liang, 2010)~~EU H2020 Green deal projects *RI-URBANS*[21] (ACTRIS) and *ICOS-cities*[22] (ICOS) or the WWRP-Endorsed Project *TEAMx*[23]).

~~The greatest seasonal amplitude of marine $ABLH$ occurs over the Arctic (Chan and Wood, 2013). A seasonal asymmetry is observed about the equator with maxima during winter in the Northern hemisphere and in autumn (dry season) in the South, respectively. Seasonal variability again is more pronounced over the Northern hemisphere driven by the larger land area (Chan and Wood, 2013). Significant seasonal amplitudes in oceanic $ABLH$ do occur over the Mediterranean Sea (von Engeln and Teixeira, or the Gulf Stream (Seidel et al., 2012). A longitudinal dependence of marine $ABLH$ has been detected by satellite and model analyses over the Pacific ocean with higher values at -85° (1.5 km) than at -70° (1 km) (Ho et al., 2015, figure 11).~~

5.0.1 ~~Coasts, islands and lakes: a mix between marine and land signatures~~

~~The ABL climatology for coastal regions reflects the combination of sea-influence and local meteorological phenomena (e.g., induced by particular topography and /or land cover). Similarly, large lakes have been associated with marine boundary layer characteristics (Liu and Liang, 2010; Moigne et al., 2013; **?**), with effects becoming more pronounced with increasing lake area. Coastal $ABLH$ show less annual variability than inland, likely due to the moderating influence of the sea (Seidel et al., 2012; **?**) . Regional dynamics such as local breeze circulations responding to thermal contrasts between the lake or ocean and the adjacent land have systematic implications for the boundary layer heights. Circulations can lead to rapid decreases in $ABLH$ at the transition between the land and sea breeze regime. Through such dynamics, shallower average $ABLH$ often occur in summer rather than in autumn or even winter, as found for e.g., Miami, Brookhaven and New York, USA (**??**), Denmark (**?**) , Athens, Greece, and Lecce in Italy (**?**), or Shanghai, China (Peng et al., 2017)). At other coastal locations (e.g., Vancouver, Canada), the average $ABLH$ is highest in summer, with peak daily maxima in spring, when added buoyancy provided by latent heat release associated with boundary-layer cloud formation enhances growth (**?**). **?** demonstrate that latitude is the most important factor determining ABL stability for larger lakes (area greater than 10 km^2) , as unstable CBL become increasingly frequent at lower latitudes. In addition to land-sea contrasts, coastal topography can play an important role. For example, **?** associate the observed minimum average $ABLH$ (41-108 m a.s.l.) during summer at four sites in California's Central Valley, USA, with topographically forced cold-air advection and local land-use characteristics.~~

~~The $ABLH$ diurnal cycle characteristics for oceanic islands presents a superimposition of the continental and marine ABL processes (Liu and Liang, 2010), with the ABL over volcanic islands combining the complex terrain effects (such as mountain venting processes) and maritime phenomena.~~

5.1 ~~Cloudy versus clear sky conditions~~

~~$CBLH$ growth rates and maxima are strongly affected by clouds as the latter modulate the incoming solar radiation and hence the energy available for buoyancy production. However, to date the majority of $ABLH$ climatology studies focus on~~

[21] https://riurbans.eu/

[22] https://www.icos-cp.eu/projects/icos-cities-project

[23] http://www.teamx-programme.org/

~~clear-sky conditions to avoid uncertainties in layerretrievals associated with clouds (Sect. 4.6). Where clouds are considered during $ABLH$ analysis, studies mostly focus on cloud cover as local information on cloud type is still scarce. Seasons with greater cloud coverage usually exhibit greater variability in measured $ABLH$ (?). Some studies (e.g., Payerne, Switzerland; Collaud Coen et al., 2014, Jülich, Germany; Schween et al., 2014) find no clear differences in CBL phenomenology, whereas others report pronounced effects of cloudy skies (such as $CBLH$ on average greater by 100-500 m on cloud-free days Guo et al., 2016). The nocturnal $SBLH$~~ As dense measurement networks are emerging across Europe and other parts of the world with high spatial coverage, harmonisation of operations, data processing and layer height retrievals are key. International operational networks (such as, e.g., E-PROFILE, ACTRIS, ICOS, ARM), not only collect and archive the observations but also strive to harmonise sensor settings and standardise file formats. Further does coordination between different sensor networks receive increasing attention as this clearly benefits synergy applications. Close collaborations with NHMS, academia and instrument manufacturers, are vital to formulate and implement standard operating procedures, to closely monitor house-keeping data, and ~~$SBIH$ climatology observed over the Swiss plateau (Collaud Coen et al., 2014) presents clear differences between clear-sky and cloudy conditions, as the layer heights are confined to ~500–700 m for cloud base heights between 500-2400 m but can expand up to 1000 m otherwise.~~ to develop both detailed correction procedures and advanced data products, including the heights of the ABL sublayers. In Europe, several EU COST actions were paramount for the exchange of knowledge and best practices, including Action 710 (Harmonisation of the pre-processing of meteorological data for atmospheric dispersion models; Seibert et , EG-CLIMET (European Ground-based observations of essential variables for CLImate and METeorology; Illingworth et al., 2015) , and TOPROF (Towards Operational ground based PROFiling with ceilometers, Doppler lidars and microwave radiometers for improving . Following the progress made in this field over recent decades, the action PROBE[24] (PROfiling the atmospheric Boundary layer at European focuses on the harmonisation of operational procedures which is necessary to ensure also higher-level products are comparable across Europe and even globally.

~~Cloud-cover alone may be insufficient to assess the impact of clouds on boundary layer dynamics (Sect. 4.6; ?). Especially differentiating between boundary layer clouds and those decoupled from~~ This review outlines how ground-based remote sensing methods are best exploited in order to gain a detailed understanding of ABL sublayer heights and dynamics. Firstly, the ~~ABL (?) greatly aids interpretation. For both Paris, France (?) and London, UK (Kotthaus and Grimmond, 2018b), aerosol-derived $MBLH$ (Sect. 3.3.1) was found to vary with cloud type. Daytime maxima are lowest for stratiform clouds and highest for convective clouds, with cloud-free $MBLH$ in-between. Peak $CBLH$ development was detected on days when broken convective clouds formed at the ABL top after a clear-sky night~~capabilities and limitations of various measurement technologies available to capture different atmospheric profile variables within the ABL are summarised. Choosing the appropriate technology for a given network not only needs to consider the physical information content of the atmospheric quantity observed (temperature, humidity, wind, turbulence, aerosol, or gas) but also whether the sensitivity, resolution, and capabilities of a given sensor are appropriate to monitor the layer(s) of interest. Such instrument characteristics not only differ between sensor types but also between models from different manufacturers and can depend even on firmware, hardware, instrument settings, instrument age, or sampling strategies.

[24] http://www.probe-cost.eu/

5.1 ~~Trends in atmospheric boundary layer heights~~

~~Trend analysis requires a homogeneous measurement time series of at least 10 years. As such time series from~~ Certain ground-based remote sensing ~~observations are only starting to become available, published trend analyses mostly rely on radio-sounding time series (??????) or on the data assimilation system ERA-interim reanalysis (??). So far, trend analyses focus on daytime conditions given the higher detection uncertainty (Sect. 4.1) combined with smaller magnitudes of nocturnal layer heights make it more challenging to assess statistically significant long-term trends.~~

~~Positive, statistically significant trends in daily $ABLH$ have been detected over Europe (1973-2010; ?)), China (1979-2003; ?)), Iran (1979-2016; ?)), the Sahara Desert and Arabian Peninsula (1979–2019 ?)) as well as in central USA, Europe, Africa, East and Southeast Asia and East Australia (1973-2018 ?)). In contrast, negative, statistically significant trends have been found over coastal US, India and West Australia (1973-2018; ?,?, 1979–2019; ?,?) , and China (2004-2016; ?). The $ABLH$ trends over China are spatially heterogeneous. At the global scale, an average increase in $ABLH$ of ~100 m per decade has been reported (1973-2018; ?)), with maxima of ~150 m per decade but also lower values of 31 m per decade over Iran (1979-2016) or 76 m per decade over Europe (1973-2010) . Trends mostly peak in spring and at times in summer seasons.~~

~~Several studies address the $ABLH$ trend over Arctic regions. For example, ? find both positive (Alaska) and negative (Canada and Europe) statistically significant changes in $SBIH$ (1990-2009) and hence conclude no clear trend dominates the whole Arctic region. ? locate a series of statistically significant trends (1981-2010~~instruments are especially suitable for the operation in automatic measurement networks as these systems are compact, tend to have comparatively lower costs and can be operated with low maintenance under all (or most) weather conditions. Namely, these are (i) microwave radiometers (MWR) and infrared spectrometers (IRS) for the profiling of temperature (and humidity), (ii) (scanning) Doppler wind lidars (DWL) or radar wind profilers (RWP) for the ~~marine boundary layer over the Atlantic Ocean, including an increase in the height of the marine boundary layer, a thinning of the temperature inversion layer and a weakening of the atmospheric stability.~~

~~Long-term trends in clear-sky $ABLH$ have been found to be correlated to a range of indicators that can help reveal further insights on the physical processes involved. Positive (negative) correlations with surface air temperature (relative humidity) are detected most widely (?????). Increasing $ABLH$ is further in line with an increasing diurnal amplitude of air temperature (India and Central Asia Liu and Liang, 2010; ?), while negative trends in $ABLH$ have been associated with reductions in 10 m wind speeds (74 WMO stations around the world ?), reductions in soil moisture and lower tropospheric stability (China ?) as well as with a decreasing vertical temperature gradient between 950-700 hPa (74 WMO stations around the world ?).~~

6 ~~Conclusions~~

~~Despite the importance for a range of applications, quantitative knowledge on the temporal and spatial variations in atmospheric boundary layer (ABL) height is still scarce. The overarching objective of this review is to emphasize how ground-based remote sensing methods are best exploited in order to gain a detailed understanding of ABL layers, their heights and dynamics. Due to advances in algorithm development, multi-sensor analysis has the potential to better quantify the relationship between layer heights based on thermodynamic, dynamic and aerosol-based retrievals, and is therefore expected to provide valuable~~

~~new insights into ABLdynamics. Advances in measurement technology, the operation of coordinated measurement networks , and developments of automatic layer height retrievals increasingly allows for ABL climatology studies to be performed. To demonstrate the vast potential of increased ABL monitoring efforts, long-term observational studies are reviewed summarising our current understanding of ABL height variations at the global scale and specifically over land or for marine environments. Interestingly, even long-term trends start to emerge~~observation wind and turbulence profiles, and (iii) (high-SNR) automatic lidars and ceilometers (ALC) for aerosol profiling in the ABL. Further could the deployment of differential absorption lidars (DIAL) in organised sensor networks increase in the future, adding profile observations of humidity or trace gases.

~~As dense measurement networks are emerging across Europe and other parts of the world with high spatial coverage, harmonisation of operations, data processing and layer height retrievals are key. This review summarises capabilities and limitations of various measurement technologies available to capture a wide range of atmospheric profile variables within the ABL as well as the numerous methods to~~ Numerous methods are available to derive the height of the ABL and its most prominent sublayers, namely the mixing boundary layer height (MBLH) and the height of the residual layer (RLH). The MBLH represents the stable (SBLH) or convective boundary layer height (CBLH), respectively, depending on atmospheric stability conditions. In addition to layer height retrievals, methods are discussed which characterise the ABL according to ~~cloud dynamics, atmospheric stability~~ atmospheric stability and turbulence, cloud dynamics, or aerosol ~~transport regimes~~ distributions based on atmospheric profile observations.

An overview is provided on the capabilities and limitations of the large number of layer height ~~retrieval methods~~retrievals, including thermodynamic methods, wind and turbulence retrievals and those based on aerosol information. Retrievals based on temperature, turbulence or wind can take into account the atmospheric stratification of the probed layer and are hence able to specifically address either SBLH or CBLH. The height of the surface-based temperature inversion (SBIH) can further be determined from temperature profile data while a low-level jet can be diagnosed from wind observations. Aerosol-based methods again analyse the result of recent mixing processes without being able to determine whether the tracers were transported as a result of thermal buoyancy or shear-driven turbulence and are hence able to track MLBH or RLH. The latter can also be assessed based on the height of the capping inversion (CI) in a temperature profile.

For the detection of shallow layers, the near-range capabilities of a ground-based remote sensing profiler are critical ~~. MWR are very suitable given their sensitivity is maximal near the sensor, ALChave better near-range capabilities than~~ while the signal strength (in relation to the noise levels) determines the maximum range observed and data quality within a deep CBL, which can reach several kilometers (3 km, or even higher). Sodars tend to have their strength in the near-range (mostly within the first km) while high-power ~~research lidars and scanning~~ aerosol research lidars provide high-quality data in greater altitudes and are not very suitable for the assessment of conditions very close to the surface. For lidar systems, there is usually an inverse relation between limitations in the near- and far-range, respectively, because high laser power associated with quality observations in greater altitudes usually goes along with a larger blind-zone near the instrument. For DWL, RWP, ALC, or DIAL, a variety of instrument models is available, with respective range extent capabilities and noise levels. Scanning DWL can reduce the blind zone by adding shallow-angle scan strategies. ~~On average, uncertainty in SBLH detection is estimated around 30-40 %. Since turbulence in the SBL is usually not uniform, the diagnosed layer heights can differ systematically~~

~~between thermodynamic, turbulence, or aerosol-based methods. Most methods are challenged when multiple layers are present as the task of layer attribution is considered more uncertain than the simple task of layer detection. Hence at night and early morning, aerosol-based methods are at risk to confuse MBLH and RLH, especially if the composition is similar within the two layers. But also turbulence-base layer detection algorithms can be challenged by the presence of intermittent turbulence in the RL. Thermodynamic layer detection from MWR profile data is less suitable for the detection of RLH (given lower sensitivity) so that the RL poses less of the problem for the detection of MBLH at night and early morning. The RLH can be tracked most reliably using thermodynamic retrievals of the CI applied to airborne in-situ sensors or by aerosol-based methods~~MWR are very suitable for the assessment of conditions near the ground, given their sensitivity is maximal near the sensor and vertical resolution of their temperature product reduces with increasing altitude. For all systems, studies show that careful processing and detailed quality control are vital to produce high-quality profile observations. This is particularly critical for measurement uncertainties that propagate to the accuracy of ABL height retrievals.

The morning growth and evening decay of the CBL pose severe challenges to numerical simulations for a range of applications, including air quality, greenhouse gas assessment and numerical weather prediction. When using observations for model evaluation or comparisons, it is crucial to carefully consider the specific uncertainties of the respective measurement used. Also, it is important to understand which atmospheric variable is used for layer detection, as it can introduce systematic biases if e.g. turbulence-derived layer heights are compared to results exploiting aerosol profiles.

Given the different stages of CBL development during morning growth, the continuous monitoring enabled by remote sensing profilers is a clear advantage for the characterization of this period compared to balloon ascends. Profiles from sodar or certain ALC have proven very useful to capture the onset of the CBLH at very low altitudes ($<$ 100 m). Approaches based on high-frequency variations of wind or aerosol are often particularly good at tracking CBLH during morning growth. Turbulence-based results from RWP usually requires longer integration times compared to DWL or ALC that range in the order of minutes. Turbulence- and aerosol-based CBLH during morning growth can be very similar provided appropriate layer attribution is performed. However, several studies report a temporal delay of aerosol-derived CBLH morning growth both relative to temperature-derived CBLH and turbulence-derived growth, with time lags of up to two hours. No clear picture has yet emerged on a potential time lag between the growth of temperature- and turbulence-derived layer heights. Method synergy of high resolution observations (both in time and vertical dimension) is a promising means to better understand and quantify the CBL growth.

Most methods show very good performance during daytime, especially when the CBL is fully developed over the entire ABL. Provided sufficient SNR and careful data processing, CBLH from all retrieval methods can agree within a few hundred metres. A large entrainment zone (EZ) can result in a weaker delineation at the CBLH, increasing uncertainty in layer detection for all methods. As convective clouds can significantly challenge layer detection, daytime maxima of the layer estimates from temperature-, turbulence-, and aerosol-based methods are most similar in cloud-free conditions. However, also strong shear layers or elevated aerosol layers above the ABL can challenge turbulence-based and aerosol-based algorithms, respectively.

Although CBLH growth rates and maxima are strongly affected by clouds, the majority of ABLH climatology studies to date focus on clear-sky conditions. Cloud cover or even cloud type are considered very rarely. Recent developments in automatic detection algorithms that now consider cloud dynamics are expected to enable more comprehensive assessments in the future. Especially differentiating between boundary layer clouds and those decoupled from the ABL greatly aids interpretation.

The evening transition of the daytime CBL into the nocturnal boundary layer around sunset is directly monitored by turbulence profiles. While turbulence-based layer retrievals and thermodynamic methods are in general agreement at this time of day, aerosol-based MBLH results are particularly uncertain during this time ~~of evening decay as it tries~~ at they try to track the history of recent turbulence activity. Reliable data in the near range, careful processing algorithms and high surface aerosol emission rates increase the likelihood of this transition time to be captured accurately by aerosol profiling methods.

~~A list of aspects is identified that require increased attention by future studies.~~ The detection and interpretation of nocturnal layer heights is still prone to significant uncertainty. On average, uncertainty in SBLH detection is estimated around 30-40 %. Since turbulence in the SBL is usually not uniform, the diagnosed layer heights can differ systematically between thermodynamic, turbulence, or aerosol-based methods. Most methods are challenged when multiple layers are present as the task of layer attribution is considered more uncertain than the simple task of layer detection. Hence at night and early morning, aerosol-based methods are at risk to confuse MBLH and RLH, especially if the composition is similar within the two layers. Turbulence-based layer detection algorithms may be confused by the presence of intermittent turbulence in the RL. Thermodynamic layer detection from MWR profile data is less suitable for the detection of RLH (given lower sensitivity) so that the RL poses less of a problem for the detection of MBLH at night and early morning. The RLH can usually be tracked reliably using thermodynamic retrievals of the CI applied to airborne in-situ sensors or by aerosol-based methods, but also trace gas observations from DAIL or Raman humidity and temperature observations can be exploited. Further investigation is required into the impact of ABL dynamics and atmospheric stability on the relative agreement between ~~methods~~SBLH results. At night, instrument synergy between ~~microwave radiometer (MWR) temperature profiling~~ radiometer temperature profiling (IRS, MWR) and aerosol observations from ~~automatic lidars and ceilometers (ALC)~~ ALC is particularly promising given their respective strengths in observing the SBLH and RLH.

A clear potential is identified for using multiple methods or sensors (*instrument synergy*). These can advance both (i) methodologies and products and (ii) processes studies and applications.

(i) Using the synergy of multiple methods and/or sensors can advance the detection of layer heights and classification procedures:

– ~~Most striking differences in detected layer heights are found for the CBLH during morning growth when there may be a time lag between retrievals that use thermodynamic profiles, turbulence profiles or aerosol profiles, respectively. Again, detailed method synergy of high resolution observations (both in time and vertical dimension) is a promising means to gain novel insights.~~ Sensors with different range capabilities can be combined to ensure high-quality data are collected along the entire extent of the ABL.

– ~~Although CBLH growth rates andmaxima are strongly affected by clouds, the majority of ABLH climatology studies to date focus on clear-sky conditions. Cloud cover or even cloud type are considered very rarely. Recent developments in automatic detection algorithms that now consider cloud dynamics are expected enable more comprehensive assessments~~

~~in the future.Especially differentiating between boundary layer clouds and those decoupled from the ABL greatly aids interpretation~~Multiple sensor types can be combined to retrieve atmospheric variables with different physical information content that can then be used to calculate advanced synergy parameters (such as the bulk Richardson number) feeding into the layer height retrieval (combined by e.g. AI or fuzzy logic).

– Combining layer estimates from a range of methods provides a valuable basis for the assessment of layer height uncertainty, the latter still being a challenging topic due to the lack of an objective reference standard.

– ~~Incorporating~~ ABL classification schemes are emerging that not only provide layer heights but also ~~assess~~ incorporate the source (surface-driven, cloud-driven) ~~and~~ or nature (buoyancy, shear) of turbulent ~~exchange into the analysis is expected to provide novel insights into the relation between MLH and cloud dynamics.~~ exchanges or can differentiate between different aerosol types. Where multiple measurements are combined that represent different physical aspects of the ABL, such tools can be particularly powerful.

(ii) Incorporating multiple sensors is extremely valuable for process studies as well as model evaluation.

– ~~To achieve a better process understanding of~~ Naturally, utilising multiple sensors across an entire measurement network adds spatial information in the horizontal domain, which allows for variations in layer heights to be assessed and interpreted (in relation to e.g., surface forcing, local- or regional circulations, synoptic conditions) but also for transport processes to be better detected (e.g. horizontal advection within the ABL ~~dynamics in heterogeneous environments (such as the urban boundary layer or the boundary layer in complex terrain) , very dense spatial measurement networks are required representing the diversity in topographic settings, surface cover characteristics, and climatological conditions.~~ or long-range transport).

– ~~In any case, careful processing and detailed quality control are vital to produce high-quality profile observations. This is particularly critical for measurement uncertainties that propagate to~~ Where observations of multiple atmospheric variables are available simultaneously, they allow for the connections between processes to be explained (e.g. understanding the vertical distribution of aerosols as a result of turbulent mixing processes that form in response to the ~~accuracy of ABL height retrievals~~thermodynamic structure of the ABL).

– ~~Choosing the appropriate technology for a given network not only needs to consider the physical information content of the atmospheric quantity observed (temperature, humidity, wind, turbulence, aerosol, or gas)but also whether the sensitivity, resolution, and capabilities of a given sensor are appropriate to monitor the layer(s) of interest. Such instrument characteristics not only differ between sensor types but also between models from different manufacturers and can depend even on firmware, hardware, instrument settings, instrument age, or sampling strategies~~Finally, synergy of multiple sensor networks with different types of profilers that operate continuously starts to portray the four-dimensional complexity of ABL dynamics.

It can be concluded that ground-based ABL profile remote sensing is a powerful means to gain high-resolution observations of the atmospheric boundary layer – to date, the most under-sampled part of the atmosphere. The diversity in measurement technology and algorithm variety bears a challenge for the quantification of retrieval uncertainty but should also be considered an advantage for powerful process studies and synergy exploitation of the increasingly rich operational measurement networks.

Code availability. Not applicable

Data availability. Not applicable

Code and data availability. Not applicable.

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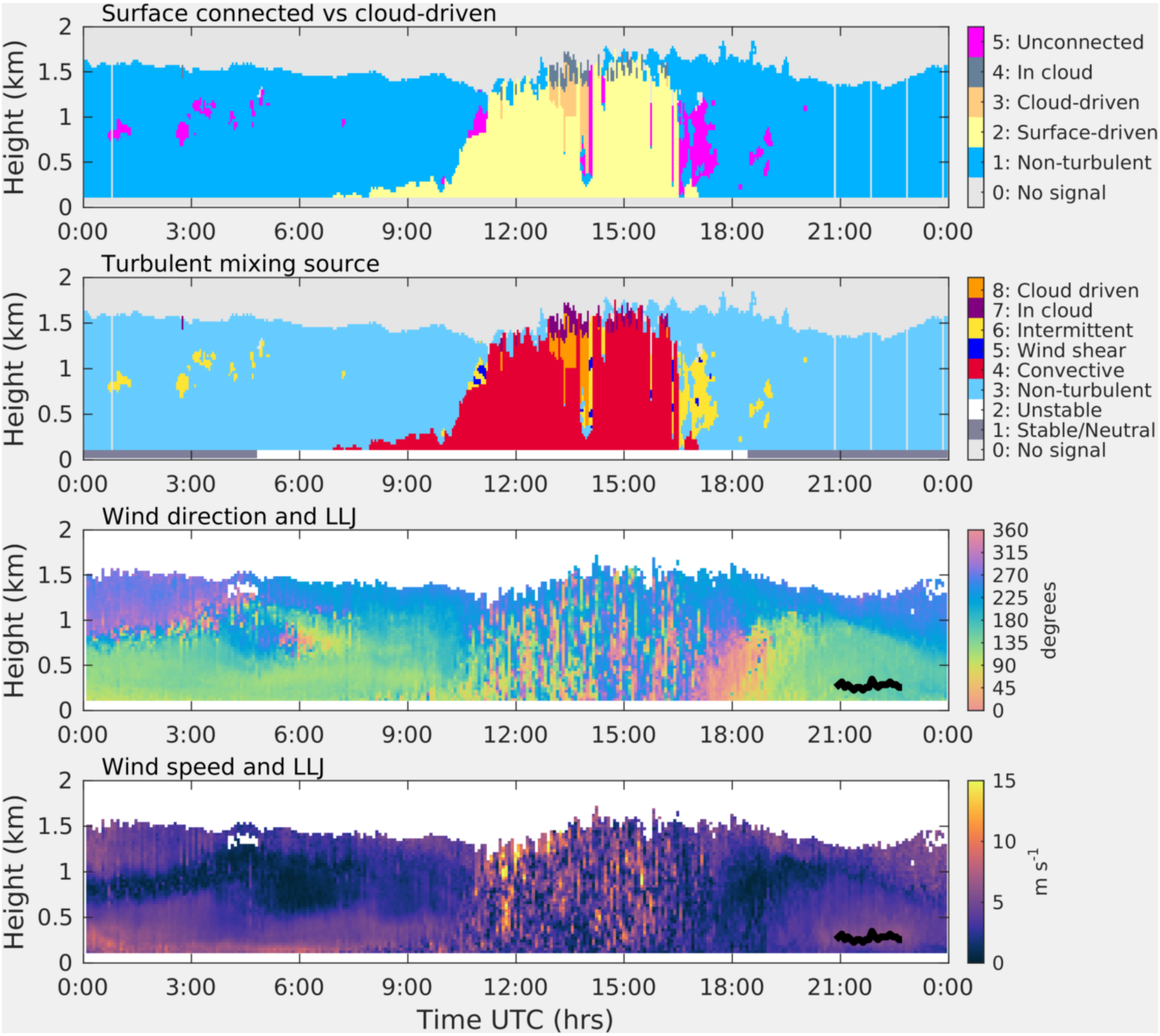
Table 2. Instrument types used to gather vertical profiles of atmospheric and measurement variables (Sect. 2.1; Table 1) in the atmospheric boundary layer. These observations are increasingly organised in national and international monitoring networks (see Sect. 2.3 for ~~definition of acronyms and~~ further details). Acronyms: ACTRIS (Aerosols, Clouds and Trace gases Research Infrastructure), ADnet (Asian Dust and aerosol lidar observation network), AMDAR (Aircraft Meteorological Data Relay), ARM (Atmospheric Radiation Measurement), EARLINET (European Aerosol Research Lidar Network), EUMETNET E-PROFILE (European Profile of the European Meteorological Network), IAGOS (In-service Aircraft for a Global Observing System), IGRA (Integrated Global Radiosonde Archive), LALINET (Latin America Lidar Network), MPLnet (NASA Micro-Pulse Lidar Network), MWRnet (Microwave Radiometer Network), NDACC (Network for the Detection of Atmospheric Composition Change), NYS Mesonet (New York State Mesonet).

Instrument type	Measurement and atmospheric variables	Network operations
~~airborne~~ Airborne in-situ meteorological sensors	$T, \theta, \theta_v, RH, u, v$ ~~IGRA, AMDAR~~, U	AMDAR, ARM, IGRA
~~airborne~~ Airborne in-situ chemistry sensors	ρ, c	IAGOS, ARM
~~microwave~~ Microwave radiometer (MWR), infrared spectrometer (IRS)	T_b, T	EUMETNET E-PROFILE, ~~ACTRIS-CCRES, MWRnet~~ ACTRIS/Cloudnet, MWRnet, NYS Mesonet, ARM
~~differential~~ Differential absorption lidar (DIAL)	~~T,~~ r, ρ, c	~~ACTRIS-CREGAR,~~ ACTRIS/NDACC
~~radio~~ Radio acoustic sounding system (RASS)	T	
Raman lidar	T, r, ρ, c, β_{att}, β_p, α_p, δ, δ_p, colour ratio	ACTRIS/EARLINET, ACTRIS/Cloudnet, ARM, NDACC
Doppler wind lidar (DWL)	v_r, u, v, w, U, $\sigma_w, \sigma_u, \sigma_v, TKE, \epsilon$	ACTRIS/Cloudnet, EUMETNET E-PROFILE, ~~ACTRIS-CCRES~~ NYS Mesonet, ARM
Radar wind profiler (RWP)	~~u, v, w~~ c_n^2, v_r, u, v, w, U, $\sigma_w, \sigma_u, \sigma_v$, TKE	EUMETNET E-PROFILE, ARM
~~sodar~~ Sodar	c_n^2, v_r, u, v, w, U, σ_w, TKE	ARM
Automatic lidars and ceilometers (ALC)	β_{att}, δ	ACTRIS/EARLINET, ACTRIS/Cloudnet, EUMETENET E-PROFILE, ~~ACTRIS-CCRES, ACTRIS-CARS~~ ARM
Aerosol lidar	β_{att}, ~~β~~ β_p, α_p, δ, δ_p ~~CARS-ACTRIS, EARLINET~~, ~~LALINET~~ colour ratio	ACTRIS/EARLINET, NDACC, LALINET, MPLnet, ADnet, ARM



~~SBI are much more frequent during nighttime and early morning, and rarely occur between noon and sunset. In polar regions, temperature inversions over ice can be very strong in winter leading to extremely stable boundary layers (?). The $SBLH$ can be as low as 20 m (?), and internal gravity waves have been observed propagating along the top of the SBL (Kouznetsov, 2009). Katabatic winds arising from drainage flows even over gentle slopes are a common feature over large ice sheets and their vertical profile usually determines $SBLH$. SBI are much more frequent at high latitudes (in polar regions: 80 % and 50 % of the time at night and at midday, respectively) than in the tropical trade-wind regions (<50 % and <15% at night and midday, respectively; Seidel et al., 2010). A clear north-south gradient of $SBIH$ is established at the continental scale for both Europe and North America with greater layer heights over warmer areas (Seidel et al., 2012; Wang and Wang, 2014).~~

4.1 ~~Atmospheric boundary layer over land~~

4.0.1 ~~Diurnal cycle~~

~~The layer most often analysed is the CBL (Sect. 1.1) as its well-mixed characteristics generally facilitate reliable detection and attribution~~