

A Raman Lidar at La Reunion (20.8°S, 55.5°E) for monitoring Water Vapor and Cirrus Distributions in the Subtropical Upper Troposphere: Preliminary Analyses and Description of a Future System

C. Hoareau^{1,6}, P. Keckhut¹, J.-L. Baray^{2,4}, L. Robert², Y. Courcoux⁴, J. Porteneuve¹, H. Vömel³, B. Morel^{2,5}

[1]{LATMOS, UMR8190, INSU-CNRS, UVSQ, UPMC, 11 Boulevard d'Alembert, 78820 Guyancourt, France}

[2]{LACy, UMR8105, 15 avenue René Cassin –BP 7151– 97715 St-Denis Cedex 09, La Réunion, France}

[3]{Deutsch Wetterdienst, Richard Assmann Observ, Lindenberg, Germany}

[4]{OSU Réunion, UMS3365, 15 avenue René Cassin –BP 7151– 97715 St-Denis Cedex 09, La Réunion, France}

[5]{now at LE2P, 15 avenue René Cassin –BP 7151– 97715 St-Denis Cedex 09, La Réunion, France}

[6]{now at LMD, UMR8539, INSU-CNRS, UPMC, Ecole Polytechnique, 91128 Palaiseau Cedex, France}

Correspondence to: C. Hoareau (christophe.hoareau@lmd.polytechnique.fr)

Abstract

A ground based Rayleigh lidar has provided continuous observations of tropospheric water vapor profiles and cirrus cloud using a preliminary Raman channels setup on an existing Rayleigh lidar above La Reunion over the period 2002-2005. With this instrument, we performed a first measurement campaign of 350 independent water vapor profiles. A statistical study of the distribution of water vapor profiles is presented and some investigations concerning the calibration are discussed. Analysis regarding the cirrus clouds is

presented and a classification has been performed showing 3 distinct classes. Based on these results, the characteristics and the design of a future lidar system to be implemented at the new Reunion Island altitude observatory (2200 meters) for long-term monitoring is presented and numerical simulations of system performance have been realized to compare both instruments.

1 Introduction

Water vapor has long been recognized as one of the most important trace gases in the atmosphere. The measurements of water vapor profiles are important for understanding and forecasting of the moisture convection and horizontal transport. Water vapor plays also a crucial role in many aspects of the Upper Troposphere and Lower Stratosphere (UTLS). It contributes strongly to the radiative balance of the atmosphere and plays an important role in global climate (Forster and Shine, 2002; Kley et al., 2000). Measuring accurately the water vapor concentration in the UTLS region is a difficult task given its very low concentration and its large variability. Water vapor can be considered for the study of tropical and sub-tropical atmospheric dynamical phenomena and their roles in the local and global circulations and in climate changes through vertical and horizontal transport in particular through tropical and sub-tropical Stratosphere Troposphere Exchanges (STE). However, condensation into ice during cirrus formation redistributes water downward through subsidence effects of heavy particles. Upper tropospheric water vapor in the tropics and subtropics is strongly influenced by the Hadley Cell and the Walker circulation (Kley et al., 2000). Widening of the tropics may also lead to changes in the distribution of climatically important trace gases in the stratosphere. The Brewer-Dobson circulation moves air upwards from the troposphere into the stratosphere in the tropics. If the area over which this upwelling occurs increases, transport of water vapor into the stratosphere might be enhanced. This could lead to an enhanced greenhouse house effect, including tropospheric warming and stratospheric cooling, and reduced ozone (Seidel et al, 2008; Forster and Shine, 2002; Kirk-Davidoff et al, 1999). Tropospheric and stratospheric water vapor has been measured over the past decades by a large number of instruments with different characteristics and limitations (Kley et al., 2000). Water vapor measurements using the Raman lidar are not new (Cooney, 1970; Ferrare et al., 1995; Whiteman et al., 1992). This technique is one of the only ones which enables to probe the upper troposphere and tropopause region. Moreover, the possibility to acquire an elastic

signal simultaneously with water vapor Raman signals is of great interest for providing information about ice crystal occurrence. Capabilities have been successively improved with larger commercial laser power availability (Sakai et al., 2007; Sherlock et al., 1999a; Leblanc et al., 2008; Whiteman et al., 2010). Some preliminary Raman channels have been setup on an existing Rayleigh lidar to perform first investigations and capabilities evaluations to measure water vapor in the upper troposphere and to design a future specific instrument. In this publication, a brief description of data retrievals and the design of the instrument have been described in Sect. 2 and 3. Data processing, regarding calibration and validation aspect, is exposed in Sect. 4. A statistical study of subtropical water vapor and scattering ratio profiles from Raman lidar is presented in Sect. 5 and the description of the design of a new specific Raman lidar to be implemented at the Maïdo altitude station is presented in Sect. 6.

2 Water vapor and cirrus cloud optical depth calculation from Raman lidar

2.1 Water vapor mixing ratio

The ratio of the mass of water vapor to the mass of dry air in a given volume, known as the water vapor mixing ratio, is a convenient measure of the amount of water vapor in the atmosphere. As atmospheric nitrogen forms a constant proportion of dry air (~78%) in the lower atmosphere, normalizing the H₂O Raman return with the N₂ Raman profile allows mixing ratio to be derived. Accounting for the atmospheric differential transmission $\Gamma(z)$ and the calibration coefficient C , it can be calculated using the following expression (Sherlock et al., 1999a; Whiteman et al., 1992):

$$q(z) = C \cdot \Gamma(z) \cdot \frac{S_{H_2O}(z)}{S_{N_2}(z)} \quad (1)$$

The calibration aspect is an important issue to insure an accurate monitoring. Various approaches have been tested to calibrate the water vapor measurements of a Raman lidar system. However, calibration issues are still pending and debated (Whiteman et al., 2011; Leblanc et al., 2011). Though an absolute calibration of the entire lidar system is theoretically possible (Vaughan et al., 1988; Sherlock et al., 1999b; Venable et al., 2011), the signal ratio is usually scaled to various external water vapor measurements (radiosonde, microwave radiometer, Global Positioning System (GPS)...) in order to deduce water vapor mixing ratio. Calibration coefficients determined from nearby radiosondes are commonly used, but their

reliability for long-term continuity is questionable (Soden and Lanzante, 1996) and independent techniques have been investigated (Sherlock et al., 1999b; Sherlock et al., 1999a). The Network for the Detection of Atmospheric Composition Change (NDACC) has recently established long-term monitoring of water vapor using Raman lidar as one of its core objectives (Leblanc et al., 2008). One of the principal needs for developing a long-term dataset for monitoring atmospheric trends is a calibration that varies randomly around some mean value and does not involve step jumps of unknown magnitude or significant drifts (Whiteman et al., 2011a). For this reason, it is essential to carefully investigate any calibration techniques developed for ensuring stable, long-term calibrations even if a greater tolerance for random uncertainty budgets in the time series for upper troposphere trend detection relaxes the need for calibration accuracy of Raman lidar, which makes the challenge of calibration perhaps easier to meet (Whiteman et al., 2011b).

2.2 Cirrus cloud optical depth retrieval

The optical thickness of cirrus is calculated in accordance with the aerosol Scattering Ratio profile (SR) which is defined as the ratio of the total (molecular and particle) backscatter coefficient divided by the molecular backscatter coefficient. Because molecular backscattering can be estimated by a dry air density profile, it can further be retrieved from the nitrogen signal, so SR can be derived from the ratio of the power in the Rayleigh-Mie and Raman vibrational N_2 channels (Ferrare et al, 2001). However, even if the Raman technique can be utilized to retrieve the extinction, in our case, the nitrogen signal backscattered is too noisy in many measurements for an accurate determination of optical depth. So we choose to calculate the optical thickness of cirrus, τ_{cirrus} , from a method similar to that described by Goldfarb et al. (2001), where τ_{cirrus} can be expressed by the following expression:

$$\tau_{cirrus} = (LR)\sigma_{rayleigh} \int_{z_{min}}^{z_{max}} n_{air}(z)(SR(z) - 1) dz \quad (2)$$

Where $\beta_{rayleigh} = \sigma_{rayleigh}n_{air}(z)$ and the air density number $n_{air}(z)$ are calculated by the Mass Spectrometer Incoherent Scatter-Extended-1990 (MSISE-90) atmospheric model and $\sigma_{rayleigh}(532nm) = 5.7 \times 10^{-32} m^2 sr^{-1}$. To be able to retrieve with the same confidence all cases including thin or thick clouds, a lidar ratio (LR) of 18.2sr (Platt and Dilley, 1984) is used. Over the year, some different values of mean LR have been retrieved; however, analysis based on Raman lidar measurements yield LR values in quite good agreement with this value

(Reichardt et al., 2002; Cadet et al., 2005). Furthermore, subdivision by cirrus generating mechanism (e.g. Sect. 5.4) seems to not influence significantly the layer mean LR consistent with the assumption in this study (Whiteman et al., 2004).

3 Technical description of the instrument

The Raman water vapor lidar system deployed at La Reunion is an upgrade of the receiving optics of the existing Rayleigh-Mie lidar system which operates on a routine basis at night, except in presence of low cloud at the Observatoire de Physique de l'Atmosphère de La Réunion (OPAR), hosted by Reunion Island university at 80m above the sea level (ASL) (Baray et al., 2006). Regular water vapor measurements have been realized with this configuration over the period 2002-2005. This long campaign allowed us to perform a preliminary study of water vapor monitoring capabilities and to evaluate the needs regarding a more specific lidar system to be implemented at the future altitude facility at Piton Maïdo mount (2200m ASL). During this period, about a hundred acquisition nights have been acquired using the Raman channels.

The Raman lidar system is based on a Nd:Yag laser source with a repetition rate of 30Hz and the second harmonic is used. The pulse energy at 532.1nm is 800mJ/pulse (9 ns pulse length). The radiation backscattered by the atmosphere is collected by optical fibers mounted on the focal plane of a 4-telescope mosaic (0.53m diameter each) of Newtonian type with a field-of-view of 1mrad and transferred to the optical ensemble. A schematic representation of the Rayleigh lidar instrument is given in Fig. 1. In this ensemble, the received wavelengths are spectrally separated through a set of dichroic beam splitters, mirrors and bandwidth pass-band interference filters (BPIF) as shown in Fig. 2. The beam reflected toward this ensemble is, firstly, filtered by a high-pass interference filter specially designed to reject the remaining 532 nm component with a rejection ratio between the transmitted and incident energy at 532 nm of 10^{-5} : this filter has a maximum transmission of 91.5% at 607 nm and of 89.2% at 660 nm. The filtered beam is then split by another dichroic beam splitter that reflects its 607 nm component toward the photomultiplier (PMT) of the Raman N₂ channel and transmits its 660 nm component toward a metallic mirror that finally reflects the 660 nm component toward an avalanche photodiode (APD) which is the detector of the Raman H₂O channel. A BPIF with a maximum transmission of 57% at 607 nm and a full width at half maximum (FWHM) of 1nm is placed in front of the N₂ PMT and aims notably to reject the residual 532 nm component

1 which passed through the high-pass interference filter. A focusing lens with a focal length of
2 50 mm is placed after this filter to focus the Raman 607 nm component onto the photocathode
3 of the N₂ PMT. Two three-cavity interference filters are used on the H₂O channel to ensure
4 that the Rayleigh-Mie contribution and Raman contributions from N₂ and O₂ are less than
5 0.1% of the water vapor signal in all measurement conditions. The BPIFs placed in front of
6 the H₂O photodiode have a maximum transmission of 80% and 57% at 660nm and
7 respectively a FWHM of 5nm and 1nm. The detector used for the Raman N₂ channel is a
8 multi-alkali 9-stages Hamamatsu R1477 circular-cage PMT that is cooled down to a
9 temperature of about -20°C through a Hamamatsu Peltier cooling mechanism. When cooled,
10 this PMT has a low dark current (<5 counts per second). PMT cooling permits here to reduce
11 the background noise (sky background and detector noise) by a factor ~2 compared to the
12 background noise obtained by the use of PMT with a typical dark count rate of ~100cps.s⁻¹.
13 Although the skylight background is much larger than detector noise for daytime
14 measurement, reduction of both noise sources for system operating at night must be
15 considered. This PMT has a quantum efficiency of about 12% at 607nm and its typical
16 voltage operation is less than 900V to limit the signal-induced bias associated with the PMT
17 response to an intense luminous pulse (Acharya et al., 2004). PMT pulses are typically 8ns in
18 duration and the maximum count rate for a Poisson signal with exponentially distributed
19 inter-arrival times is 45MHz. An avalanche photodiode has been used for the Raman H₂O
20 channel because of the better quantum detection efficiencies of these detectors at visible
21 wavelengths compared to classical PMTs. In addition, the technology of photodiodes
22 generally offers better overall detection efficiencies in the red and near-infrared parts of the
23 spectrum. The APD model used for the H₂O channel is an EG&G SPCM AQ-232 photodiode.
24 This photodiode is uncooled but has a very low temperature dependence of its characteristics.
25 Its quantum efficiency is about 35% at 660nm. Contrary to PMTs, it is not possible to set the
26 supply voltage of the photodiode and it must be noticed that the output of this detector is a
27 standardized logic pulse: a 2V logic pulse with a temporal length of about 40ns correspond to
28 each detected photon. The main drawback of this photodiode is its detection surface which is
29 a disc with a diameter of 480μm: this entails an important sensitivity of the H₂O channel to
30 optical misalignment and also do not ease the alignments of the photodiode. It is essential to
31 use very-low-noise detectors for detecting the very weak water-vapor returns. The maximum
32 count rate for a Poisson signal with exponentially distributed inter-arrival times is 10 MHz.

4 Description of data processing

4.1 Methodology and validation

Raman lidar profiles of water vapor mixing ratio are determined by taking the ratio of Raman backscatter by water vapor to Raman backscatter by one of the well mixed gases such as nitrogen (e.g. Sect 2.1). In the middle and upper troposphere, aerosols densities are generally small and ice clouds do not exhibit large wavelength attenuation dependence. Though it can be estimated with additional channels (Faduilhe et al., 2005), it has been proved that the relative transmission of the Raman returns, at 607nm (N₂) and 660nm (H₂O), corresponds to a 0-5% overestimation in extreme aerosol loading conditions. This systematic bias introduced in water vapor profiles has been corrected from molecular attenuation and absorption by ozone coefficients of the US Standard Atmosphere. Due to the large bandwidth of the interferential filter (1nm), no temperature dependence corrections have been applied. The water vapor mixing ratio profiles are obtained by averaging pre-accumulated lidar signals (typically 2 min) over an a-priori period of quasi-stationary conditions regarding statistical variability. Assuming stationary atmospheric conditions, the backscattered photons hit the counter independently and the counting follows a Poisson process. Thus, from a statistical point of view, the mixing ratio derived from the signal averaging is equivalent to the averaging of the individual mixing ratio profiles. Because the photon-counting process is described by a Poisson statistics, long enough sampling period provide a better statistical estimator of the water vapor mixing ratio. However, if the sampling period is too long, information about the variability is lost. To achieve a reasonable compromise between accuracy and atmospheric variability, the applied method consists of adjusting the integration time with the discontinuity of the flow sounded. The identification of discontinuities in the time series is based on the test of non-stationarity of the series due to a change in dispersion in the H₂O/N₂ ratio time series as described by Hoareau et al. (2009). Under the assumption that two successive profiles give a variability of the same order, they can be considered as independent measurements. In clear sky condition, this Raman lidar system allows to perform measurements of water vapor mixing ratio profiles that extend up to the tropopause region in nighttime conditions.

In 2005-2006, a campaign of CFH (Cryogenic Frost point Hygrometer) measurements was organized at Reunion Island. The CFH is sensor carried under a balloon which measures

water vapor continuously between the surface and the lower stratosphere (Vömel et al., 2007). It is based on the chilled mirror principle and measures the temperature of a mirror carrying a thin dew or frost layer, which is maintained in equilibrium with the ambient water vapor. The optical phase sensitive detector measures the bulk reflectivity of the mirror and the microprocessor feedback controller regulates the mirror temperature such that the bulk reflectivity and hence the condensate layer remain constant. Under this condition the condensate layer on the mirror is in thermal equilibrium with the vapor phase of the air passing over the mirror. The mirror temperature is then equal to the ambient dew point or frost point temperature and the water vapor mixing ratio and relative humidity can be calculated from this observation using a variation of the Clausius Clapeyron equation. Like many chilled mirror instruments, CFH does not need to be calibrated for water vapor and can be considered as an absolute reference for water vapor measurements. The total uncertainty in frost point is better than 0.5K throughout the entire profile, which means a mixing ratio uncertainty of about 4% in the lower tropical troposphere and about 10% in the middle stratosphere and tropical tropopause (Vömel et al, 2007). In Fig. 3, an intercomparison between H₂O Raman lidar, (CFH) sonde and the European Center for Medium-Range Weather Forecast (ECMWF) operational is shown. Calibration has been realized here using ECMWF Operational data (e.g. Sect. 4.2). The water vapor mixing ratio profile obtained during the descent of the CFH sonde is used here to validate the lidar profile in the upper troposphere. The integration time of the water vapor mixing ratio profile is ~50min, and the random error of the profile is inferior to 1% up to ~5km, reached 10% from ~12km and superior to 30% from ~14km. Regarding the vertical integration of the data, which reduce the statistical noise and extends the altitude range in the upper troposphere, it consist to an averaging window increasing with altitude. In the lower troposphere, since the backscattered signal is large, the initial resolution (150m) is not degraded. In the middle and upper troposphere, the vertical resolution increases up to 2km.

4.2 Calibration

One of the most important issues concerning the water vapor monitoring from Raman lidar technique is the calibration of the instrument (e.g. Sect. 2.1). Even if some procedures of independent calibration as well as calibration using H₂O vertical total column have been explored (Sherlock et al., 1999b; Leblanc et al., 2008; Hoareau et al., 2009), the most commonly approaches consist in the normalization by radiosonde measurements. The lack of

simultaneous and collocated radiosonde measurements over the site during the period 2002-2005 does not permit to use them in this first investigation. Consequently, the calibration procedure has been performed using the ECMWF operational water vapor profiles. These data from the archive are re-sampled on a latitude-longitude resolution grid of 1.125° . Until August 2002, data used are provided on a vertical resolution of 21 pressure levels, after this date, the resolution is defined on 23 levels. To compare and calibrate lidar profiles with ECMWF profiles, relative humidity data provided by ECMWF were converted to water vapor mixing ratio by means of the empirical saturation vapor pressure over liquid water formulas of Hyland and Wexler (1983). The calibration coefficient is determined by adjusting the water vapor lidar profile to the ECMWF one by the median of both profiles ratio at pressure levels of ECMWF data. Although this calibration method does not appear as the most appropriate method for the long-term commitment, it allows to get a first assessment of the water vapor mixing ratio profiles data set above La Reunion. The results indicate that the calibration coefficient values seem to be quite stable, with an observed mean variance of $\sim 13\%$. When a major instrumental change occurs, the mean variance observed in the jump of the coefficient medians can be very important and change by a factor more than 10, as shown in Fig. 4.

5 Preliminary Data Investigations

5.1 Dataset

During 2002-2005 period, the water vapor Raman lidar system recorded about one hundred measurement nights. The H_2O Raman channel being an additional channel of the existing Rayleigh-Mie lidar system, simultaneous measurements of water vapor and cirrus clouds have been realized. Based on the methodology regarding the integration time period described in section 4.1, around 350 independent water vapor profiles have been obtained over this period with an average of 24 independent water vapor profiles per month. However, as seen in Fig. 5, there are more profiles from February to March and October to November. The reason for this is the duration of measurement acquisitions that were longer during these periods. This allows to derive more profiles in a same night given the high temporal and spatial variability of water vapor. On average, according to the water vapor variability, the integration time of the water vapor mixing ratio profiles is around 34 min (Fig. 6).

5.2 Water vapor seasonal cycle

The climatic context of the lidar station at La Reunion is typically that of an oceanic site dominated by the southern Hadley cell circulation (Baldy et al., 1996). Two typical seasons can be identified depending of the position of the Inter Tropical Convergence Zone (ITCZ) with respect to the island location. During the austral winter (May to October), the Inter Tropical Convergence Zone (ITCZ) position is distant from the island, and a strong influence of the Hadley and Walker cells generates steady easterly trade winds at low altitudes (<2km) and westerly winds above the trade wind inversion (Taupin et al., 1999). During austral summer (November to April), the ITCZ comes within reach of the island and thus the trade wind influence is weaker. The trade wind inversion is almost entirely disappeared during the summer. Water vapor from the marine boundary layer can also be vented to the upper troposphere by intensive deep convection. Relating to these seasons and to these dynamical considerations, an analysis of the water vapor profiles is compared for the moist and dry seasons (Fig. 7). A mean ratio between both seasons about 1.6 up to 9km is observed, above this altitude the ratio is in mean inferior to 0.4 until 15km. From 9km up to 15km, the water vapor contents have a similar decrease for both seasons with mean values between 0.25g/kg at 9km and 0.02g/kg at 15km.

5.3 Diurnal cycle

Based on water vapor distribution study at several altitudes, results have shown a systematic bimodal distribution in logscale of water vapor in lower layer of the troposphere at all altitudes up to ~4km regarding mean distribution using all data available. However this bimodality in the distribution does not occur for each measurement. From one night to another, the distribution of water vapor mixing ratio can remain in one of the two modes or move from one mode toward another (Fig. 8). This variability seems in connection with the dynamical context of the island which is located in the influence of very regular east-south-easterly trade-wind. A wind inversion resulting from the descending branch of Hadley cell circulation is the main characteristic of the wind vertical distribution over Reunion: low-level easterlies are opposite to upper-level westerlies. The transition layer which delimits low-level trade-wind regime below and the upper westerly flow regime above, is known as the trade wind inversion (TWI) which is located between 2 and 5km (Baray et al., 1998; Taupin et al., 1999). However, more investigations need to be performed for a better interpretation of these

1 results. Steep topography, dynamical influence on the synoptic trade-wind flow and various
2 diurnal thermal effects make complex this analysis.

3 **5.4 Cirrus clouds analysis**

4 In the upper troposphere, a fraction of the water condenses to generate cirrus clouds. Cirrus
5 clouds are a main uncertainty in climate change assessments (Houghton et al., 2001). They
6 have been identified as one important regulator of the radiation balance of the earth-
7 atmosphere system (Twomey et al., 1991). It is important to investigate the altitude range and
8 vertical extension of cirrus clouds, which are critical parameters for the radiative balance of
9 the atmosphere. A cirrus cloud at high altitudes and, hence, a cold cloud, influences more
10 strongly the infrared flux than the same cirrus at lower altitudes. In contrast, a cirrus cloud at
11 low altitudes has a weaker effect (cirrus cloud reflecting back to space the incoming solar
12 radiation). Currently, the vertical transport of water vapor and ice particles in the vicinity of
13 the tropopause is not perfectly known. The processes involved are debated (Pommereau et al.,
14 2011; Kiemle et al., 2008) and different formation processes could lead to different cloud
15 characteristics that require to be identified before specific statistical analysis (Keckhut et al.,
16 2006). A first climatology of sub-tropical cirrus clouds from Reunion Island lidar dataset for
17 the period 1996-2001 has been already published (Cadet et al., 2003). The updated analysis
18 realized here for the period 2002-2005 is in good agreement with these previous results
19 regarding the cirrus clouds optical depth distribution (Fig. 9). In both cases, the percentages
20 have to be associated with the value of lidar ratio equal to 18.2 sr and subvisible cirrus (SVC)
21 are defined with an optical depth $\tau \leq 0.03$ (Sassen et al., 1989). However, Cadet et al. (2003)
22 indicated that the cirrus clouds occurred 7% of the time for the total observation period versus
23 15% for the updated analysis. Cirrus occurrence frequency is obtained as the ratio of cirrus
24 detection time versus the total measurement time.

25 Moreover, complementary analyses have been performed regarding cirrus cloud classification
26 according to geometrical macrophysic properties and some investigations regarding origin
27 have been regarded. The optical thickness of cirrus cloud is calculated in accordance with the
28 scattering ratio profile (SR) using a method as described in Sect. 2.2. In order to identify
29 different cirrus cloud classes, a probability distribution study of diverse parameters of cirrus
30 clouds (optical thickness, top, mean altitude and geometric depth of cirrus cloud) is realized.
31 Results regarding the probability density functions (PDFs) show no single mode Gaussian

distribution which suggests possible different types (Fig. 10). To discriminate the different classes, a cluster analysis is performed. The approach used is similar to the multivariate analysis performed by Keckhut et al. (2006) to discriminate different cirrus classes. It consists here in applying a Hierarchical Ascendant Classification (HAC) analysis on geometrical macrophysic cirrus clouds parameters (optical depth, top, mean altitude and geometric depth of cirrus clouds) derived from lidar measurements. Principle of the HAC consist to organize the observations, defined by a certain number of variables, in aggregating them hierarchically. This method assumes that dissimilarity measure between the observations is existing and distance can be used as dissimilarity measure. In this study, dissimilarity is calculated using euclidean distance and the Ward's method, described in Ward (1963), is used for the agglomerative clustering method. Based on this analysis, HAC lead to three distinct classes as shown in Fig. 11. The mean and standard deviation for all parameters of each cirrus class are listed in Table 1. To ensure the robustness of these results, a discriminant factor analysis (DFA) is performed. DFA permits the identification of the optimal set of orthogonal projection axes which best separate the classes; these axes are the discriminant factor. As three classes have been previously identified, the analysis is done according two discriminants axes, F1 and F2. Results show a better discrimination of the different classes regarding the discriminant factor F1 which represents 93.31% of discrimination (Fig. 11). Other results indicate 99% of correlation between the discriminant factor F1 and the top of cirrus clouds which seems to be the most important parameter for the discrimination of the classes. DFA corroborates to the HAC results at 98.85% with only one different affection on 87 observations. A complementary analysis regarding the origin of the different classes is realized. Images from geostationary METEOSAT satellite are used in this analysis. During the 2002-2005 period, EUMETSAT (Europe's Meteorological Satellite Organization) was operating the Meteosat 5 satellite providing centered observations over the Indian Ocean (repositioned at 63°E in 1998 for the Indian Ocean Experiment: INDOEX). Images used here are from infrared channel in spectral range 10.5-12.5 μ m with resolution at nadir of 5km and observations taken every 30 minutes. From the images, which have been linked with lidar observations, three types of atmospheric motions have been identified and appear to be associated with the different classes : middle latitude front, tropical convection and tropical cyclone. Among the different cirrus clouds classes obtained, the first one (Class I) is related to a middle latitude front with a top of cirrus cloud located at 11.3km in mean and an occurrence of 44%. The Class II and III are related to the tropical cyclone and tropical convection

1 respectively (Fig. 12). For the Class II which is associated to the tropical cyclone, the top of
2 cirrus clouds is located at 15.9km and represents less of 20% of occurrence. The mean
3 geometrical thickness of the cirrus clouds belonging to this class is 3km. The Class III
4 indicates a mean altitude of 14.3km for the top of cirrus clouds with thickness of cirrus
5 around twice lower than the class II, 1.4km, and an occurrence of 37%.

6 **6 Future lidar system**

8 The results presented in the Sect. 5 demonstrate the capabilities of water vapor monitoring in
9 upper troposphere from lidar instrument over a subtropical site. The challenge is now to
10 monitor water vapor with a lidar system able to measure in the upper troposphere with a
11 smaller random error and to reach the lower stratosphere. The future Raman lidar which will
12 be implemented in Piton Maïdo facility at Reunion Island at the end of 2011 is designed to
13 reach UTLS. The station is located above the boundary layer, in the less cloudy part of the
14 island at an altitude of 2200m ASL. The future instrument, under construction, is principally
15 dedicated for the water vapor measurements in the lower stratosphere and upper troposphere
16 but also for the measurements of stratospheric temperature using Rayleigh scattering. It will
17 detect light backscattered by molecules and particles in the atmosphere from outgoing laser
18 beam at 355nm. Inelastic Raman backscatter from nitrogen will be detected at 387nm. It will
19 allow the direct retrieval of the aerosol extinction coefficient α and will be used to retrieve the
20 water vapor mixing ratio using the channels at 407nm which will detect vibrational Raman
21 scattering from H₂O molecules.

22 **6.1 Choice of the excitation wavelength**

23 The emitted wavelength was chosen to improve the overall efficiency and is based on several
24 factors: water vapor backscattering cross-section, laser source availability and power, detector
25 efficiency. Molecular scattering follows a λ^{-4} law, therefore short wavelengths are more
26 efficient and the near ultraviolet (UV) band is the most indicated. UV bands allow better
27 detector efficiency than in the visible and near infrared bands. The light source of this lidar
28 consists in two commercial Quanta Ray Pro-290-30 Nd:Yag lasers with frequency tripling,
29 generating laser pulses with about 375mJ at 355nm with a repetition rate of 30Hz and a
30 duration pulse of 9ns. Pulses of both lasers can be synchronized and output beams can be
31 coupled through polarization cubes to emit 750mJ pulses at 355nm. This flexibility will

enable a power increase, if needed, to reach UTLS. The lidar design was built with the opportunity to use both emitting wavelength. While capabilities are expected to be similar in theory, the design will allow direct comparisons including calibration and operational issues. In the laser coupling and optics design, we also thought about using both lasers emitting at 532nm for potential intercomparison and if for operational reasons it is decided to change.

6.2 Optical ensemble

Regarding the optical ensemble, the radiation backscattered by the atmosphere is collected by a 1.2m – diameter telescope that was previously used at Biscarrosse for Rayleigh temperature and Raman measurements (Hauchecorne et al., 1991; Keckhut et al., 1990). A narrow field of view of 0.25mrad will be used to reduce as little as possible sky background and detector noise. Contrary to the lidar system used during the period 2002-2005, the future one will not use optical fibers in the focal plane of the telescope to transfer the backscattered signals to the optical ensemble. This will avoid a systematic bias in water vapor measurements due to fluorescence contribution in fiber-optic cables, even if this transfer protocol permits to obtain quasi-constant illumination conditions at the optical fiber output and that even telescope alignment changes. This is not the case in conventional systems (without optical fiber) where optical alignment change can lead to important variations (~2-5%) in response system due to detector spatial inhomogeneities, and thus the calibration coefficient (Whiteman et al., 1992; Nedeljkovic et al., 1993). However, even if OH-rich fibers were used for the preliminary system, resulting in a reduction in the estimated fluorescence contribution, the bias introduced by the fiber fluorescence can be significant to be corrected for on a routine basis (Sherlock et al., 1999a). Consequently, we designed a configuration with a direct optical path between the secondary mirror of the receiver and the detection box although fluorescence is not limited to the fiber-optics and could arise in any optical component. Careful tests will be required when lidar will be implemented.

6.3 Rayleigh-Mie and Raman signals separation

The backscattered signals collected by the telescope are, firstly, transferred toward an optical ensemble through a set of lens and mirrors. A spectral separation of the light is then realized through a set of dichroic beam splitters as well as BPIF. A schematic view of the optical ensemble is shown in Fig. 13. The beam reflected is split by dichroic beam splitter that reflects 532nm component toward an auxiliary optical ensemble. The filtered beam is then

split by another dichroic beam splitter that reflects its 355nm component toward the photomultiplier of the Rayleigh-Mie channel coupled with a BPIF with a maximum transmission of 55.3%. The transmitted beam is filtered by a high-pass interference filter designed to reject the remaining at 355nm. This filter has a maximum transmission of 90% and 85% at respectively 407nm and 387nm. Another dichroic beam splitter is then used to reflect the 387nm component toward the photomultiplier of the N₂ Raman channel and transmits its 407nm component toward the photomultiplier of the H₂O Raman channel. A BPIF with a maximum transmission of 63.7% at 387nm and a FWHM of 3nm is placed in the front of the N₂ PMT. A high-pass interference filter designed to reject the remaining 387nm component and a BPIF are placed between the last dichroic beam filter and the lens which focalized the beam onto the photocathode of the H₂O PMT. The BPIF has a maximum transmission of 60.6% at 407nm and a FWHM of 1nm. In the case we decided to work in visible wavelengths in the future, the optical subsystem containing the splitters, dichroic mirrors and detectors can be upgraded with a second ensemble adapted to the useful wavelengths.

6.4 Photo detection

In theory, the limiting factor for a PMT in photon counting mode is the dark current. This current that contributes to the noise in the measurements can be used to categorize the limit of detection of a PMT. It can be expressed in term of Equivalent Noise Input (ENI) that is an indication of the photon-limited signal-to-noise ratio. It refers to the amount of light to produce a signal-to-noise of unity in the output of PMT and can be expressed as follows:

$$ENI = \frac{\sqrt{2 \cdot q \cdot ldb \cdot G \cdot \Delta f}}{S} \quad (3)$$

Where q is the electronic charge, ldb is the anode dark current (A), G is the gain, Δf is the bandwidth of the system (Hz) and S is the anode radiant sensitivity at the wavelength of peak response (A.W⁻¹).

Relating to the photon detector, we will use new Hamamatsu R7400- 03g or 20g mini-photomultiplier tubes depending of the wavelengths we will emit; R7400-03g for an emitted wavelength at 355nm and the R7400-20g at 532nm. Although an emitted wavelength at 355nm is most indicated and will be principally used, we will dispose of an auxiliary optical ensemble which can be used simultaneously for received wavelength at 607nm and 660nm

1 since a simultaneous emission at 532nm and 355nm can be performed (Sect. 6.1). The R7400-
2 03g has a typical spectral response in UV-Visible with a cathode radiant sensitivity of
3 $\sim 60\text{mA/W}$ for the wavelength 387nm and 407nm. Regarding the R7400-20g, the spectral
4 response is located in the Visible-Near IR range with a cathode radiant sensitivity of
5 $\sim 75\text{mA/W}$ for the wavelength 607nm and 660nm. Although these tubes are less sensitive than
6 old cooled R1477, smaller ENI have been indicated for Hamamatsu R7400-03 compared to
7 Hamamatsu R1477 at wavelengths 355nm, 387nm and 407nm, with respectively a decrease of
8 $\sim 40\%$, $\sim 30\%$ and $\sim 20\%$ of the ENI values. However, we will need check which PMT give
9 better performance with different emitted energy per pulse. Regarding data acquisition for
10 Raman channels, we will use TR 20-160 transient recorders for both analog (lower altitude)
11 and photon-counting (upper altitude) combination increasing the dynamical range of acquired
12 signal compared to conventional system.

13 **6.5 Calibration**

14 Regarding the future calibration of the instrument, coaxial geometry for emission and
15 reception was chosen, as shown in Fig. 14; indeed this design will avoid parallax effect,
16 extend measurement down to the ground and contribute to ease the alignment. It will allow to
17 perform, in better conditions, calibration using H_2O total column measurements from
18 collocated ancillary instruments (GPS, radiometer...). Radiosonde data, which could be
19 realized on daily basis at the meteorological station (20km faraway), could be used for
20 comparison on the full height range. These measurements will be used as a fully independent
21 and systematic duplicate measurement for comparison with water vapor lidar profile
22 calibrated from H_2O total column measurements. Indeed, the distance between lidar and
23 radiosonde sites could be not very effective for that concerns the calibration. However, to
24 compare one calibration method to another, some methodology to calibrate through
25 noncollocated radiosoundings (with similar range of noncollocation) can be used if need
26 (Dionisi et al., 2009). In the mechanical design, we defined and built an integrated and
27 removable support for a calibration lamp to complement the calibration with total H_2O
28 measurements to use the hybrid technique. Although instrumental changes can be detected
29 following calibration method using radiosondes, those using calibration from lamp or passive
30 zenith daytime observations are better. It is therefore important to implement one of these
31 methods for the instrument monitoring, which is necessary for long term monitoring.

6.6 Numerical evaluation of lidar signals

Numerical simulations of lidar signals can be performed for studying the expected system capability. The range interval for which a lidar can be operational is limited by the geometrical form factor function and saturation in the lower altitude and the fast decreasing signal-to-noise ratio (SNR) at far altitude range. For a first performance simulation, we suppose the R7400-03 Hamamatsu PMT with a typical dark count rate of 80s^{-1} to evaluate the detector noise contribution compared to sky background in measurements. The quantum efficiency of the R7400-03 PMT is around 22% for 407nm wavelength. Based on these component properties and H₂O channel efficiency (Table 2), simple numerical simulation of the system performance has been performed and compared to the previous system which used SPCM AQ-232 APD for H₂O channel.

For measurements at a given emitted wavelength λ_0 , the lidar signal $N(\lambda_H, z)$ in photon counts, corresponding to the range z , can be calculated by the following equation:

$$N(\lambda_H, z) = \frac{K}{z^2} \cdot n_H(z) \cdot \frac{d\sigma_H(\pi)}{d\Omega} \cdot e^{-[\int_0^z \alpha(\lambda_0, z') dz' + \int_0^z \alpha(\lambda_H, z') dz']} + N_0 + N_D \quad (4)$$

Where K is a proportionality constant for H₂O channel that accounts for the system optical efficiency, the telescope receiver area, the photomultiplier tube (PMT) spectral efficiency, the overlap function and the laser output energy; $d\sigma_H(\pi)/d\Omega$ is the Raman backscattering cross section for H₂O molecule; $n_H(z)$ is the H₂O number density; the exponential factor gives the two-way atmospheric transmission, where α is the total extinction coefficient; N_0 is the sky background and N_D represents the detector dark counts.

The background term can be evaluated as follows:

$$N_0 = \eta_0 \cdot \eta_q \cdot \frac{\lambda_H}{h \cdot c} \cdot W \cdot A \cdot f \cdot \Delta t \cdot \Delta \lambda_H \cdot \frac{2\Delta R}{c} \cdot \Omega \quad (5)$$

Where W is the background radiance ($\text{W} \cdot \text{sr}^{-1} \cdot \text{nm}^{-1} \cdot \text{m}^{-2}$), $\Delta \lambda_H$ is the receiver bandwidth (nm), and Ω is the observation solid angle (sr), η_0 the efficiency of the received channel, η_q the quantum efficiency, Δt the integration time (s), ΔR the vertical resolution (m), h the Planck constant and c the speed of light. And the dark count rate, expressed in function of d (s^{-1}), can be estimated as follows:

$$N_D = \frac{2f}{c} \cdot d \cdot \Delta R \cdot \Delta t \quad (6)$$

To perform the numerical simulations, water vapor mixing ratio profiles from ECMWF ERA-40 re-analysis is used as reference profile, and the atmospheric profile has been derived taking account extinction due to Rayleigh scattering. Rayleigh scattering profiles have been computed using International Standard Atmosphere. Absorption by ozone and other trace gases is negligible at the wavelengths used and the extinction due to aerosol do not have been considered due to altitude station (2200m). Due to the large bandwidth of the interferential filter (1nm), no temperature dependence corrections have performed. Based on measurements to the 15th March 2005 for the current system, the detector noise (N_D) to sky background (N_0) ratio has been shown of around 2. Indeed, based on the lidar measurements for this date which have been performed in clear sky and moonless conditions, the background noise indicated a value of ~ 20.5 photons for 30 minutes time integration as shown in Fig. 15. From this observation and based on APD dark count tests conducted in the same weather conditions, with values found close to 250s^{-1} for the APD dark counts, the derived value is ~ 13.5 and 6.9 photons for the detector dark counts and sky background respectively according to the Eq. 5 and 6. The derived night sky radiance of $4.6 \cdot 10^{-9} \text{ W.m}^2.\text{sr}^{-1}.\text{nm}^{-1}$ is in agreement to the night sky radiance range values indicated in literature at return wavelength and similar weather conditions (Höhn and Büchtemann, 1973; Turnrose, 1974). Höhn and Büchtemann (1973) measured the sky brightness from 400 to 800 nm in a range of weather conditions and under various aspects both experimentally and theoretically. We chose their spectrum to estimate night sky radiance and perform the numerical simulations for the future lidar system (Table 3). Usually to reduce the statistical noise, in addition to the temporal integration and to extend the altitude range in the upper troposphere, vertical integration is applied. In the numerical simulations, only temporal integration has been regarded according to mean time integration (~ 30 min) which corresponds to the mean time integration to access water vapor variability, as described in Sect.5.1.

Concerning the water vapor signal strength, simulations indicate a ratio of ~ 150 at lower altitude ($\sim 3\text{km}$) with a decrease to ~ 20 around the tropopause ($\sim 16\text{km}$) between both lidar systems for Raman H_2O channel for one laser in operation at 355nm (Fig. 15). This ratio decrease is essentially due to the altitude squared dependence in return signal which is not the same due to the altitude difference between both stations ($\sim 2100\text{m}$). Although performances improvement are the result of the emission wavelength which implies a greater backscattering of molecules at return wavelength, the receiver area and detection efficiency, the altitude of the station permits to improve the backscattered power principally in lower altitude since the

factor is ~ 11 at $\sim 3\text{km}$ and decreases to ~ 1.3 at $\sim 16\text{km}$. In this development stage, the altitude expected within an error of 15% for H_2O measurements, regarding random error (principally sky background and photon counting error) and a pre-accumulated signal of 30 minutes, is of $\sim 15.6\text{km}$ in clear sky and moonless condition and $\sim 15.5\text{km}$ with thinly covered sky and moonlit. For the same measuring conditions, random error reaches 30%, respectively at around $\sim 18\text{km}$ and $\sim 17.6\text{km}$. Coupling both lasers, the altitude expected within an error of 15% and 30% are 16.8km and 21km respectively for clear sky and moonless condition.

Concerning the actual system, simulation results have been compared to the experimental measurements. Comparison indicated a ratio of ~ 4 for H_2O channel between the expected results from simulation and those from the instrument. And regarding the numerical simulation, the expected altitude for a random error at 15% and 30% is respectively around 11.8km and 13km in clear sky and moonless conditions. In comparison, assuming the same factor of 4 between both experimental and simulation results and that seems better correspond with real-world expectations, we could expect reach an altitude of 14.6km and 16.3km within 15% and 30% respectively and good weather conditions.

7 Conclusions

We have presented preliminary data of water in upper troposphere from lidar instrument over a subtropical site. Performing analysis of the preliminary system lidar dataset, advantages and drawbacks have been pointed out. Present configuration of the lidar system permits to cover a large altitude range, from the ground up to the upper troposphere ($\sim 14\text{--}15\text{km}$) within a random error of 30% for a temporal integration of $\sim 50\text{min}$, vertical integration up to $\sim 2\text{km}$, during nighttime and presence of any clouds. Regarding statistical analysis, some results based around 350 independent water vapor profiles have been presented. Mean vertical profiles of water vapor in respect to the seasons have shown a mean ratio of 1.6 up to 9km . This ratio decreases within 0.4 in mean between both seasons above 9km up to 15km . About cirrus cloud analysis, according to different parameters of cirrus clouds, 3 distinct classes have been obtained. Following to these different classes and complementary analysis based on Meteosat images, origin of the classes have been identified and related to the tropical convection (class III), tropical cyclone (class II) and midlatitude front (class I); with respectively 37%, 19% and 44% of occurrence. The most discriminatory parameter associated to the classification seems to be the top of the cirrus cloud. These altitudes have been

demonstrated at 11.3km for the first class I, 15.9km for the second, and 14.3km for the third. Although these analyses have shown good results regarding water vapor and cirrus data, some aspects needed to be resolved in order to improve the measurements from the future lidar system. Based on these analyses and encountered difficulties to derive water vapor mixing ratio profiles, the characteristics and the future design of the instrument deployed to the Maïdo station has been presented. According to a simple numerical simulation of the system performance the results have shown a ratio of 4 for H₂O channel between theoretical curves and instrumental measurements with one laser in operation. The maximum altitude expected, taking account to a random error within 15% and pre-accumulated signal of 30min, has been indicated at 11.8km for H₂O channel in clear sky and moonless condition. Based on characteristics and design of the future lidar system, and regarding the numerical simulation of both systems, the performances seem to be improved by a factor 20 for H₂O channel around 16km. The improvements of these performances are principally due to a better detection efficiency of the optical ensemble and backscattering cross-section at shorter wavelength, the size of the telescope and the altitude of the new observatory (2200m ASL) which permit to improve the backscatter power of a ratio ~150 at ~3km. Concerning the design of the future instrument, some modifications have been chosen compared to the preliminary system in order to avoid systematic bias in water measurements removing the optical fiber and to facilitate the calibration choosing coaxial emission. Consequently, with an expected altitude between 14.6km and 16.8km for H₂O channel, within a random error of 15% and a temporal integration of 30min, and regarding the characteristics and the design of the future lidar system and different calibration aspects, this lidar could be suitable for long-term monitoring of water in the upper troposphere – lower stratosphere. The future system will allow us to provide data for the validation of present satellite experiences like AURA-MLS, AQUA-AIRS and more recently the Indian-French project MEGHA TROPIQUES which launched at the end of 2011 and which host the instrument SAPHIR (micro-wave radiometer). Flexibility in the design (emitted power, wavelengths, calibration techniques...) will enable to improve the performances of this instrument on the long-term to fully reach on operational system in the tropic for water vapor monitoring up to the low stratosphere.

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8

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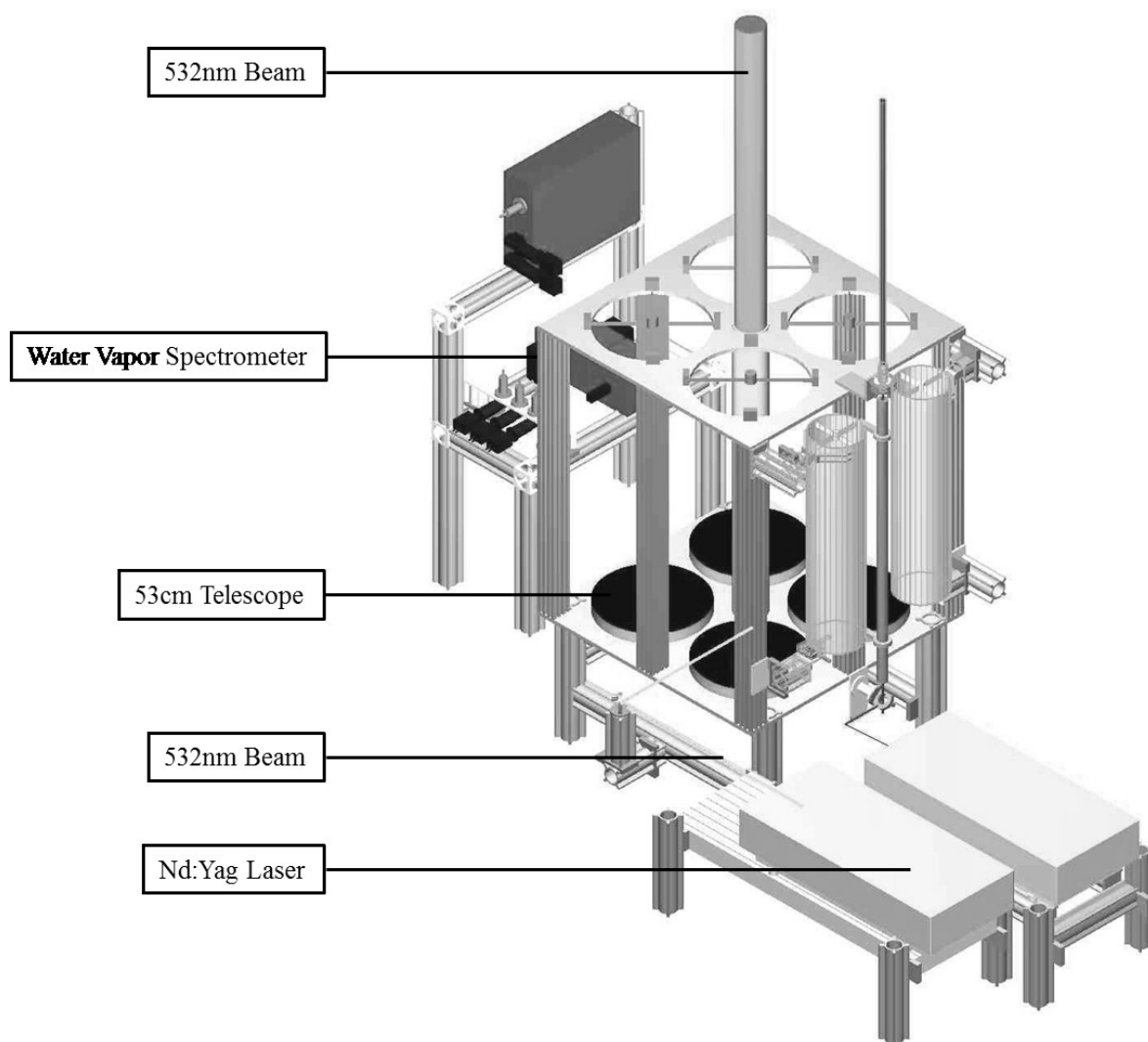


Figure 1. Representation of the lidar instrument at Reunion Island university during the 2002-2005 period.

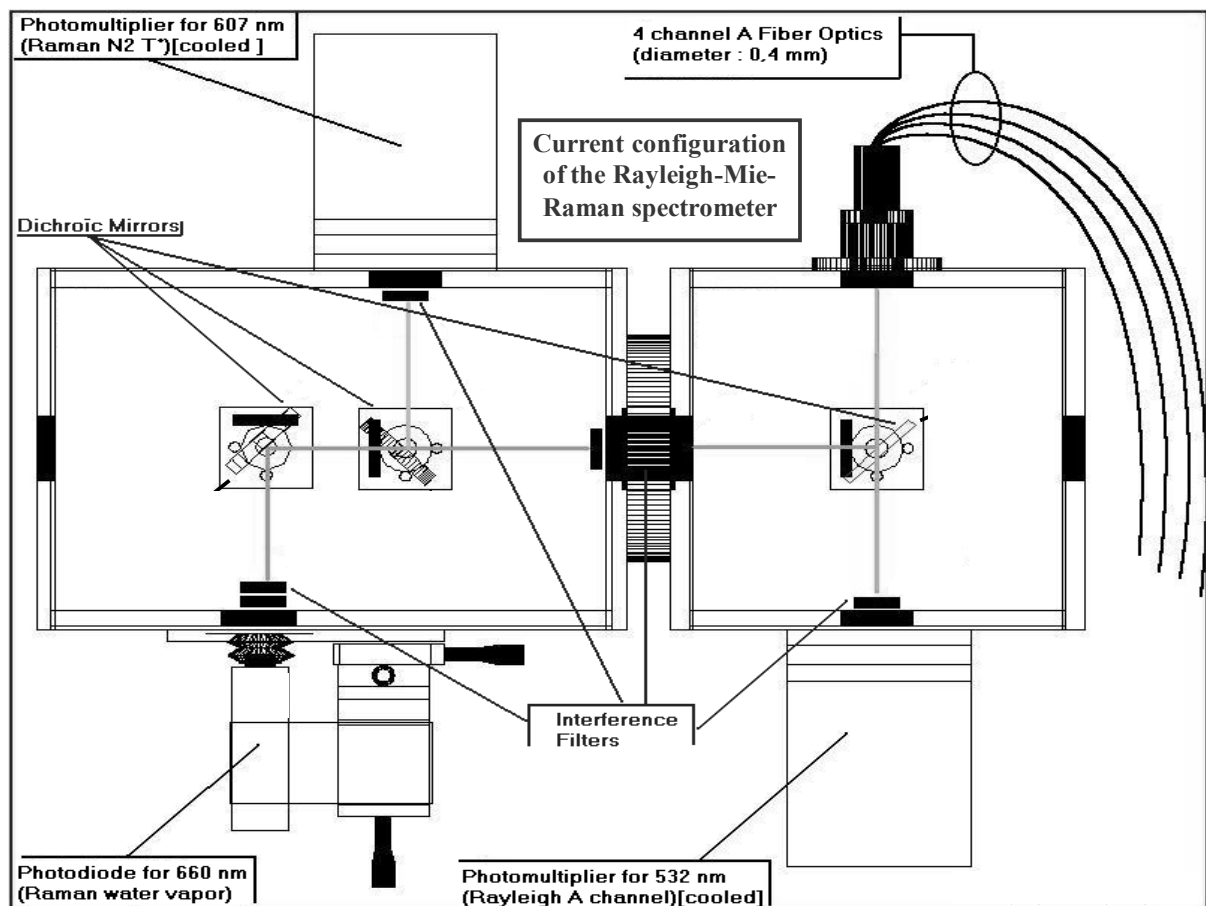


Figure 2. Schematic representation of the current configuration of the Rayleigh-Mie-Raman spectrometer of the preliminary lidar system.

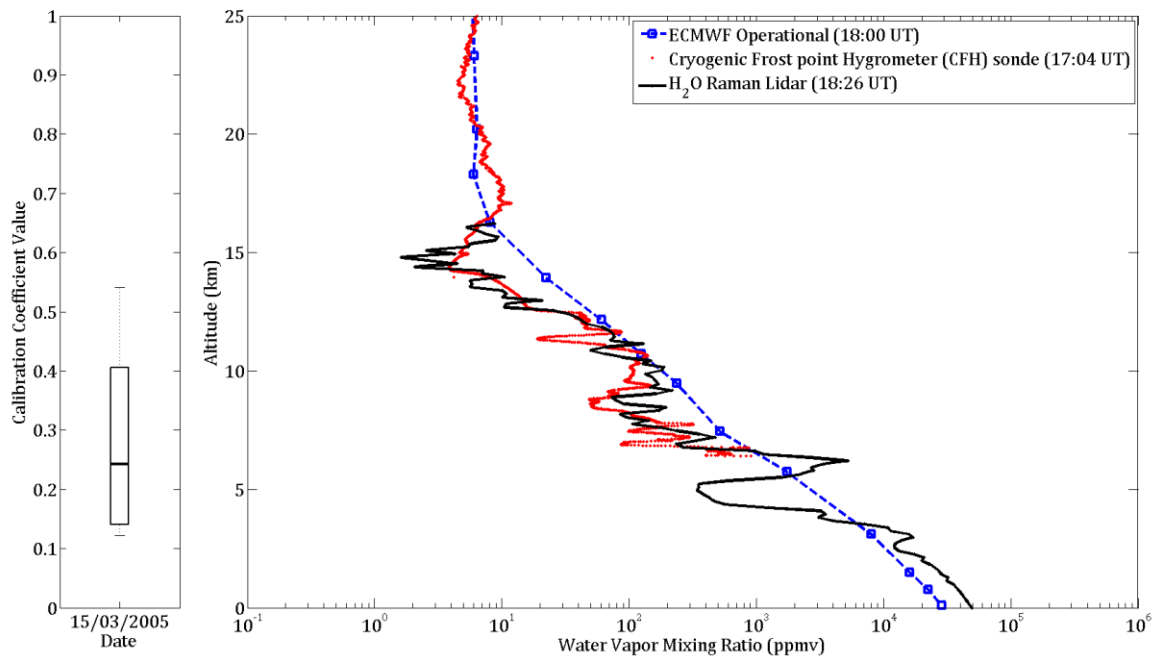


Figure 3. Intercomparison between H₂O Raman lidar (black line), Cryogenic Frost point Hygrometer (CFH) sonde (red dots) and ECMWF operational analysis (blue line) on 15 March 2005. The calibration coefficient and his error are shown on the left. Calibration has been realized using ECMWF Operational data before comparison with CFH.

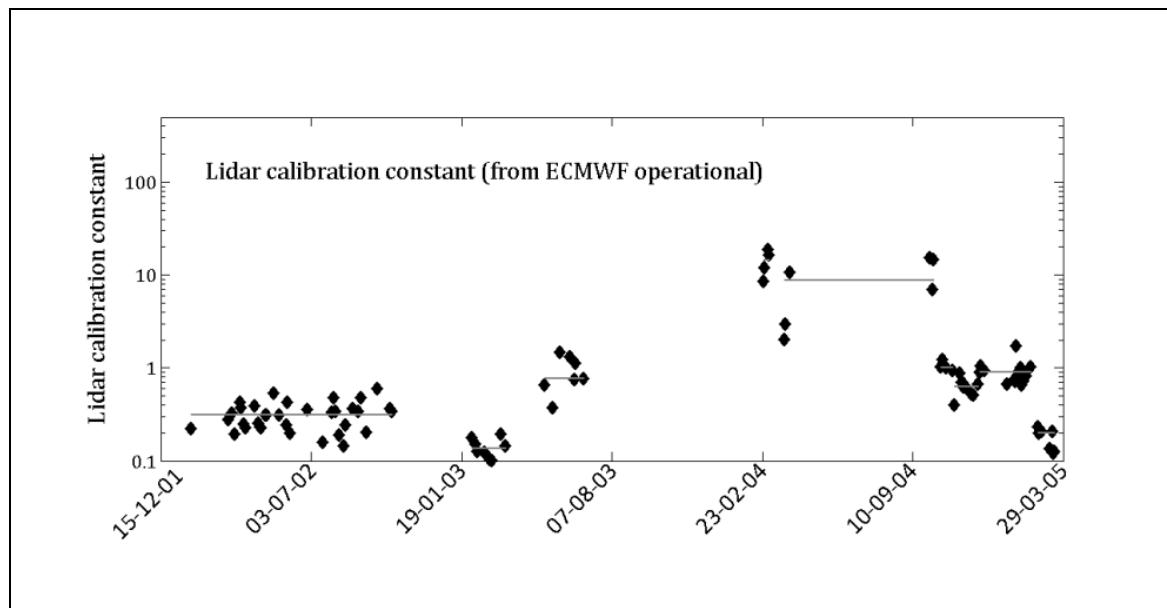


Figure 4. Time evolution of the actual lidar calibration constant over the period 2002-2005. The horizontal grey lines represent the median of these coefficients.

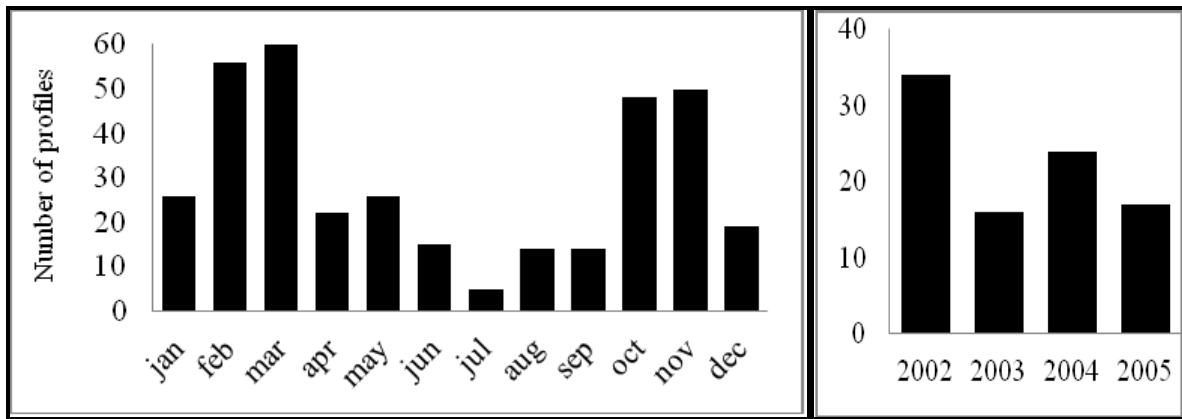
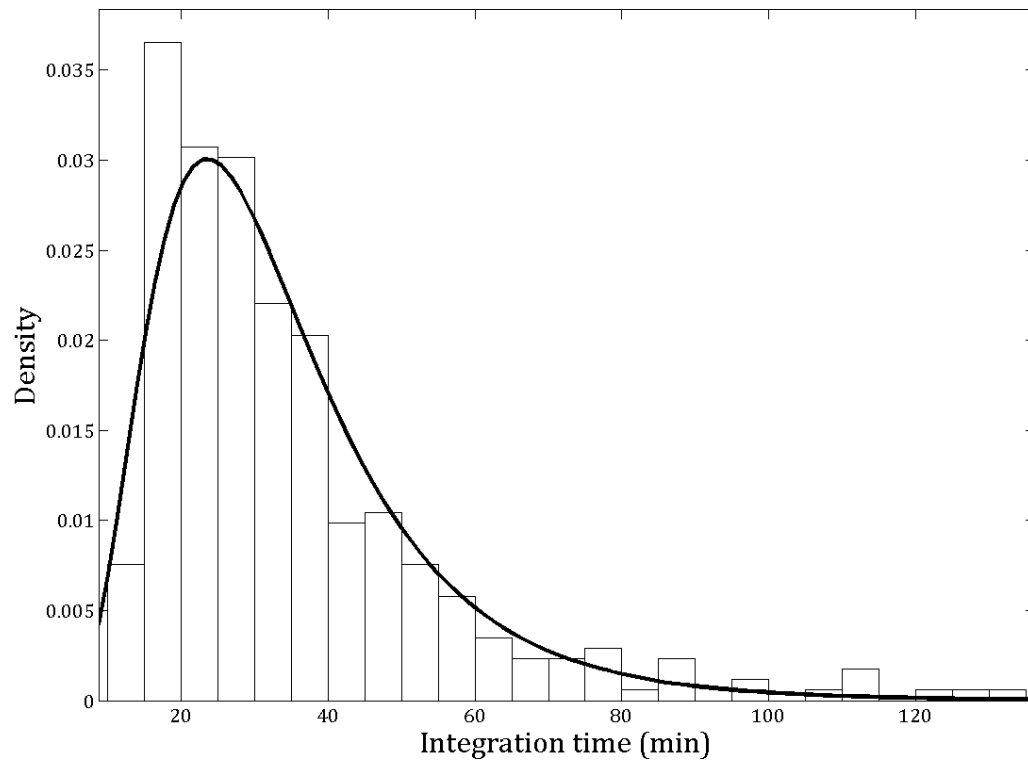


Figure 5. Dataset histogram over the period 2002-2005.



1

2 Figure 6. Integration time distribution of the vertical profiles of water vapor mixing ratio. The
 3 bold black line represents a log-normal fit of the distribution which represents an adequate
 4 approximation of the histogram.

5

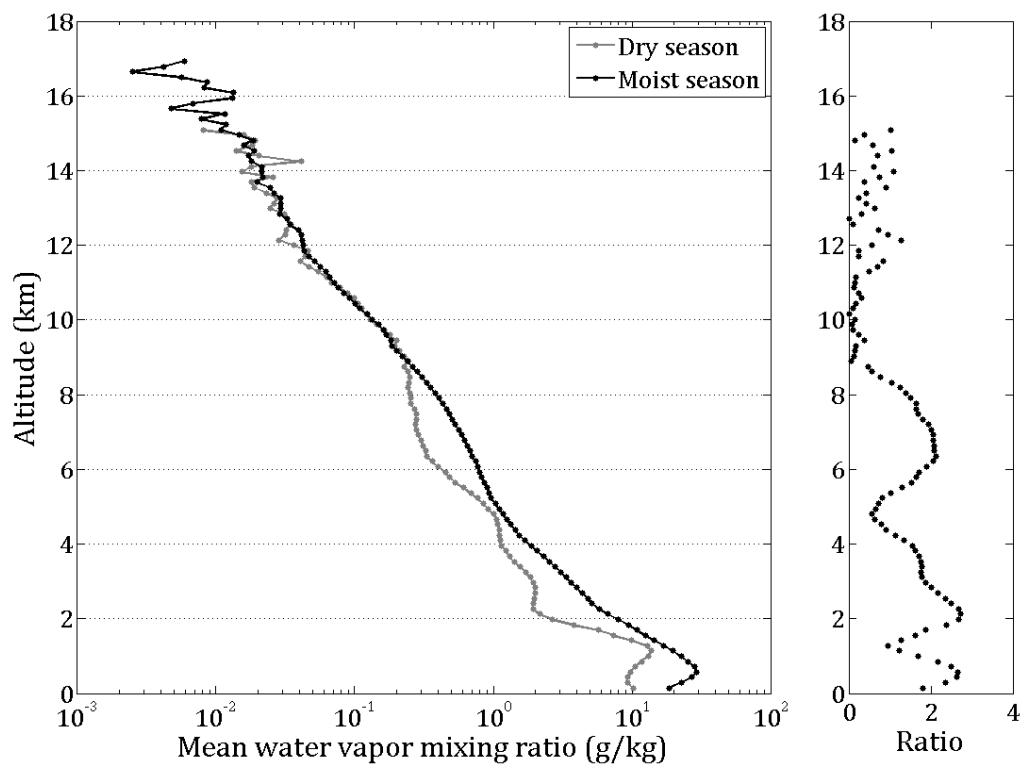
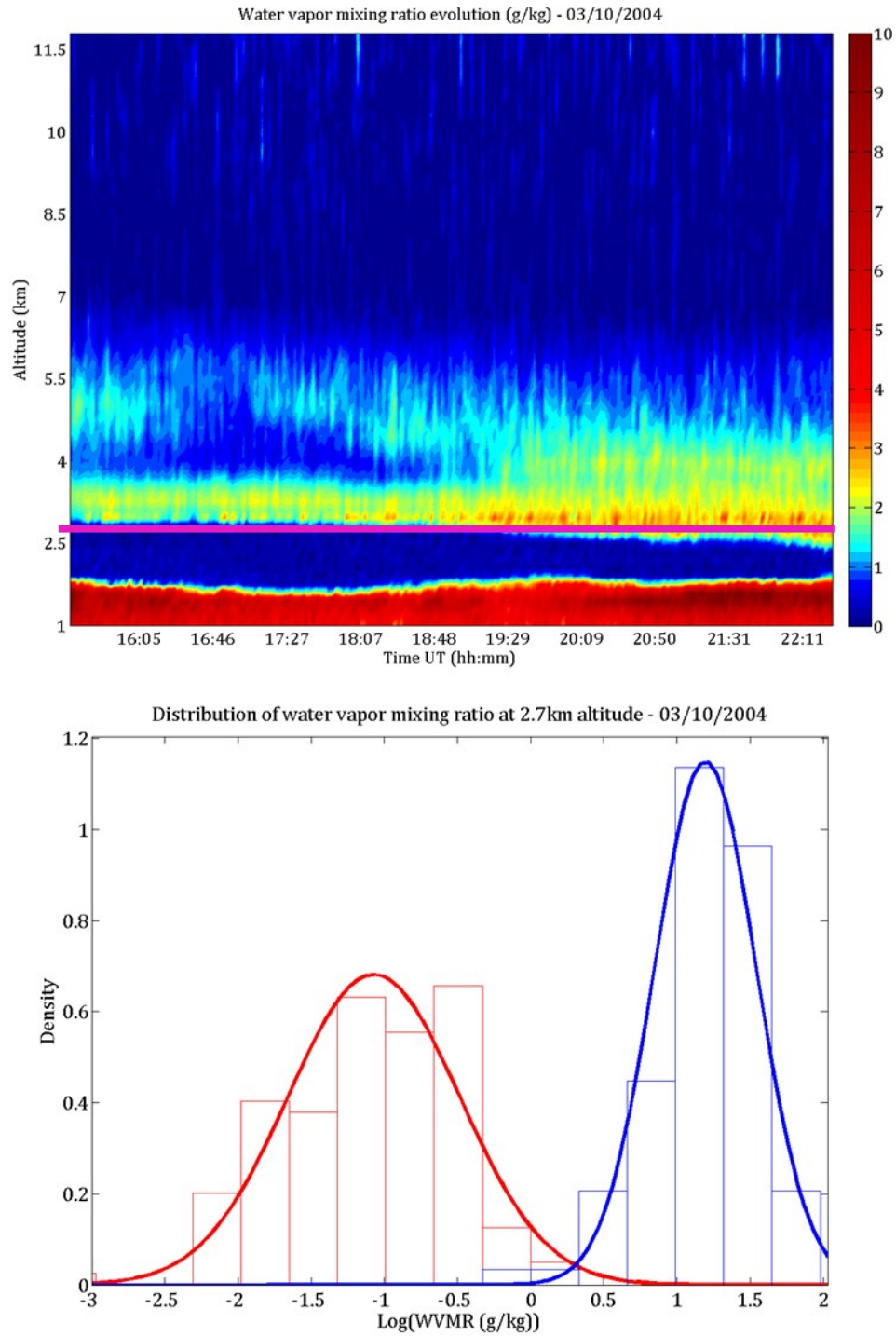


Figure 7. Mean water vapor mixing ratio vertical profiles during the period 2002-2005 regarding the moist and dry season (left panel) and the ratio between the seasons (right panel)



1

2 Figure 8. Evolution of water vapor mixing ratio to 3rd October 2004 (top panel). Pink dotted
 3 line represents an example of location (here at 2.7km altitude) where the distribution of water
 4 vapor are bimodal (bottom panel). On the right panel, red line represents the water vapor
 5 distribution before 19:30 and the blue line the distribution after 19:30.

6

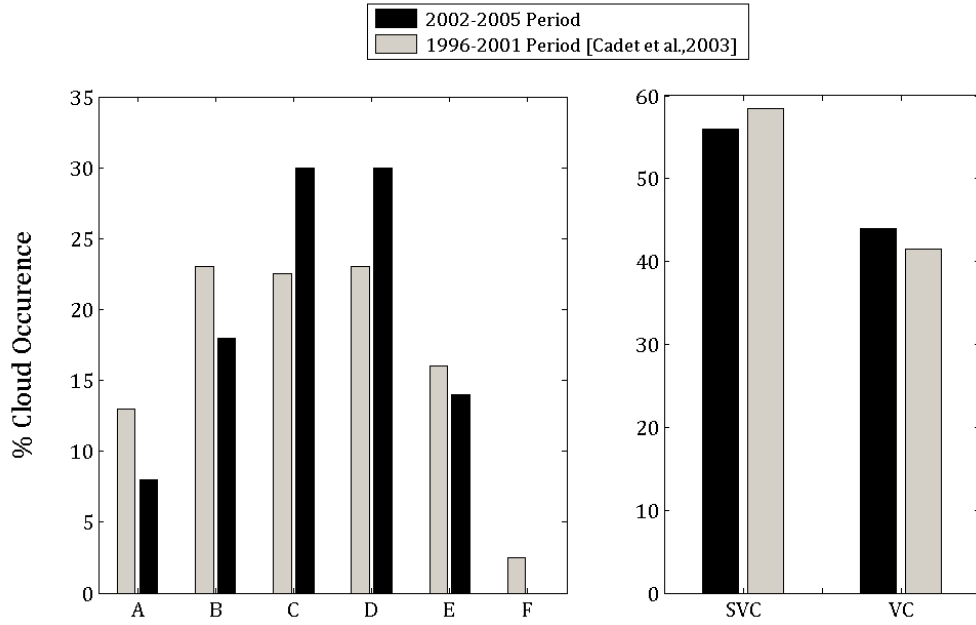


Figure 9. Histograms of cirrus optical thickness binned on a log scale. Lettering along the abscissa (left panel) corresponds to the following optical thickness intervals: A=(0.001-0.003), B=(0.003-0.01), C=(0.01-0.03), D=(0.03-0.1), E=(0.1-0.3), F=(0.3-1.0). The right panel shows the combined results according to subvisible cirrus (SVC) and visible cirrus (VC). Categories A-C comprises the SVC component and D-F comprises the VC component.

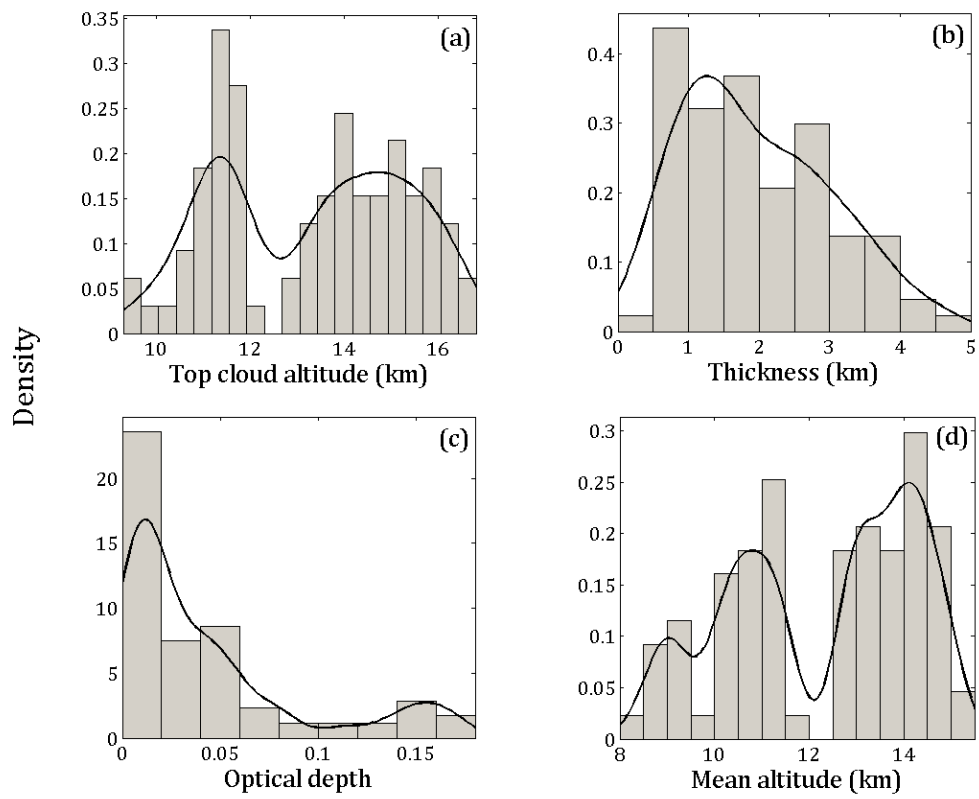


Figure 10. Probability Density Function (PDF) of the different characteristics of cirrus clouds observed at Reunion Island over the period 2002-2005. The panels (a), (b), (c) and (d) represent respectively the top of cirrus clouds, geometrical thickness, the optical depth and the mean altitude.

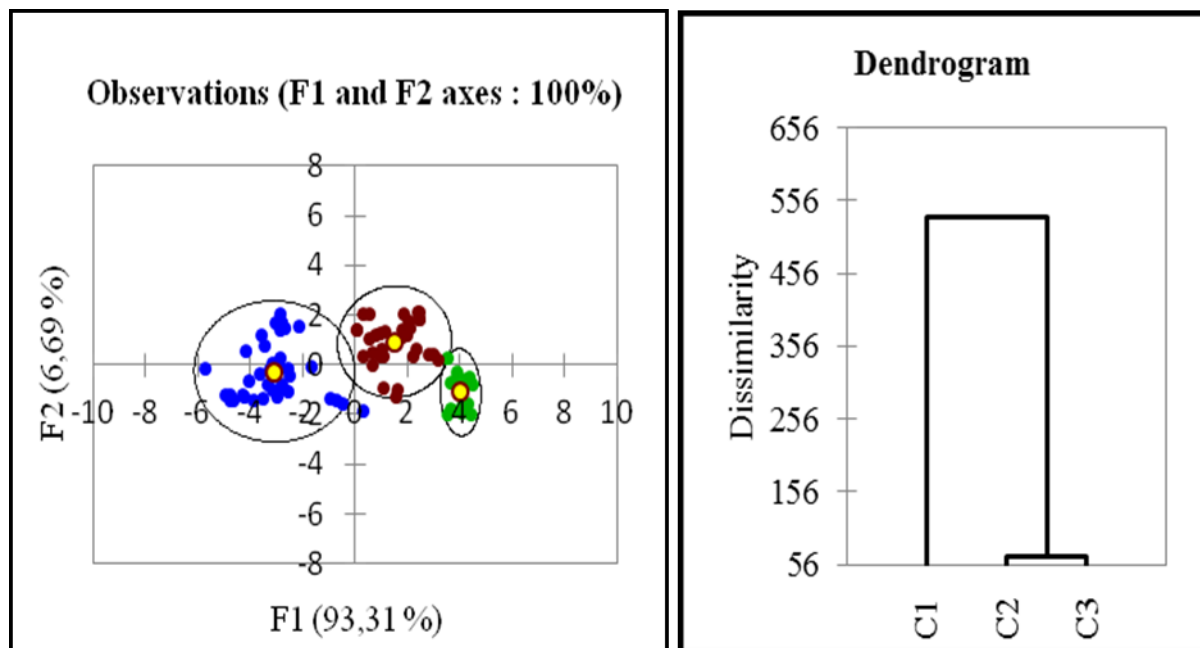


Figure 11. Virtual representation of the observations on discriminated axis showing the different classes and performed from the discriminant factor analysis (on the left panel). Circles represent different identified classes with a confidence interval of 95%. Blue dots represent the first class (midtroposphere thin cirrus), red dots represent the second one (thick upper troposphere cirrus) and the green dots represent the third class (upper troposphere thin cirrus). Dendrogram derived from the hierarchical ascendant classification is shown on the right panel.

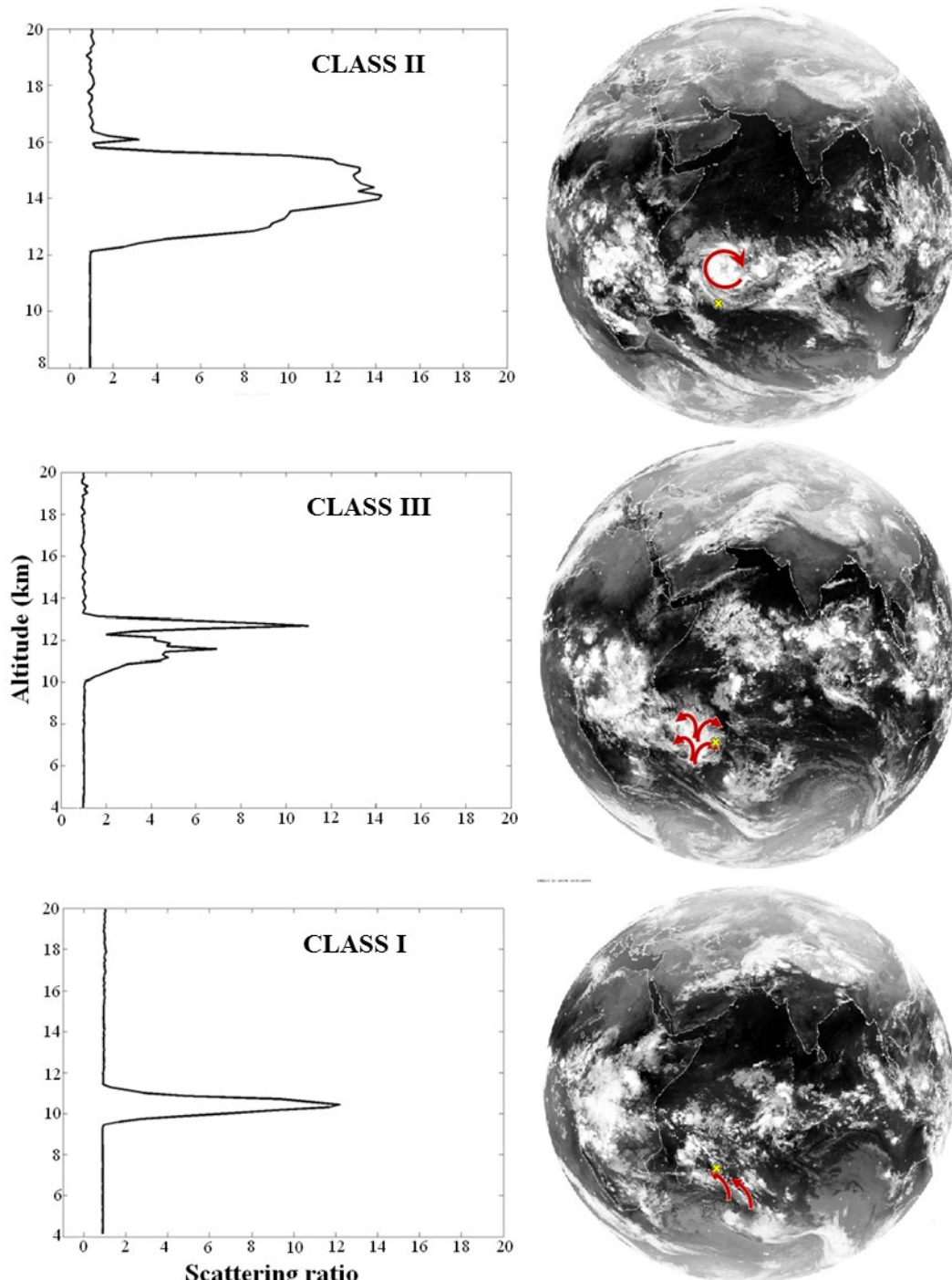
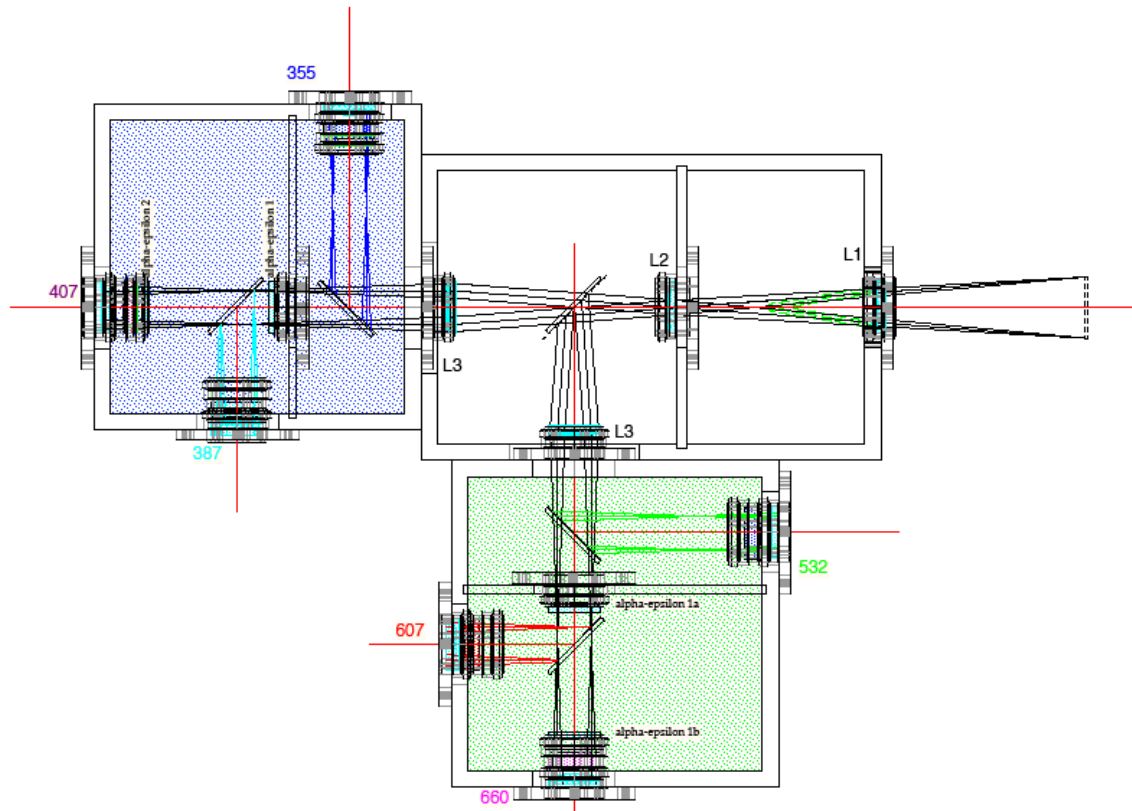


Figure 12. Representation of different cirrus cloud classes. Left panels represent the scattering ratio profiles for each class and right panels represent the corresponding MeteoSat image. Reunion Island is indicated by yellow cross. Atmospheric move is represented by red arrows. The corresponding dates are 16th March 2005, 15th December 2004 and 9th February 2003 (from the top to the bottom).



1
2

3 Figure 13. Schematic view of the optical ensemble of the future lidar implemented at
4 Observatoire de Physique de l'Atmosphère de La Réunion (OPAR).

5

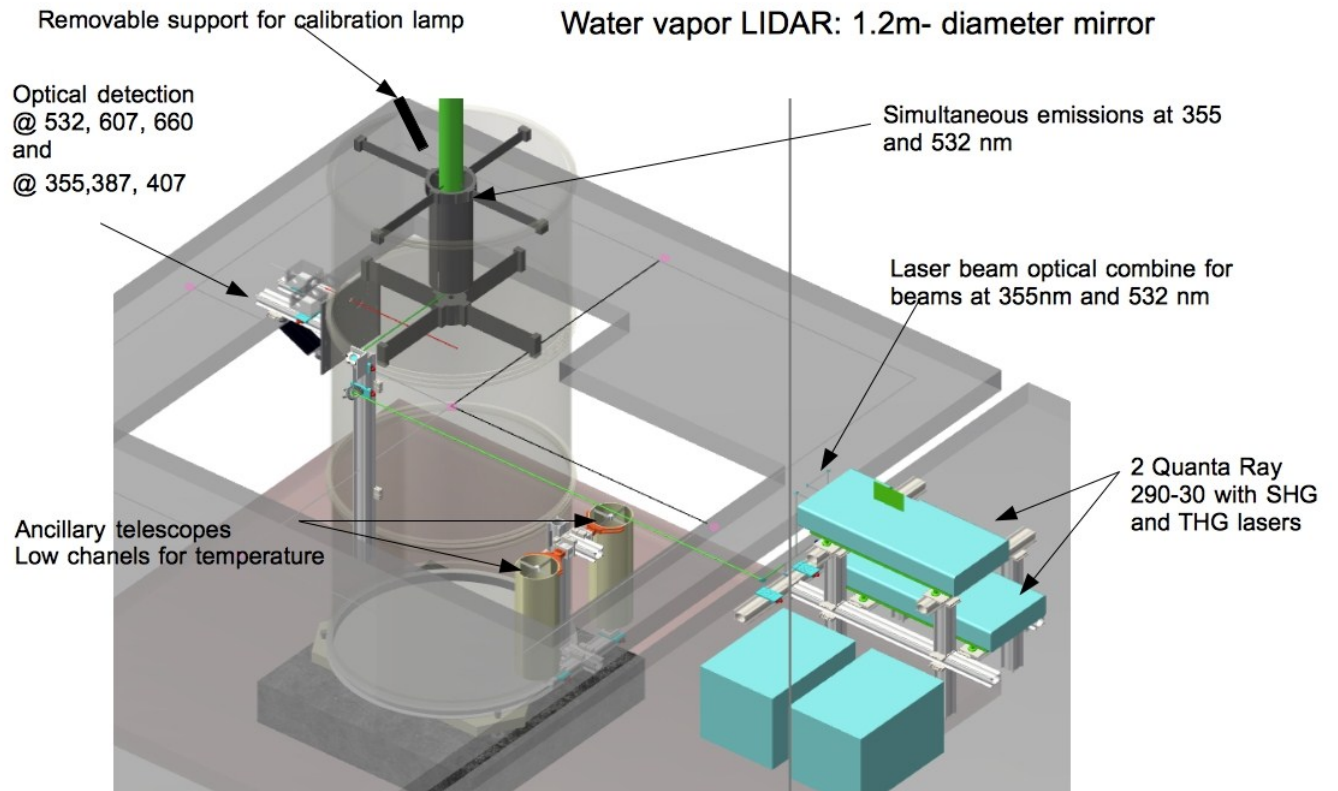


Figure 14. Representation of the design of the future lidar implemented at the Reunion Island altitude observatory.

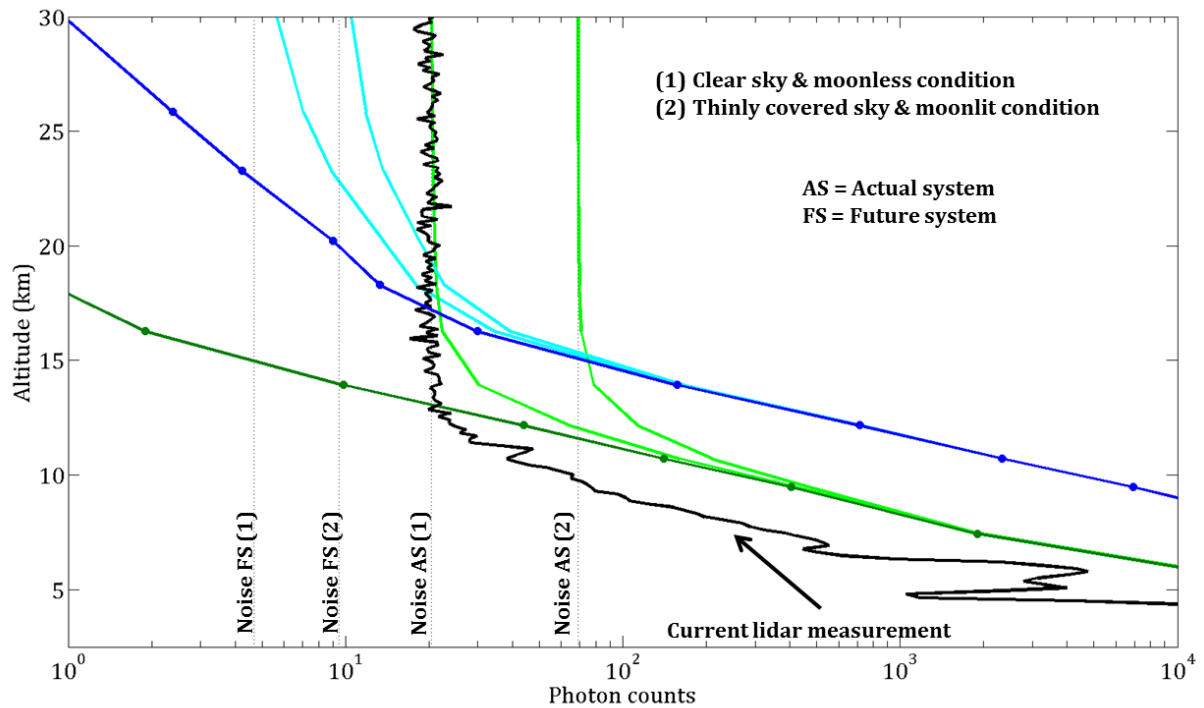


Figure 15. Raman H₂O backscattered signal comparison (30 minutes time integration). The bold black line with dots corresponds to the numerical simulation regarding the current lidar system and the bold black line with square for the future one with one laser. The grey lines correspond to the numerical simulation + noise (sky background and detector noise) following different weather conditions: dark grey lines for the actual system and light grey lines for the future system. Weather conditions are indicated on the figure.

TABLE 1. Characteristics of the three cirrus classes

Class type	I. Midtroposphere cirrus	thin	II. Thick upper troposphere	III. Upper troposphere thin ci
Occurrence (%)	44		19	37
Mean altitude (km)	10.3 ± 0.9		14.3 ± 0.6	13.6 ± 0.6
Thickness (km)	2 ± 1.1		3 ± 0.7	1.4 ± 0.5
Optical depth	0.04 ± 0.04		0.09 ± 0.06	0.02 ± 0.02
Top altitude (km)	11.3 ± 0.8		15.9 ± 0.4	14.3 ± 0.6

1 TABLE 2. System Parameters for H₂O channel for the actual and future lidar system

System Parameter	Parameter Value	
	Actual lidar	Future lidar
Received wavelengths	660.5nm	407.5nm
Filter Bandwidth	1nm	1nm
Filter transmission	0.14	0.39
Detector	EG&G SPCM AQ-232 photodiode	R7400-03 Hamamatsu photomultiplier
Quantum Efficiency	0.35	0.22
Total H ₂ O channel efficiency	3.2%	7.6%

2

3

1 TABLE 3. Sky Spectral Radiance for different weather conditions and corresponding sky
2 background for the H₂O channel future lidar system (integration time of 30 min is considered)

Reference	λ [nm]	Period	L_λ [W.m ⁻² .sr ⁻¹ .nm ⁻¹]	N_0	N_0/N_D
Höhn and Büchtemann (1973)	400	Night (1,2)	$3.4.10^{-9}$	0.34	0.08
Höhn and Büchtemann (1973)	400	Night (3)	$1.9.10^{-8}$	1.95	0.45
Höhn and Büchtemann (1973)	400	Night (4)	5.10^{-8}	5.14	1.19

3

4 (1) Clear sky (moonless), (2) Thinly covered sky, haze or thin fog (moonless)

5 (3) Clear sky (moonlit), (4)Thinly covered sky, haze or thin fog (moonlit)

6 The typical dark count rate of R7400-03 Hamamatsu photomultiplier $d=80$ s⁻¹ is used.

7 N_0 and N_D are, respectively, the night sky background and the detector noise (in photon).