

Response to Editor

Dear Editor & Prof.

We greatly thank you and the two reviewers for the thorough and valuable suggestions to our work, and thank you valuable comments and suggestions by peer experts in the open discussion. The manuscript has been polished and modified by professional organizations, and English has been greatly improved. We have made a point-to-point response to these opinions and suggestions, and believe that the quality of the manuscript has been promoted now. All the comments have been addressed in the revised manuscript, and the responses to each comment are given below.

Thank you very much for considering our work!

Yours sincerely,

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Response to reviewer #1 (in open discussion)

General comments:

This manuscript presents an observational combined lidar (1064 nm) and radar (8.6 mm) data set to determine the cloud boundaries over the ground station located at the Xian region. The authors use signal enhancement techniques to avoid background and aerosol signal thereby improving the SNR for the identification of cloud top and base boundaries. Analysis of one-year data set over the Xian is presented characterizing the cloud cover and single/multiple cloud layer occurrences. Overall, this is an interesting manuscript and has the potential to be published.

We appreciate the reviewer's thoughtful review and constructive comments, which have greatly helped to enrich the details and improve the quality of the paper. These comments have been revised and supplemented in the manuscript, and the responses to each comment are given below. The manuscript has been polished and modified by professional organizations, and I believe that English has been greatly improved.

1. It is well established that combined lidar and radar measurements are essential to monitor the local cloud cycle. This is also acknowledged by the authors here in the manuscript with some references (Line 76) and demonstrated with few case studies (section 4.1). However, the one-year data presented here do not have simultaneous measurements from lidar and radar for about one-third of the total time (7248 hours). This could introduce bias: (a) in the cloud base boundaries as also shown in fig 10 (19:00 – 00:00 hrs) for the cases of cirrus to altostratus transition where the cloud particles would eventually grow into large sizes producing precipitation, and (b) in cloud top boundaries where the ice crystals are too small to be detected by the radar (fig 10, 19:00 – 20:00 hrs). This bias should be mentioned in the abstract and needs to be discussed in the main text.

Response: As discussed and analyzed in Section 4.1, MMCR can effectively measure cloud tops compared, but it has no advantage over lidar in detecting cloud bottom. We analyzed the correlation of cloud bottom (0.803) between obtained by MMCR after data control and detected by lidar, and considered that the two instruments have a high correlation for the detection of cloud bottom height. Therefore, we used MMCR data (7248 hours) aided by lidar data (872.5 hours) to improve the accuracy of cloud bottom detection. To reduce the error caused by directly employing MMCR to analyze the cloud bottom. Two biases (a) and (b) have been added to the abstract and discussed in the main text.

Specific modifications are as follows:

1) L 12-L 17: We analyzed three typical cases (e.g., single-layer clouds, multilayer clouds, and precipitating clouds), case one presents two interesting phenomena: a) at 19:00~20:00, the ice crystal particles at the cloud top boundary are too small to be detected by MMCR, which is well detected by lidar. b) at 19:00~00:00, the cirrus cloud transits to the altostratus where the cloud particles would eventually grow into large sizes producing precipitation, and MMCR has more advantages than lidar in detecting the cloud top boundary within this period.

2) L 248-L 254: At 19:00 ~ 20:00, in cloud top boundaries where the ice crystals are too small to be detected by the MMCR, but the lidar detects the real cloud top. The main reason is that the echo intensity of MMCR is proportional to the 6th power of particle diameter, and the lidar echo signal is proportional to the square of particles. From 19:00 to 00:00, the cirrus cloud transits to the altostratus cloud, where the cloud particle size increases in the form of collision and finally produce precipitation. In this process, the lidar beam entering the cloud is attenuated, but MMCR has a good advantage in cloud top detection.

2. I suggest including the lidar wavelength or spectral region in the title (and abstract) of the manuscript, since this often gives the impression that lidar is operated at visible channel (532 nm) – if specific wavelength or type of instrument is not mentioned. Further the term ‘statistical analysis’ in the title is misleading. To my understanding there is no statistical analysis in this manuscript, rather the authors just present the frequency of cloud top/base altitude occurrences and its seasonal variability.

Response: We changed the title of the manuscript to “Detection and analysis of cloud boundary in Xi’an, China employing 35 GHz cloud radar aided by 1064nm lidar”

3. Extensive editing of the manuscript is required for the proper English usage.

Response: The manuscript has been polished and revised by professional institutions.

Specific Comments:

1. L 49: ‘Pal et al’ is repeated. There are several instances throughout the manuscript where the citations embedded in the sentence has repeated words.

Answer: We have modified all the similar situations in manuscript, as follows:

- 1) L 49 ‘The differential zero-crossing method proposed by Pal et al. (Pal et al.,1992)’ is changed to ‘ The differential zero-crossing method proposed by Pal et al. (1992)’
- 2) L58 ‘Morille et al. (Morille et al., 2007) determined the local maxima on both sides of the cloud peak as the cloud...’ is changed to ‘ Morille et al. (2007) determined the local maxima on both sides of the cloud peak as the cloud’
- 3) L 60 ‘underestimated, respectively. Mao Feiyue (Mao et al., 2011)’ is changed to ‘ underestimated, respectively. Mao Feiyue (2011)’
- 4) L65 ‘Kollias et al. (Kollias et al., 2007) judged the SNR value’ is changed to ‘Kollias et al. (2007) judged the SNR value’
- 5) L67 ‘Clothiaux et al. (Clothiaux et al.,1999) used 35 GHz millimeter wave cloud measuring radar’ is changed to ‘Clothiaux et al. (1999) used 35 GHz millimeter wave cloud measuring radar’
- 6) L170 ‘Referring to the empirical formula proposed by Riddle (Riddle et al., 1989)’ is changed to ‘Referring to the empirical formula proposed by Riddle (1989)’
- 7) L379 ‘Zhao et al. (Zhao et al., 2014) at the SGP site and Hailing Xie (Xie et al., 20217)’ is changed to ‘Zhao et al. (2014) at the SGP site and Hailing Xie (20217)’

2. L 115: Please use appropriate standard literature reference for the elastic backscattering lidar equation. For example: Measures, R.M., Laser remote sensing: Fundamentals and applications, Willey Publishers, 510 pp, 1984.

Answer: The reference ‘Laser remote sensing: Fundamentals and applications, Willey Publishers, 510 pp, 1984’ does not point out the standard lidar equation, so we refer to the standard radar equation in the reference ‘Wandinger U.: Introduction to Lidar, Brooks/Cole Pub Co, doi:10.1007/0-387-25101-4_1, 2005.’ as follows

$$P(\lambda, r) = P_0 \frac{c\tau}{2} A\eta \frac{O(r)}{r^2} \beta(\lambda, r) \cdot \exp\left[-2\int_0^r \sigma(\lambda, r) dr\right], \quad (1)$$

where λ is the wavelength of the emitted light, r represents the detection distance, and $\beta(\lambda, r)$ and $\sigma(\lambda, r)$ are the atmospheric backscattering and extinction coefficients, respectively. $O(r)$ is the laser-beam receiver field-of-view overlap function, c is the speed of light, P_0 is the average power of a single laser pulse, τ is the temporal pulse length, η is the overall system efficiency, and A is the area of the primary receiver optics responsible for the collection of backscattered light.

3. L 133: Details on the wavelet function used should be mentioned here.

Answer: See the answer 4 below.

4. L 135: Complete description of the flow chart processes – variables are missing. For example, the

variables/symbols R_s , id , Pe , Ma , and Mi shown in figure 2 are nowhere defined.

Answer: After the statistical analysis of the system noise, we set $k = 4$ in this study. The algorithm flow chart of detecting cloud boundary by lidar is shown in Fig. 2. Usually, the moving average of $P_{new}(\lambda, r)$ of lidar echo signal is calculated to reduce the influence of random noise. However, the selection of a sliding window directly affects the signal quality. Therefore, $P_{new}(\lambda, r)$ is denoised by wavelet transform, threshold function is a soft threshold, wavelet base is sym7, and the number of decomposition layers is 5. Using wavelet function to reduce noise can avoid too much smoothing remove sharp signal changes due to clouds, and can also avoid the improper selection of moving average window. Obtaining cloud boundaries mainly includes three parts. The first part is signal preprocessing. $P_{new_s}(\lambda, r)$ after wavelet de-noising is discretized based on the estimates of noise, and get useful signals $P_{new_s1}(\lambda, r)$ and $P_{new_s2}(\lambda, r)$. The second part is to enhance the signal to make the cloud signal sharper from the background noise and aerosol signal. Average signals $P_{new_s1}(\lambda, r)$ and $P_{new_s2}(\lambda, r)$ to obtain $P_{new_sf}(\lambda, r)$. Ascending arrangement are conducted for $P_{new_sf}(\lambda, r)$ and the new sequence R_s and the corresponding index id are recorded. The maximum and minimum R_s are denoted as Ma and Mi , respectively. By building a new mapping proportion coefficient $Pe(i)$, the enhanced signal $P_{new_sp}(\lambda, r)$ is obtained. Obtain slope of baseline 1, and obtain baseline 2 based on this slope. Signals exceeding baseline 2 are regarded as candidate cloud signals as shown in Fig. 3b) and Fig. 4b). The third part is to extract cloud signal and realize boundary detection by combining the SNR of echo signal. By fitting the echo signal slope in the height range of 15–20 km, the slope is used as the bottom slope to distinguish the cloud and aerosol layers (as shown by the magenta line in Fig. 3b and Fig. 4b). Without considering the bottom echo signal (0–2 km), the amplitude of the echo signal received by the lidar decreased with increasing detection height according to the fitted slope, as shown by the blue line baseline in Figs. 3b) and 4b). When the beam senses the presence of clouds, the amplitude of the echo signal will exceed the blue baseline.

5. L 138: What about the cases when high level clouds exist? It is well known that cirrus types of clouds occur high in the troposphere extending to the tropopause (on average ~17 km during summer over the subtropics).

Answer: In line 138, the 15-20km height range used to fit the slope is applicable to Xi'an region (in the data of the past two years, there are few clouds higher than 12km). When there are high-level clouds, the range of slope fitting should be the echo signal above the high-level clouds.

6. L 206: Do you mean the 2 min time-resolution lidar profiles are duplicated 24 times to make it look like 5 sec temporal resolution data? Please mention this clearly.

Answer: No, we perform linear interpolation on the lidar data within 2min, and keep its time-resolution consistent with MMCR, that is, the 2min one group of data becomes 2min 24 groups.

7. L 355: I don't think these are statistical rules. Please replace the term 'statistical rules' with 'logic-based' rules or something like that throughout the manuscript

Answer: L 355 ‘The statistical rules shown in Table 3’ has been replaced by ‘The cloud bottom height recording guideline in the Table3’. Other ‘statistical rules’ in the manuscript have been modified or replaced, and the specific details are given in the answer 15.

8. L 359: The case1 in Table 3, please be specific if you mean ‘optically thin’ or ‘geometrical thin’ cloud? If it is optically thin than both cloud top and base can unambiguously be determined from lidar.

Answer: L 359: ‘Thin cloud’ is changed to ‘geometrical thin’ cloud in Table 3.

9. L 376: Clouds above 8 km has highest frequency in autumn, and these are likely stratus and cumulus clouds? This sentence doesn’t make any sense. Please refer to the WMO cloud classification – stratus and cumulus are low-level clouds those formed within 2 km above the surface level.

Answer: According to the expert opinion, we have re-defined the division of seasons (answer 12), so the observation data has been slightly adjusted, and the corresponding expression has changed. See the answer 10 below for details.

10. L 378: “height range of clouds is narrow, and the numerical range is wide”? Please re-write this sentence for more clarity mentioning the height range you are referring to.

Answer: For vertical distribution of cloud base, the first narrow peaks is boundary layer clouds (≤ 1.5 km) , and the second peak is 2.5 ~ 3.5 km, and the third peak has a big range in vertical height, which is around 4.7-10 km a in spring. Fig.18 b) expresses that the cloud bottom height in summer is mainly distributed at 3-9.5km, indicating that the middle and high cloud may be dominated. The distribution of cloud bottom shows the bimodal, the first peak is the boundary layer cloud peak, and the second peak is located at 2.7-3.7 km and 3.6-8.3 km in autumn and winter, respectively.

11. L 383: Why the data presented in the figures showing vertical distribution of frequency of cloud occurrences are limited to 12 km? Or is this an underestimation of cloud top boundaries owing to the sensitivity of 8.6 mm radar? It is not uncommon to have high level clouds extending up to 15 km or more in the region of interest. Please present the results upto the tropopause level.

Answer: In July, 2021, the detection distance base of MMCR increased from 420 to 600, that is, the maximum detection range increased from 12.6km to 18km. We checked the echo data of MMCR (the maximum detection range is 18km) for 199 days one by one. Among them, only one day's data show that cirrus clouds existed at about 13km, and only four days' data show that the cloud top was slightly more than 12km. Therefore, our analysis of cloud boundary is limited to 12km. At the same time, we have added the specific time when the maximum detection range of MMCR changes in the manuscript. Add description as ‘During the 12-month observation, the maximum detection range of MMCR has changed. From December 2020 to June 2021, the maximum detection range of MMCR is 12.6km, and then the maximum detection height is changed to 18km.’

12. L 390: Before discussing the result, it would be beneficial to briefly describe how normalized cloud cover is computed here. Also, indicate the months of the season – spring (MAM), and summer (JJA). How is the maximum cloud cover 2.46 in summer?

Answer:L 356-358 ‘The experimental data of 302 days (65 days in spring (January-March), 84 days in summer (April-June), 65 days in autumn (July-September) and 88 days in winter (October-December) observed in 2021 are classified and sorted out to ease the statistics and analysis of the variation characteristics of cloud boundary height’, which is no solar term to define the season, so we re-describe it as ‘From the above three cloud observation cases, it can be seen that MMCR has more advantages than lidar in detecting cloud-top boundaries. Therefore, when calculating the cloud boundary height distribution characteristics over Xi'an, we only counted the cloud top boundary height detected by the MMCR and considered it as the actual cloud top boundary. From December 2020 to November 2021, MMCR and lidar stored 302 d (7248 h) and 126 d (872.5 h) of observational data, respectively. During the 12-month observation period, the maximum detection altitude of the MMCR changed. From December 2020 to June 2021, the maximum detection range of MMCR is 12.6 km, and the maximum detection height is changed to 18 km. The total observation hours of MMCR and lidar for each month are shown in Fig. 15. The hours of lidar, MMCR, and simultaneous measurements are 872.5 h. In this study, the four seasons were defined as follows: spring from March to May (MAM), summer from June to August (JJA), autumn from September to November (SON), and winter from December to February (DJF).’

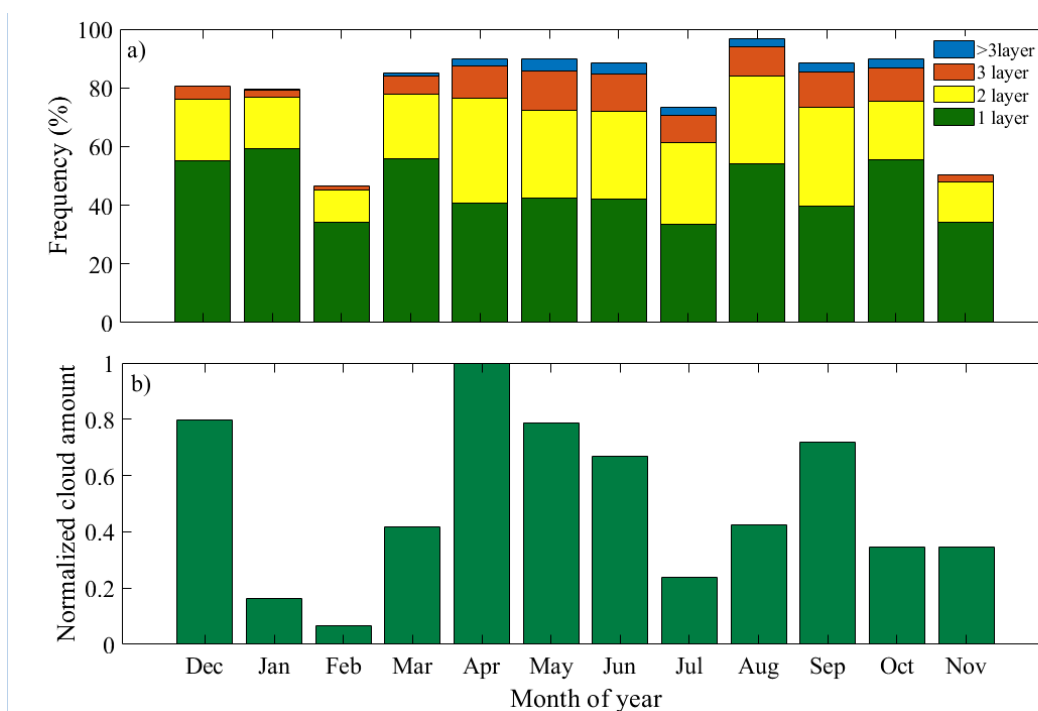


Fig. 19 Monthly variation in cloud frequency distribution and cloud cover from December 2020 to November 2021
a) monthly variation in the frequency of the number of cloud layers. b) monthly variation in cloud cover

MMCR defines cloud cover as the percentage of cloud obscuring sky field of vision. Cloud cover observation includes total, low, medium and high cloud cover. Total cloud cover refers to the total number of cloud cover in the sky during observation (Fig.18b shows the total cloud cover in every month). Generally, the sky is divided into 10 parts. When there is no cloud in the clear sky or less than 0.5 parts are covered, the cloud cover is zero. The cloud covers half of the sky and the cloud cover is 5. Cover the whole sky with clouds and the cloud cover is 10. Calculation steps: 1): divide

the cloud layer into high, medium and low families through the radial cloud base height. 2): average each cluster for 30 minutes. 3): Weighted Processing of data in 10 minutes to obtain the integrated cloud cover. Because the calculated cloud cover is a relative value, it does not mean the real cloud cover. Figure 18b shows that the cloud cover is the largest in April. Therefore, the cloud cover in April is set to 1, and the cloud cover in other months is calculated to represent the relative change trend of cloud cover in each month.

‘The maximum cloud cover 2.46 in summer’ is changed to ‘It can be seen from the distribution of cloud cover in every month that there are relatively more cloud cover in summer and the least in winter, indicating that warm atmospheric conditions are more conducive to the formation and development of clouds.’

13. L 394: I suggest adding fig 18c showing the total monthly hours of lidar, radar and simultaneous measurements in this figure. This is essential to understand the reported cloud characterization.

Answer: The total observation hours of MMCR and lidar in each month are shown in Fig. 17. The hours of lidar, MMCR and simultaneous measurements is 872.5 hours. Considering the logic of the manuscript, we decided to put the Figure 17 in L364 in subsection 4.1.

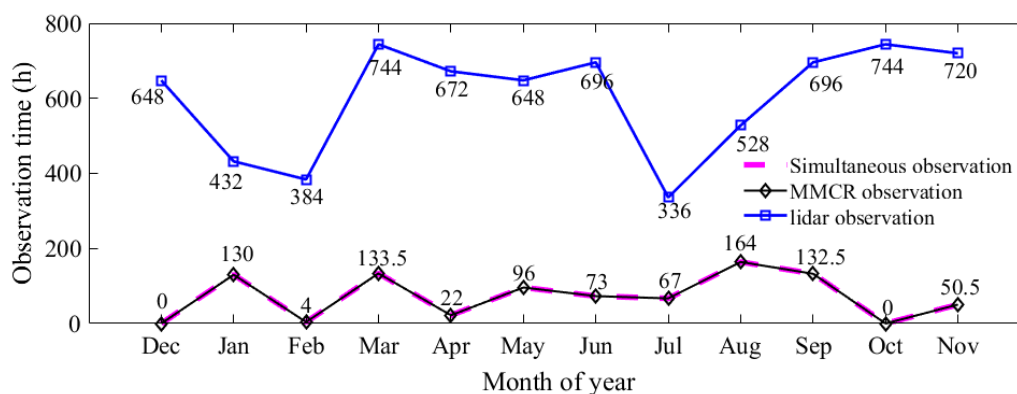


Fig. 17 Monthly observation hours of lidar and MMCR

14. L 396: ‘frequency change characteristics...’? This does not make any sense. As the figure caption says it is the frequency distribution of cloud boundaries observed over Xian in 2021.

Answer: L 396: ‘Fig. 19 shows the frequency change characteristics of the cloud boundary vertical height distribution in 2021’ is changed to ‘As the Fig.20 caption says it is the frequency distribution of cloud boundaries observed over Xian from December 2020 to November 2021’.

15. L 424: Remove the word ‘statistical’.

Answer: L 424: The word ‘statistical’ has been removed. The modified expression is ‘Based on the analysis of the changes and distribution of cloud boundaries in Xi'an from December 2020 to November 2021.’ At the same time, we have modified and replaced the word ‘statistical’ in other parts to make it closer to the aim of the manuscript. Such as ‘Table 3 Statistical rules of cloud bottom boundary information’ is changed to ‘Table 3 Cloud bottom height recording guideline.’ The word ‘statistical’ in L14 has also been removed.

Response to reviewer #2 (in open discussion)

Major comments:

This manuscript combines lidar and Ka-band millimeter-wave cloud radar (MMCR) to study the cloud macrophysical properties in Xi'an. The authors propose a local method for lidar and MMCR, but without enough details. It would be more interesting if detailed descriptions are added in this manuscript. The statistical analysis is kind of superficial and the English writing needs a full editing. It is difficult to follow for several times. Thus, I recommend a major revision and suggest the authors to rearrange this manuscript carefully.

We appreciate the reviewer's thoughtful review and constructive comments, which have greatly helped to enrich the details and improve the quality of the paper. We have added more details to the manuscript to make the logic of the article clearer and the details more perfect. The manuscript has been polished and modified by professional organizations, and I believe that English has been greatly improved. These comments have been revised and supplemented in the manuscript, and the responses to each comment are given below.

1. The title "Lidar and MMCR applied for the study on cloud boundary detection" indicates the manuscript will mainly focus on instruments and method, while the "statistical analysis of cloud distribution in Xi'an region" imply a systematically study for the local cloud distribution. This causes the keynote of the whole text not clear. Which part the authors want to focus, the method or statistical analysis? This would affect the structure of manuscript. Additional, both the method and statistical analysis of the manuscript as current form are not very clear.

Response: According to your requirements and suggestions, a series of modifications have been made in the manuscript. Reorganize the structure of the manuscript, reorganize the highlights and reorganize the language. In order to more clearly express the aim of the manuscript and consider the observation duration of MMCR and lidar, we have changed the title of the manuscript to "Detection and analysis of cloud boundary in Xi'an, China employing 35 GHz cloud radar aided by 1064nm lidar".

2. The two flow charts of lidar and radar, i.e., Figure 2 and 6, are complex, but the text is too short. I can't tell if they are novel compared with previous methods. If the authors emphasize their method is well-improved, they should carefully introduce this part and show the difference and improvement from others'.

Response: We have modified figures 2 and 6 and added corresponding detailed descriptions as follows.

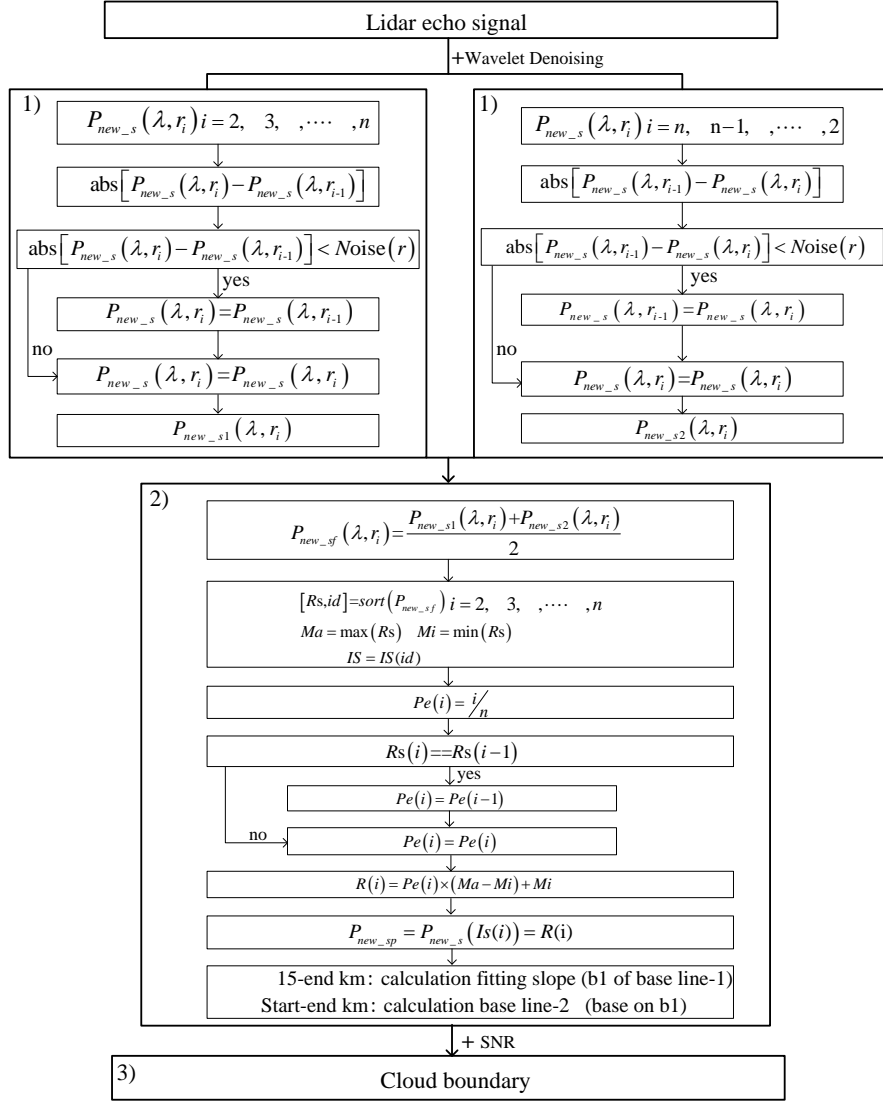


Fig. 2 Use lidar to detect cloud boundary. 1) signal preprocessing, 2) baseline determination based on enhanced signal, 3) identifying cloud boundary with SNR

For revised figure 2: After the statistical analysis of the system noise, we set $k = 4$ in this study. The algorithm flow chart of detecting cloud boundary by lidar is shown in Fig. 2. Usually, the moving average of $P_{new}(\lambda, r)$ of lidar echo signal is calculated to reduce the influence of random noise. However, the selection of a sliding window directly affects the signal quality. Therefore, $P_{new}(\lambda, r)$ is denoised by wavelet transform, threshold function is a soft threshold, wavelet base is sym7, and the number of decomposition layers is 5. Using wavelet function to reduce noise can avoid too much smoothing remove sharp signal changes due to clouds, and can also avoid the improper selection of moving average window. Obtaining cloud boundaries mainly includes three parts. The first part is signal preprocessing. $P_{new_s}(\lambda, r)$ after wavelet de-noising is discretized based on the estimates of noise, and get useful signals $P_{new_s1}(\lambda, r)$ and $P_{new_s2}(\lambda, r)$. The second part is to enhance the signal to make the cloud signal sharper from the background noise and aerosol signal. Average signals $P_{new_s1}(\lambda, r)$ and $P_{new_s2}(\lambda, r)$ to obtain $P_{new_sf}(\lambda, r)$. Ascending arrangement are conducted for $P_{new_sf}(\lambda, r)$ and the new sequence R_S and the corresponding index id are recorded. The maximum and minimum R_S

are denoted as Ma and Mi , respectively. By building a new mapping proportion coefficient $Pe(i)$, the enhanced signal $P_{new_sp}(\lambda, r)$ is obtained. Obtain slope of baseline 1, and obtain baseline 2 based on this slope. Signals exceeding baseline 2 are regarded as candidate cloud signals as shown in Fig. 3b) and Fig. 4b). The third part is to extract cloud signal and realize boundary detection by combining the SNR of echo signal. By fitting the echo signal slope in the height range of 15–20 km, the slope is used as the bottom slope to distinguish the cloud and aerosol layers (as shown by the magenta line in Fig. 3b and Fig. 4b). Without considering the bottom echo signal (0–2 km), the amplitude of the echo signal received by the lidar decreased with increasing detection height according to the fitted slope, as shown by the blue line baseline in Figs. 3b) and 4b). When the beam senses the presence of clouds, the amplitude of the echo signal will exceed the blue baseline.

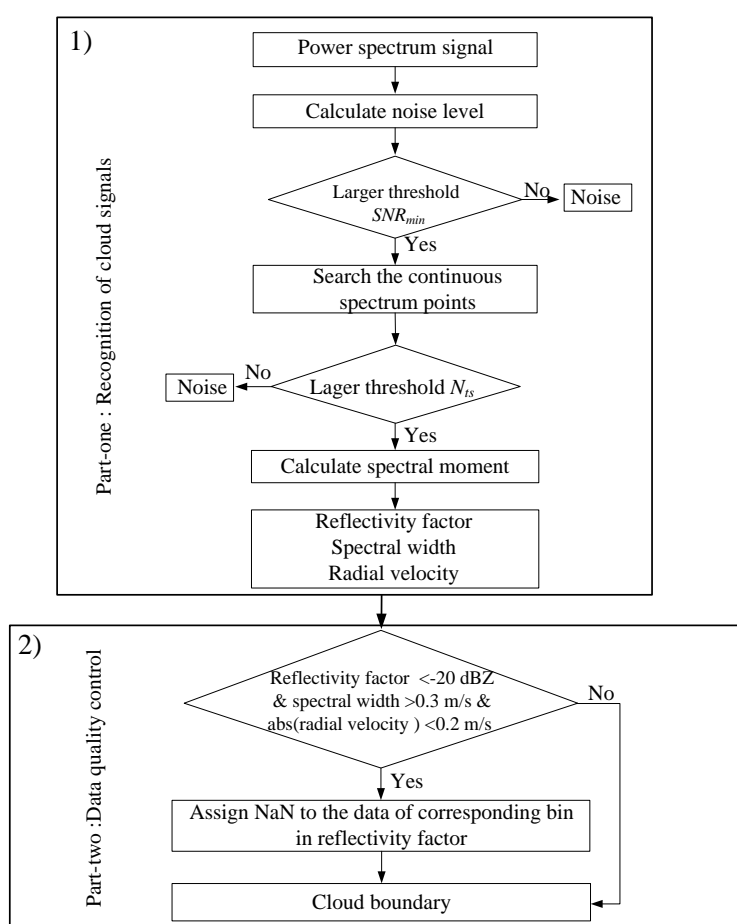


Fig. 6 Flow chart of MMCR cloud boundary detection. 1) recognition of cloud signals from Doppler spectra of MMCR, and 2) cloud boundary with data quality control

For revised figure 6: Identifying cloud signals from Doppler spectra of the MMCR is affected by the noise level, particularly when the SNR is low. As shown in Fig. 5, if all spectral points above the noise level are integrated, it will result in a large error in the inversion of its characteristic parameters (reflectivity factor, spectral width, radial velocity, etc.). Therefore, it is necessary to carefully identify cloud signals in Doppler spectra signal. There are two parts in Figure 6 includes two parts:

recognition of cloud signals from Doppler spectra of MMCR and data quality control for MMCR. Part one is mainly to prepare for obtaining effective cloud signals. Generally, cloud signals have a certain number of continuous spectral points and SNR . With the part one of Fig.6, We use the segmental method to calculate the noise level, and take it as the noise and signal boundary (as shown is Fig. 5). If spectral data amplitude is greater than SNR_{min} , and search for consecutive velocity bins in its spectral data and record the number of bins. When the number is larger than N_{ts} , and the corresponding spectral signals is determined as an effective spectra segment. Intersections of effective spectral segment and noise and signal boundary are left and right endpoints of cloud spectral, that is, the starting and end point of the spectral moment calculation.

The echo signals of floating debris in the low-level atmosphere have the characteristics of a small reflectivity factor, low velocity, and large spectral width. To further eliminate interfering wave information, we obtained the data quality control threshold by counting the characteristic changes in planktonic echoes in the boundary layer under cloud-free conditions. As shown in 2) of Fig. 6, when the subjective echo intensity $Z < -20$ dBZ, the absolute value of the radial velocity < 0.2 m/s, and the velocity spectra width > 0.3 m/s are used as the threshold for removing non-cloud information; thus the expected data quality control requirements can be met. Cloud boundaries are detected using data quality-controlled cloud echo reflectivity factors.

3. The authors claimed several times “This study will combine the advantages of lidar and MMCR in detecting clouds”. While it seems that the results are just simply calculated from the two instruments, respectively. I was hoping some more in-depth combination, like DARDAR for the space-born radar and lidar (Delanoe and Hogan, 2008), whose method is associated with the specific radar/lidar raw observational value.

Response: Dear reviewer, thank you for these valuable references, which we have carefully read and revised the manuscript accordingly based on your suggestions. We have cited these references at the corresponding places in the manuscript. As well as, these references have provided great help for my follow-up research. Obtaining accurate cloud information from echo signals is the premise of in-depth study of cloud micro parameters and analysis of special meteorological variation characteristics. In this study, we propose new methods for cloud boundary detection by lidar and MMC, and combined with special cases to verify and apply those methods. According to this main line of study, the research contents are full.

1). In the manuscript, based on the signal characteristic lidar and MMCR. We propose a new algorithm which suitable for accurately extracting cloud information from lidar echo signals. In order to improve the detection accuracy of MMCR, the cloud signals in Doppler spectra are identified in detail. The appropriate data quality control thresholds are established to effectively eliminate the floating debris echo signal.

2). Using lidar to identify cloud boundaries (cloud bottom and cloud top) is easily affected by aerosol and background noise. We extract the cloud signal effectively by wavelet change noise reduction, signal enhancement combined with the SNR of lidar echo signals. This method is not easy to be affected by noise and interfering signals, and also avoids the problem that cloud base and cloud top being overestimated or underestimated due to improper threshold selection. Compared with the previous research literature that directly uses the reflectivity of MMCR for cloud boundary recognition, the manuscript analyzes and calculates the noise level, SNR_{min} , and continuous common points from the initial Doppler spectra data of MMCR. Those make the recognition of meteorological signals more accurate.

3). Based on three special cases, we verified the proposed algorithm, and also clarified the detection advantages of lidar and MMCR under different conditions. Based on three special detection cases, the correctness and reliability of the proposed algorithm are verified, and the detection advantages of lidar and MMCR under different conditions are illustrated.

4). By processing and analyzing the accumulated observation data, a preliminary analysis of the changing characteristics of the cloud boundary is carried out in Xi'an.

So, it is unlikely that more research content needs to be added to the manuscript at present.

4. One-year observation might be too short for statistics analysis of cloud in section 4.2, especially only 302 days of MMCR and 126 days of lidar.

Response: At present, the amount of lidar and MMCR data in the manuscript is not enough to comprehensively and deeply analyze the cloud change distribution characteristics in Xi'an. Therefore, we have replaced or deleted 'statistics' in the text, and re-determined that the purpose of the manuscript is cloud boundary detection method research. The data analysis in the section 4.2 is the application of cloud boundary detection method. It provides a preliminary analysis for the distribution characteristics of cloud boundary in Xi'an. We also point out the contents to be studied in the future.

5. Most of the conclusions (line 413-423) are not new. There are many studies using collocated radar and lidar observation for cloud research, e.g. (Borg et al., 2011) (Dong et al., 2010) (Protat et al., 2011) and so on, which have shown similar results.

Response: L413-423 expresses some well-known advantages and disadvantages of lidar and MMCR for investigation of cloud, which makes the conclusions not detailed and in-depth. Therefore, we re-describe the conclusions (L 404 - 423), and also the points that can be improved in the follow-up of the manuscript are list.

Based on the observation data of lidar, a new algorithm is proposed which can effectively extract cloud signals. Compared with the previous method of identifying cloud bottom and cloud top from echo signals, the new method mainly obtains effective cloud signals through suppressing noise

signals and enhancing effective signals to realize cloud boundaries. The algorithm has two main characteristics: 1) in the signal preprocessing, wavelet transform is used for the original signal to avoid the defect of effective information loss caused by improper selection of smooth window; 2) The SNR of the signal is considered.

The cloud signals in Doppler spectra are effectively extracted by analyzing the noise level, SNR_{min} , and the continuous spectral points of Doppler spectra. The data quality control conditions for MMCR (reflectivity factor < -20 dBZ, spectra width >0.3 m/s and radial velocity < 0.2 m/s) were established by analyzing the characteristic of the interference of floating debris signals. By analysing the correlation of cloud bottom height between MMCR and lidar, and the cloud bottom height detection by MMCR with data quality control have a good agreement with lidar (the correlation coefficient is 0.803). Therefore, quality control is an important factor to improve signal accuracy of MMCR.

In this study, combined with the respective advantages of MMCR and lidar in cloud detection, the cloud cover and distribution of cloud boundaries characteristics are analyzed based on the observation data in Xi'an from December 2020 to November 2021. The result reveals that more than 34% of the clouds appear in the form of a single layer every month. The cloud cover was lowest in spring and highest in summer. The seasonal variation in cloud boundary height showed that the distribution characteristics of cloud boundaries in spring and summer were similar, and the frequency of high-level clouds in the range of 8–10 km was greater than autumn and winter. The stratiform clouds appearing below 3.5 km in autumn have the highest frequency, and high-level ice clouds or cirrus clouds above 8 km in winter are less likely to appear. The findings can provide a preliminary analysis of cloud boundary changes in Xi'an. If there are huge amounts of simultaneous observation data of lidar and MMCR, the comprehensive statistics and analysis of cloud macro and micro parameters can be realized, which can provide better support for the study of climate change characteristics in Xi'an.

Minor Comments:

1. Line 9, “he” should be “the”

Answer: Write error has been changed to “the”.

2. Line 11, The SNR and SNR_{min} in the abstract should be explained and given the full description.

Answer: The *SNR* (Signal-to-noise ratio) is the ratio of lidar echo signal to noise signal, dimensionless. The *SNR_{min}* refers to the noise ratio of the smallest measurable cloud signal in Doppler spectra signal. These have been described in the abstract.

3. Line 14, what does the “rules” mean?

Answer: The “rules” originally expressed the records of observation data in Table 3. We have changed Table 3 to “Cloud bottom height recording guideline”. In L13-15 “Based on the advantages and disadvantages of the two devices in detecting cloud boundaries under different conditions, cloud boundary statistical rules are established to analyze the characteristics of cloud boundary changes in Xi'an in 2021” is changed to “Based on the respective advantages of the two devices, the change characteristics of cloud boundary in Xi'an from December 2020 to November 2021 are analyzed with MMCR detection data as the main data and lidar data as assistant data.”

4. Line 33, what is “high change rate”

Answer: “However, the vertical structure distribution of clouds has great temporal and spatial heterogeneity and a high change rate, which leads to great challenges....” is changed to “However, the vertical structure distribution of clouds has great temporal and spatial heterogeneity, which leads to great challenges....”

5. Line 35-36, remove “direction”, ...has always been important for cloud physics.

Answer: “Notwithstanding, research on the characteristics of cloud vertical structures has always been an important direction of cloud physics research. ” is changed to “Notwithstanding, research on the characteristics of cloud vertical structures has always been an important for cloud physics.”

6. Line 50,” dP/dr ”, what is P and r , what is “negative to positive”, you mean the value of dP/dr , from negative to positive? Please rephrase this sentence.

Answer: Re-describe L50 as, “Calculation of dP/dr using lidar backscattering intensity P and range r , and the first derivative of backscatter intensity dP/dr changes sign from negative to positive and this zero crossing is cloud bottom. ”

7. Line 54, what is “detail debugging”

Answer: The 'detail debugging' means that the threshold method needs to be changed according to experience in the calculation process. Unclear expression has been modified in the manuscript.

Line 54, “It is easily affected by noise, and some indicators must be introduced in the specific implementation process to determine the cloud boundary through complex detail debugging, which brings certain difficulties to accurate cloud boundary detection” is changed to “It is easily affected by noise, and restrictive parameters must be introduced in the specific implementation process to determine the cloud boundary by adjusting the parameters, which brings certain difficulties to accurate cloud boundary detection”

8. Line 59, “but the cloud bottom and cloud top detected by this method will be overestimated and underestimated respectively”. Does this mean the method would miss some part of cloud, i.e., detect some real cloud signal as noise? This manuscript really needs complete English editing.

Answer: “but the cloud bottom and cloud top detected by this method will be overestimated and underestimated respectively.” means that the real signal at the cloud bottom may be considered as noise, and the real signal at the cloud top may be miss. It has been re-expressed as “, but this method takes some real signals at the cloud bottom as noise and miss information at the cloud top, and resulting in overestimation and underestimated of cloud base and cloud top height respectively.”

The English language of the manuscript has been polished and modified by professional institutions.

9. Line 65, what is “library”. Line 65-67 is different to understand.

Answer: The “distance library” in the line 65 is changed to “gate”.

Line 65-67 is re described as: “Kollias et al. (2007) judge step by step from the bottom to the top of the reflectivity. If the *SNR* of 9 consecutive distance gates is greater than the set threshold, these gates represented as cloud signals. Otherwise, they are non-cloud signal.”

10. Figure 1. The area could be lager, at least shows some of the “Guanzhong Basin”, “Weihe River Basin”, “Loess Plateau”, “Qinling Mountains” as you described in line 88-90. What does the white line mean? Is it really necessary to show the negative elevation in your color bar?

Answer: The white line in the original diagram originally denotes the Xi'an Region. The revised Fig. 1 contains the “Guanzhong plain”, “Weihe River”, “Loess Plateau” and “Qinling Mountains” as follow. In the Fig. 1, the black line represents Shaanxi Province, the dark blue represents the Yellow River, and the wathet blue represents the Weihe River.

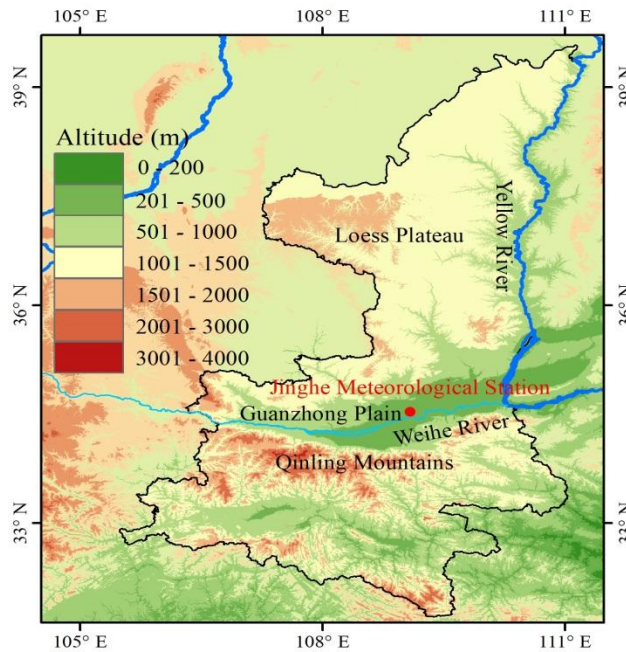


Fig. 1. Geographical coverage of Shaanxi Province (105°29'-111°15'E, 31°42'-39°35'N). The red dot indicates the location of the Jinghe National Meteorological Station in Xi'an.

11. Line 99, what is “HT101”?

Answer: TH101 is the model of the MMCR

12. Line 113-114 is difficult to understand.

Answer: “When using lidar for detection, the laser beam propagates in a clear atmosphere, and the received echo power continuously decreases with increasing detection height. However, the beam into the clouds (or aerosols, etc.), the echo power increases suddenly and becomes stronger at a distance above the cloud bottom. The lidar equation owing to elastic backscattering can be written as (Motty et al., 2018),” was re-described as “The lidar equation owing to elastic backscattering (Motty et al., 2018) can be written as,”

13. Line 116 and line 120, should the N_{bcak} be N_{back} ?

Answer: “ N_{bcak} ” is changed to “ N_{back} ”

14. Figure 2. “yes” and “no” may be marked in the wrong place. They should be marked after a judgment statement, i.e., “>”, “<” or “==”, rather than equations. The symbols in the text should be explained. What is “sort”, “Pe”? What is the relationship between the three main boxes? It is hard to follow just from line 134-135.

Answer: Sorry, “yes” and “no” are misplaced in the flow chart 2. The revised figure 2 and text description are shown in the Major comments two.

15. Figure 3. The box, axis, tick should be black. The other figures in the manuscript should be changed too. Why the time in title is different with the time in figure? What is the unit of x axis? I notice there are some signal below the blue base line in figure b, especially below cloud base height, around 8 km, 6 km and 4 km. The slope is obviously different with the fitting slope. Does this influence your detection? What is the vertical dashed line in figure c?

Answer: The box, axis, tick of all figures have changed black in the manuscript. The time of the legend in the figure is correct, and “Fig. 3 Detection results of the lidar at 19:15 on March 4, 2021” is wrongly written due to negligence, and has been modified in the paper. The situation in Fig.3b) does not affect the identification of subsequent cloud boundaries, and the signals below the blue baseline (especially at 8 km, 6 km and 4km) are considered as aerosol signals or interference information and will be eliminated. Fig.3c) S/N in Shannon formula is the power ratio of signal to noise, which is a dimensionless unit. The blue vertical dotted line is only a schematic auxiliary line in Fig.3c), indicating that the S/N of the cloud should be greater than 5 in this case.

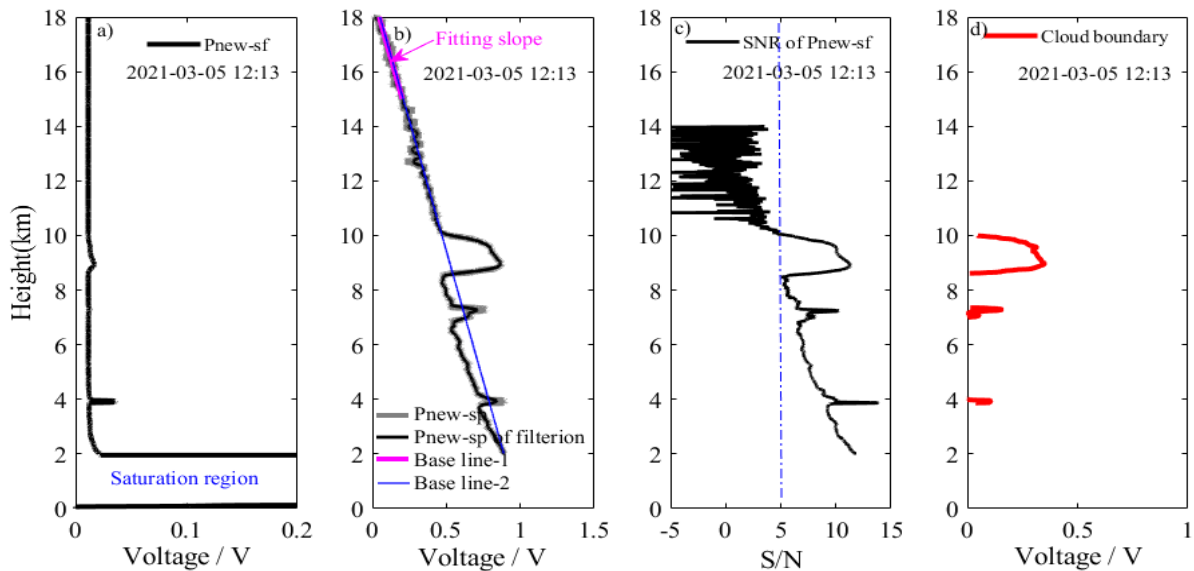


Fig. 3 Detection results of the lidar at 12:13 on March 5, 2021: a) P_{new_sf} of the 1064 nm signal, b) P_{new_sp} of the 1064 nm signal, c) SNR of P_{new_sf} , and d) cloud information detected

16. Figure 6. The “thresh of XXX” should be “Larger/Smaller than thresh of XX”. Generally, it should be a judgment statement.

Answer: Figure 6 has been modified. The revised figure 6 and text description are shown in the Major comments two.

17. Line 176, what is N_{ts} ?

Answer: N_{ts} represents the threshold value of continuous spectral points. The “ N_{ts} ” has been described in the manuscript.

18. Line 184-185, please do not use both “>” and “less than” in one sentence. What is the unit of “velocity” and “velocity spectrum width”? And why you choose such thresholds?

Answer: “As shown in Fig. 6b), when the subjective echo intensity $Z < -20$ dBZ, the absolute value of radial velocity is less than 0.2, and the velocity spectrum width > 0.3 is used as the threshold for removing nonmeteorological information, the expected data quality control requirements can be met.” is changed to “As shown in Fig. 6b), when reflectivity $Z < -20$ dBZ, the absolute value of radial velocity < 0.2 m/s, and the velocity spectra width > 0.3 m/s are used as the threshold for removing non-cloud information, the expected data quality control requirements can be met.”

19. Figure 7. Is the unit of velocity spectrum width in figure c “m/s”? Figure a, echo emissivity factor is the same as “reflectivity factor”?

Answer: Yes, the unit of velocity spectra width is m/s in Figs 7. c). “Figure 7 a), echo emissivity factor” and “reflectivity factor” in Figs 7. a) and d) are consistent, and they are uniformly expressed as ‘reflectivity’ in the manuscript.

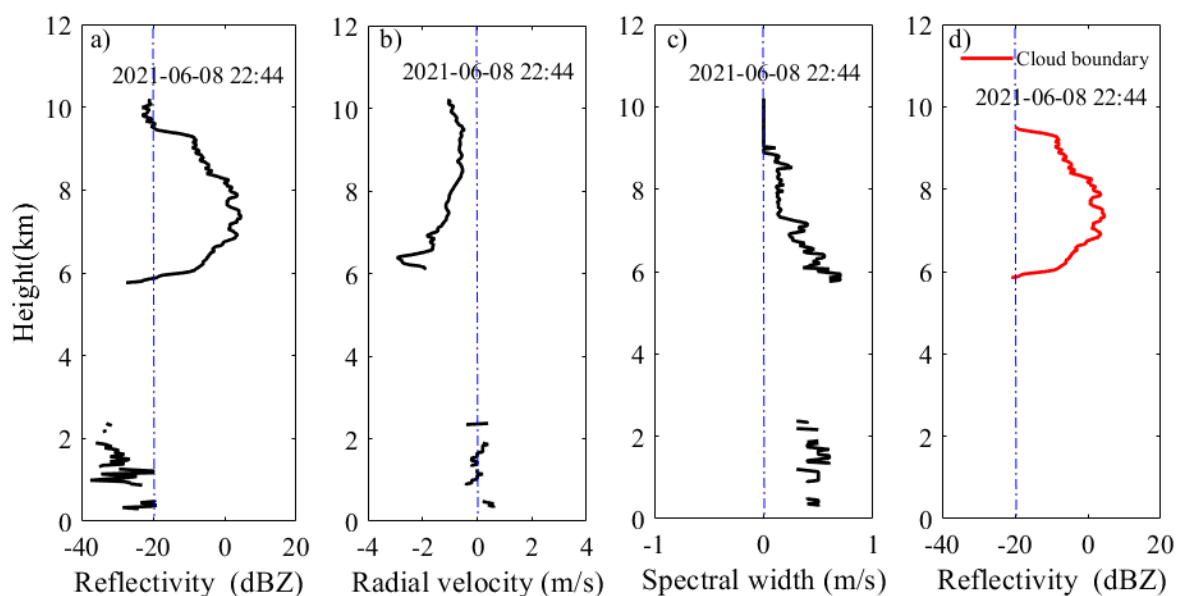


Fig. 7 Meteorological signals of MMCR at 22:44 on June 8, 2021. a) reflectivity, b) radial velocity, c) velocity spectra width, d) echo emissivity factor after quality control

20. Line 212, What is “time-height-indicator information”? Do you mean “vertical profile”?

Answer: No, “time height indicator information” is used to describe the long-term observation results in Fig. 8.” The sentence is re-described in the manuscript.

“According to the data method described in Section 3.1, the SNR of P_{new_sf} and P_{new_sp} of the echo signal of the lidar @1064 nm are obtained time-height-indicator information (THI) and are shown in Figs. 8a) and 8b).” is changed to “According to the data method described in Section 3.1, we can obtain cloud change information of time-height-indicator (THI) for SNR of P_{new_sf} and P_{new_sp} of lidar @1064nm with a duration of 7 hours, as shown in Figs. 8a) and 8b).”

21. Line 214-215, “After 05:00, the cloud layer developed deeper”. Does this infer from Figure 9, the MMCR observation? It would be clearer if you combine Figure 8, 9 and 10 together to see the difference of the two instruments. Same as Fig 11-13, and Fig 14, 15.

Answer: Yes, this phenomenon can be seen from Figure 9 that the clouds are developing deeply. We have combined figure 8 and Figure 9 in the original text and described them again. It can be seen from Fig.8 d) that the cloud layer developed deeper after 5:00, and the laser beam penetrated 0.1 km into the cloud layer and was quickly attenuated.

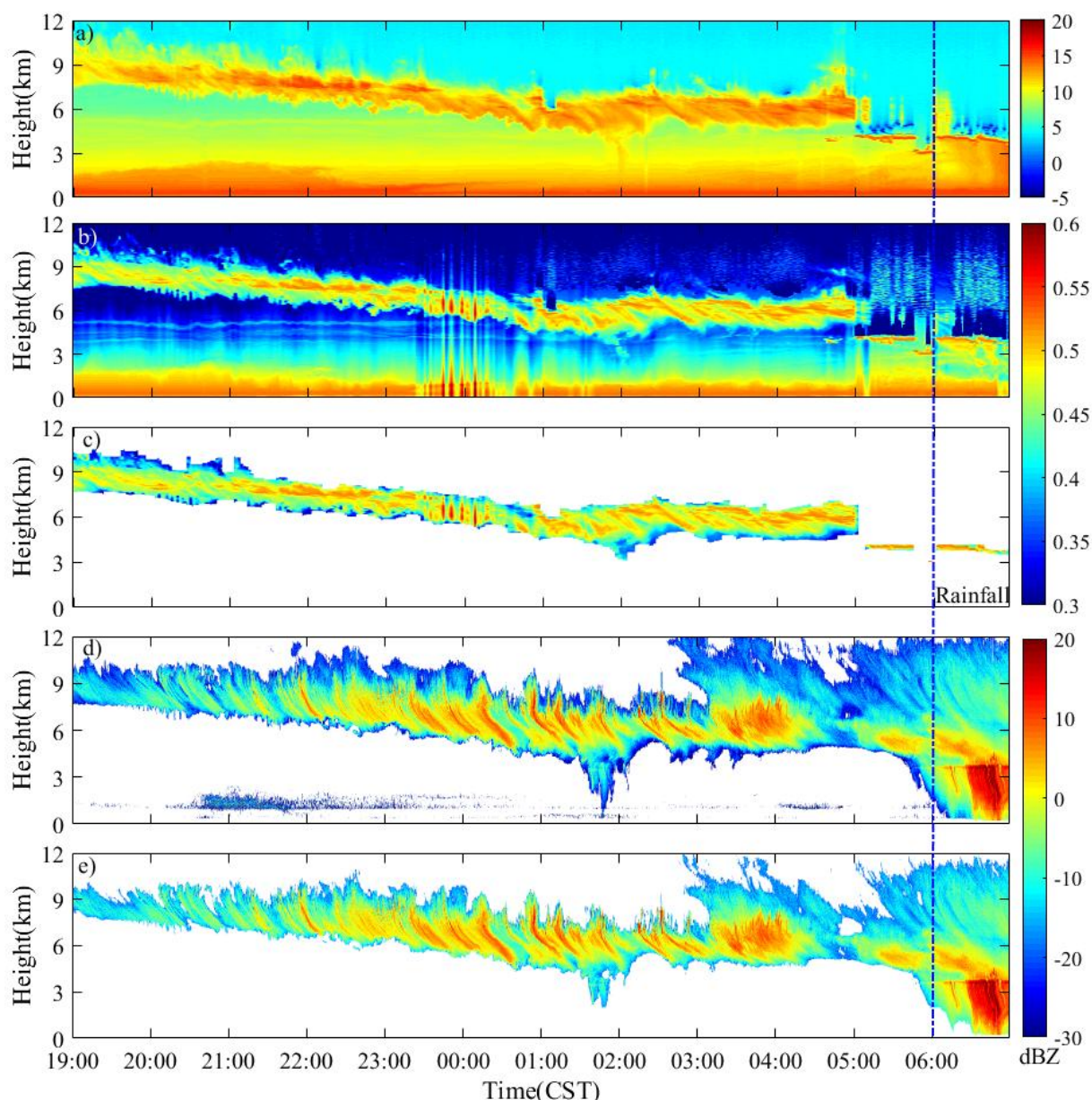


Fig. 8 The THI of echo signal of the lidar and MMCR on March 4 to 5, 2021. a) SNR of the 1064 nm signal, b) P_{new_sp} of the 1064 nm signal, c) cloud information detection results of the lidar, d) reflectivity of the MMCR without quality control, e) reflectivity the MMCR with quality control (dotted line indicates rainfall time)

22. Line 216, “Rainfall begins at 06:00”, how do you get the time of rainfall, do you have rain gauge or other observations? Please explain this in Section 2.

Answer: We checked the time of rainfall recorded by microwave radiometer, which is close to MMCR. The record of rainfall time has been described in the manuscript.

23. Figure 8. What is the stripe in figure b around 23:00-01:00? Does this affect your detection results? What does the “SNR>5.2” in figure c stand for?

Answer: The stripes around 23:00-01:00 in Fig. 8b are caused by the instability of the laser seed, which causes slight fluctuations in the emitted light energy, but this does not affect lidar detection of clouds, nor does it affect the recognition of cloud boundary. “SNR >5.2” in Fig.8 c) indicates that we get the cloud boundary shown in Fig. 8c), we only retain the effective data lattice with SNR >5.2 (regardless of the underlying signal saturation region) in Fig. 8a).

25. Line 232, “the cloud layer starts at 03:00”, does this mean the signal before 03:00 is not cloud?

Answer: No. The information displayed is cloud signal from 19:00 to 06:00 in Fig.9b). “From the THI of the echo reflectivity of the cloud, the cloud layer starts at 03:00 and gradually develops from 7 km to 12 km (the lidar signal fails to show this detail).” is changed to “According to the echo emissivity factor of the MMCR, from 03:00 to the end of observation, the cloud layer developed deeper, the cloud bottom height gradually decreased from 7 km to 300m, and the cloud top height developed to ~12 km (the lidar signal fails to show this detail). ”

26. Line 253-254, “From the characteristic distribution of the $P_{\text{new_sp}}$ signal in Fig. 11b), the low-level cloud rained from 18:30 to 18:45”, how does this be concluded, just by the sudden decrease of cloud base?

Answer: In the observation experiment at 18:30 on March 4, 2021, we felt that there were small showers on the ground and the duration was ~10 mins. Then we checked the rainfall time recorded by microwave radiometer (recording every 2 min), and the specific rainfall period was 18:30~18:45 CST and the precipitation reached the ground. At the same time, the radial velocity of MMCR showed that the velocity reaches ~-4m/s in this period.

27. Line 273, “During the period from 15:00 to 01:00”, where is “15:00” in figure 12?

Answer: Sorry, I mistakenly wrote 17:00 as 15:00 due to negligence. “During the period from 15:00 to 01:00...” is changed to “During the period from 17:00 to 01:00...”

28. Figure 13, Could you please at least use one specific color/line style/marker to represent one property (cloud base or top/first or second layer/lidar or MMCR)?

Answer: The changed Figure 13 as below.

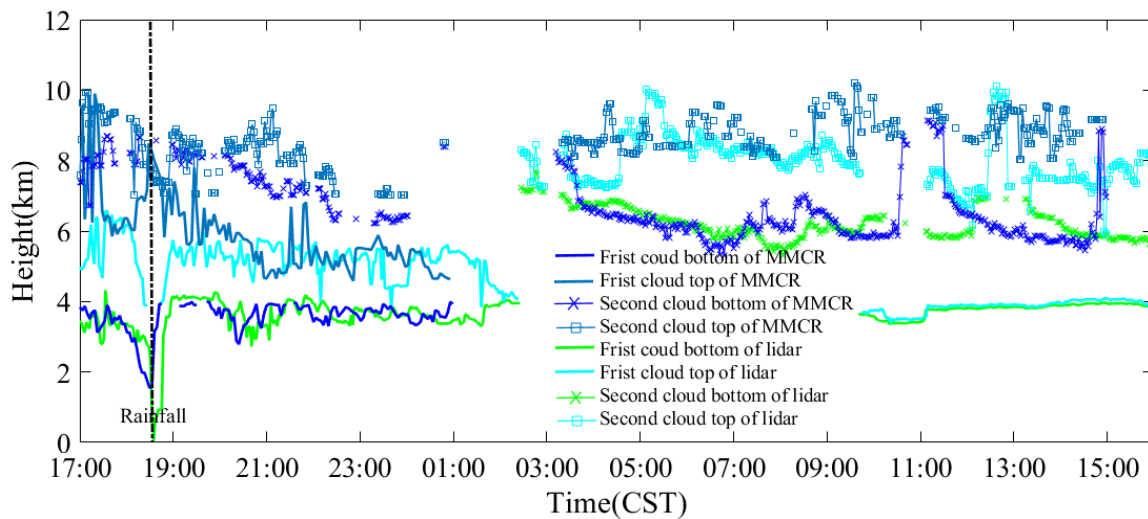


Fig. 13 Cloud boundary detected by the lidar and MMCR from March 4 to 5, 2021

29. Line 294, “Case three studies of precipitating cloud”, the figures of case one and two are also have been marked with rainfall. If you want to discuss precipitating cloud separately, the case one and two should be non-precipitating cloud.

Answer: We changed the objectives of the three study cases to the following,

“1) Case one studies of double-layer clouds” is changed to “1) First case study period”.

“2) Case two studies of double-layer clouds” is changed to “2) Second case study period”.

“3) Case three studies of precipitating cloud” is changed to “3) Third case study period”.

30. Line 310, what is “rain storage”?

Answer: The “rain storage” means “rain virga”. “As the observation time progresses, the phenomenon of rain storage (reflectivity >-15 dBZ) occurs in the cloud” is changed to “As the observation time progresses, the phenomenon of rain virga (reflectivity >-15 dBZ) occurs in the cloud”

31. Figure 15. How the cloud base height being determined for precipitating cloud, such as after 11:00? I don’t think the cloud base height around 0 km is appropriate. This may explain why the cloud base height in figure 19 has a such huge peak at lower level.

Answer: When rainfall is slightly more intense, neither laser radar nor millimeter radar achieves an accurate assessment of cloud base height (visiting a balloon perhaps achieves an approximate detection but is not part of this paper's research content). Figure11: the cloud base height after 11:00 is 0.27 km instead of 0 m. In the Fig. 15, because there are a larger number of plotted points, the cloud bottom height around 0 km is appropriate.

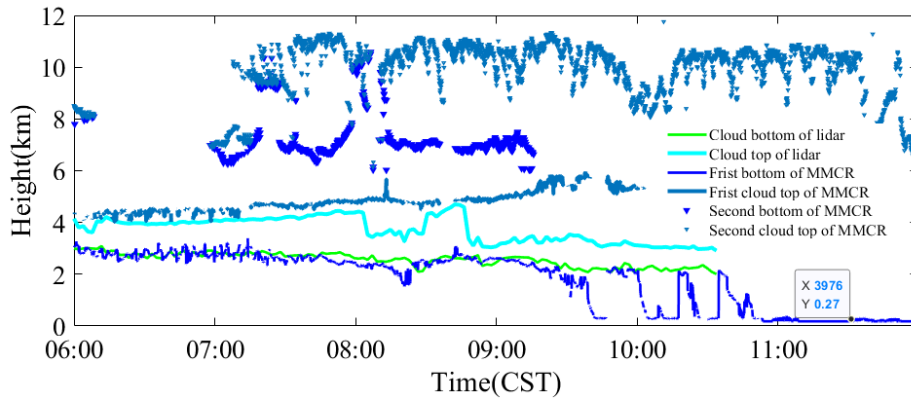


Fig. 15 Cloud boundary detected by the lidar and MMCR on March 10, 2021

32. Line 338, the 126 days of lidar observations seems too short for one year. Can the authors explain why is that? Is there any issue of the lidar, if so, does this issue affect the observed results?

Answer: The main task of the lidar in the manuscript is to monitor special weather changes, so the data volume is only 126 days in 2021. This does not affect our cloud boundary analysis for the whole year, because MMCR data are mainly used in cloud boundary analysis.

33. Line 341-342, “we plan to employ MMCR data to replace the data of periods when the lidar is not running” What do you mean by “replace”? You mean the MMCR data are only useful when lidar is not running? Generally, I am not sure the purpose of Figure 16 and Table 3. “bottom of MMCR is blurred” in Table 3, what does this mean? Are the results of table 3 accomplished by manual selection?

Answer: According to the results discussed in the previous chapters, lidar has more advantages than MMCR in cloud bottom detection. Therefore, lidar (detecting cloud bottom) and MMCR (detecting cloud top) can be combined to detect cloud boundary (cloud bottom and cloud top), but considering the continuous observation time of lidar, it is not enough to analyze the change of cloud bottom all the year. Therefore, we analyzed the correlation between the cloud bottom detected by MMCR with quality control and lidar, and the correlation coefficient is 0.803. Therefore, the cloud bottom height during the period when the lidar is not running is provided by MMCR to realize the annual cloud boundary change in Xi'an.

Figure 16 mainly shows that the cloud bottom height is good agreement with the lidar and MMCR with data quality control. Therefore, when the lidar is not operational, the cloud bottom information can be provided by MMCR.

The “bottom of MMCR is blurred” in Table 3 indicates that the MMCR cannot accurately identify the cloud bottom in light rain or drizzle. “Bottom of MMCR is blurred” is changed to “bottom of MMCR is invalid”.

The data selection in Table 3 is provided by our developed algorithm.

34. Line 379, “20217” should be “2017”.

Answer: Sorry, “20217” has been changed to “2017”.

35. Line 385-386, “The months with the largest (96%) and smallest (42%) cloud occurrence frequencies are August and December, respectively.” Does this mean the Jinghe National Meteorological Station are nearly covered by cloud during the whole month of August? Does it make any sense?

Answer: “The months with the largest (96%) and smallest (42%) cloud occurrence frequencies are August and December, respectively.” indicates that 96% and 42% of all profiles detected in the 22 days of August and 30 days of December contain cloud profiles, indicating that the frequency of cloud formation is the highest and lowest in August and December respectively. This number of ‘96%’ is relatively large, and Line 1389 explains why “96%” is large.

36. Line 390-391 and figure 18b, how the “normalized monthly distribution” be calculated? “the minimum cloud amount is 0.65 in spring and the maximum is 2.46 in summer”, how do these two numbers be inferred?

Answer: MMCR defines cloud cover as the percentage of cloud obscuring sky field of vision. Cloud cover observation includes total, low, medium and high cloud cover. Total cloud cover refers to the total number of cloud cover in the sky during observation (Fig.18b shows the total cloud cover in every month). Generally, the sky is divided into 10 parts. When there is no cloud in the clear sky or less than 0.5 parts are covered, the cloud cover is zero. The cloud covers half of the sky and the cloud cover is 5. Cover the whole sky with clouds and the cloud cover is 10. Calculation steps: 1): divide the cloud layer into high, medium and low families through the radial cloud base height. 2): average each cluster for 30 minutes. 3): Weighted Processing of data in 10 minutes to obtain the integrated cloud cover. Because the calculated cloud cover is a relative value, it does not mean the real cloud cover. Figure 18b shows that the cloud cover is the largest in April. Therefore, the cloud cover in April is set to 1, and the cloud cover in other months is calculated to represent the relative change trend of cloud cover in each month.

‘the minimum cloud amount is 0.65 in spring and the maximum is 2.46 in summer’ is changed to ‘It can be seen from the distribution of cloud cover in every month that there are relatively more cloud cover in summer and the least in winter, indicating that warm atmospheric conditions are more conducive to the formation and development of clouds.’

Reference:

Borg, L. A., Holz, R. E., and Turner, D. D.: Investigating cloud radar sensitivity to optically thin cirrus using collocated Raman lidar observations, *Geophysical Research Letters*, 38, L05807, 10.1029/2010gl046365, 2011.

Delanoe, J., and Hogan, R. J.: A variational scheme for retrieving ice cloud properties from combined radar, lidar, and infrared radiometer, *Journal of Geophysical Research-Atmospheres*, 113, 10.1029/2007jd009000, 2008.

Dong, X., Xi, B., Crosby, K., Long, C. N., Stone, R. S., and Shupe, M. D.: A 10 year climatology of Arctic cloud fraction and radiative forcing at Barrow, Alaska, *Journal of Geophysical Research*, 115, 10.1029/2009jd013489, 2010.

Protat, A., Delanoe, J., May, P. T., Haynes, J., Jakob, C., O'Connor, E., Pope, M., and Wheeler, M. C.: The variability of tropical ice cloud properties as a function of the large-scale context from ground-based radar-lidar observations over Darwin, Australia, *Atmospheric Chemistry and Physics*, 11, 8363-8384, 10.5194/acp-11-8363-2011, 2011.

We have carefully revised the manuscript base on the opinions of reviewers and public discussion. The modified part has been marked with blue font. The specific modification list is as follows:

Title and authors

Number	original manuscript	Number of line	revised manuscript	Number of line
1	Lidar and MMCR applied for the study on cloud boundary detection and the statistical analysis of cloud distribution in Xi'an region	line 1-2	Detection and analysis of cloud boundary in Xi'an, China employing 35GHz cloud radar aided by 1064nm lidar	line 1-2
2	Yun Yuan , Huige Di *, Tao Yang , Yuanyuan Liu , Qimeng Li, Qing Yan, Dengxin Hua*	line 3-4	Yun Yuan , Huige Di *, Yuanyuan Liu ,Tao Yang , Qimeng Li, Qing Yan, Wenhui Xin, Shichun Li, Dengxin Hua*	line 3-4

0) Modification the part of Abstract

Number	original manuscript	Number of line	revised manuscript	Number of line
1	Lidar	line 6	Lidar @1064 nm	line 7
2	Equipment to detect	line 6	tools for detecting	line 7
3	which can monitor the whole life	line 7	and can monitor the entire life	line 8
4	In this paper, we employ lidar	line 7	In this study, lidar and	line 8
5	under different conditions (e.g., single-layer clouds, multilayer clouds, and precipitating clouds)	line 9-8	under different conditions	line 9
6	SNR	line 9	Signal-to-noise ratio (SNR)	line 10
7	using SNR_{min}	line 11	using the noise ratio of the smallest measurable cloud signal (SNR_{min})	line 12
8	threshold, and the quality control of the meteorological signal (echo reflectivity factor)	line 11-12	threshold (N_{th}). Moreover, the quality control of the reflectivity factor of MMCR	line 13
9	Based on the advantages and disadvantages of the two devices in detecting cloud boundaries under different conditions, cloud boundary statistical rules are established to analyze the characteristics of cloud boundary changes in Xi'an in 2021. The seasonal variation characteristics of clouds show that the frequency distribution of cloud boundaries in vertical height in spring and summer has a similar variation trend. The normalized cloud amount is the lowest in spring (0.65) and the highest in summer (2.46). The frequency distribution of high-level clouds (at 11~12 km) is the highest in autumn, and the clouds in winter are mainly distributed below 8 km. Furthermore, the cloud boundary frequency distribution results for the whole year of 2021 show that the cloud bottom boundary below 1.5 km is more than 10%, the frequency within the height range of 3.06 km~3.6 km is approximately 3.24%, and the frequency above 8 km is less	line 13-23	We analyzed three typical cases studies; case one presents two interesting phenomena: a) at 19:00–20:00 CST (China standard time), the ice crystal particles at the cloud top boundary are too small to be detected by MMCR, but they are well detected by lidar. b) at 19:00–00:00 CST, the cirrus cloud tranists to altostratus where the cloud particles eventually grow into large sizes, producing precipitation. Further, MMCR has more advantages than lidar in detection the cloud top boundary within this period. Considering the advantages of the two devices, the change characteristics of the cloud boundary in Xi'an from December 2020 to November 2021 were analysed, with MMCR detection data as the main data and lidar data as the assistant data. The seasonal variation characteristics of clouds show that, in most cases, high clouds often occur in summer and autumn, and the low clouds are usually in winter. The normalised cloud cover shows that the maximum and minimum cloud cover occur in summer and winter, respectively. Furthermore, the cloud boundary frequency distribution results for the whole of observation	line 14-28

	<p>than 2%. The cloud top boundary frequency distribution has the characteristics of a bimodal distribution. The first narrow peak lies at approximately 1.5~3.1 km, and the second peak appears at 7.5~10.5 km.</p>		<p>period show that the cloud bottom boundary below 1.5km is more than 1%, the frequency within the height range of 3.06~3.6 km is approximately 0.38%, and the frequency above 8 km is less than 0.2%. The cloud top boundary frequency distribution exhibits the characteristics of a bimodal distribution. The first narrow peak lies at approximately 1.0~3.1 km, and the second peak appears at 6.4~9.8 km.</p>	
10	Ka-band Millimeter-Wave Cloud Radar (MMCR)	line 24	Ka-band millimeter-wave cloud radar (MMCR)	line 14-18

1) Modification the part of Introduction

Number	original manuscript	Number of line	revised manuscript	Number of line
1	Stephens et al., 2012	line 29	Sherwood et al., 2014; Dong et al., 2010	line 37
2	change rate	line 33	rate of change	line 38-39
3	Notwithstanding, research on	line 35	Research on	line 40
4	cloud vertical structures	line 35	vertical cloud structures	line 40
5	of cloud physics research	line 36	in cloud physics research	line 41
6	of course including the side boundary	line 37	including the side boundary.	line 42
7	in this paper mainly refers to the cloud bottom and cloud top boundary. In the case of multilayer clouds, it also includes the boundary information of intermediate discontinuous clouds	Line38-39	in this study mainly refers to the cloud bottom and top boundaries. Multilayer clouds also include boundary information of intermediate discontinuous clouds	line 43-44
8	MMCR	line 41	Ka-band millimeter-wave cloud radar (MMCR)	line 46
9	are effective instruments for cloud boundary detection	line 43	have become effective instruments for cloud boundary detection	line 48
10	The common methods of	line 44	Common methods for	line 49
11	by lidar	line 44	using lidar	line 49
12	the amplitude of the echo signal	line 45-46	the echo signal amplitude	line 50-51
13	However, in fact, due to the existence of noise, the point with an obvious increase in amplitude may not be found under the condition of a low signal-to-noise ratio (SNR), so the cloud bottom boundary cannot be judged. The differential zero-crossing method proposed by Pal et al. (Pal et al.,1992) differentiates the echo signal to obtain dP/dr , and the zero crossing point from negative to positive is the cloud bottom boundary. The threshold method, differential zero crossing method and variant detection method are all based on feature points of cloud boundaries (Streicher et al., 1995). It is easily affected by noise, and some indicators must be introduced in the specific implementation	line 47-54	However, because of the existence of noise, a point with a marked increase in amplitude may not be found under the condition of a low signal-to-noise ratio (SNR); therefore, the cloud bottom boundary cannot be judged. Pal et al. (1992) proposed the differential zero-crossing method through Calculation of dP/dr using lidar backscattering intensity P and range r , and the first derivative of backscatter intensity dP/dr changes sign from negative to positive and this zero crossing is cloud bottom. The threshold, differential zero-crossing, and variant detection methods are all based on the feature points of cloud boundaries (Streicher et al.,	line 52-60

	process to determine the cloud boundary through complex detail debugging, which brings certain difficulties to accurate cloud boundary detection.		1995). They are easily affected by noise, and some indicators must be introduced in the specific implementation process to determine the cloud boundary by changing the experience threshold frequently during calculation, which causes difficulties in accurate cloud boundary detection.	
14	, but the algorithm	line 56	However, the algorithm	line 62
15	WCT (wavelet covariance transform)	line 257	wavelet covariance transform method,	line 63
16	Morille et al. (Morille et al., 2007)	line 58	Morille et al. (2007)	line 63
17	detected by this method will be overestimated and underestimated, respectively.	line 59-60	but this method takes some real signals at the cloud bottom as noise and miss information at the cloud top, and resulting in overestimation and underestimated of cloud base and cloud top height respectively. Mao (2011)	line 64-66
18	, and realized the	line 61	, and detected the	line 67
19	to detect the cloud boundary (Haper et al., 1966; Hobbs et al.,1985; Platt et al., 1994; Brown et al., 1995). Kollias et al. (Kollias et al., 2007) judged the SNR value of a 5×5 grid centered on a distance library. If the SNR of more than 9 consecutive libraries reaches the threshold, the distance library is a cloud signal; otherwise, it is judged as a noncloud signal.	line 64-67	used to detect the cloud boundary (Hobbs et al., 1985; Platt et al., 1994). Kollias et al. (2007) judge step by step from the bottom to the top of the reflectivity. If the SNR of more than nine consecutive distance gates reaches the set threshold, these gates represented as cloud signals; otherwise, it is deemed a noncloud signal.	line 70-72
20	Due to the existence of certain ground object	line 69	The existence of certain ground object	line 74
21	in the lower atmosphere, it will interfere with the real cloud echo signal	line 70	in the lower atmosphere interferes	line 75
22	, it will result in large errors in the detection of cloud boundaries.	line 73	large errors in the detection of cloud boundaries result.	line 78-79
23	At present,	line 75	Currently,	line 81
24	Sasse et al., 2001)	line 76	Sasse et al., 2001; Borg et al., 2011; Delanoe and Hogan, 2008)	line 82-83
25	in this paper,	line 81	in this study,	line 87

2) Modification the part of Observation and Instrument

Number	original manuscript	Number of line	revised manuscript	Number of line
1	Xi'an (107.40 ~ 109.49 E and 33.42 ~ 34.45 N)	line 88	Xi'an City (107°40'-109°49'E, 33°42'-34°45'N), Shaanxi Province (105°29'-111°15'E, 31°42'-39°35'N)	line 94
2	of sky clouds. Fig. 1 shows the topography of Xi'an and the site location of the Jinghe Meteorological Station.	line 93-94	of clouds. Black line represents Shaanxi Province, dark blue represents the Yellow River, wathet blue represents the Weihe River, and red dot indicates the location of the Jinghe National Meteorological Station in Fig. 1.	line 99-101
3	Figure 1	line 95-97	Figure 1	line 102-104

3) Modification the part of Method

Number	original manuscript	Number of line	revised manuscript	Number of line
	in the actual observation process	107	during the actual observation	107
	of echo signals.	110	in the echo signals.	117
1	When using lidar for detection, the laser beam propagates in a clear atmosphere, and the received echo power continuously decreases with increasing detection height. However, the beam into the clouds (or aerosols, etc.), the echo power increases suddenly and becomes stronger at a distance above the cloud bottom. The lidar equation owing to elastic backscattering can be written as (Motty et al., 2018),	line 112-115	The lidar equation owing to elastic backscattering (Wandinger, 2005; Motty et al., 2018) can be written as,	line 119
2	Add formula		$P(\lambda, r) = P_0 \frac{c\tau}{2} A\eta \frac{O(r)}{r^2} \beta(\lambda, r) \cdot \exp\left[-2\int_0^r \sigma(\lambda, r) dr\right]_{(1)}$	line 110
3	<p>where λ is the wavelength of the emitted light, r represents the detection distance, and C is the system constant, which is determined by the laser energy, the receiving area of the telescope, the quantum efficiency of the detector, etc. Δr is the detection range resolution of the system, and $\beta(\lambda, r)$ and $\sigma(\lambda, r)$ are the atmospheric backscattering coefficient and atmospheric extinction coefficient, respectively. $N_{back}(\lambda, r)$ is the background noise received by the system. $E(\lambda, r)$ represents noise brought to the detection system obtained by calibration.</p> <p>To avoid amplifying the high-level noise signals, we do not perform the distance square correction Eq. (1) and directly process it as follows:</p>	line 17-123	<p>where λ is the wavelength of the emitted light, r represents the detection distance, and $\beta(\lambda, r)$ and $\sigma(\lambda, r)$ are the atmospheric backscattering and extinction coefficients, respectively. $O(r)$ is the laser-beam receiver field-of-view overlap function, c is the speed of light, P_0 is the average power of a single laser pulse, τ is the temporal pulse length, η is the overall system efficiency, and A is the area of the primary receiver optics responsible for the collection of backscattered light.</p> <p>Considering the influence of the background noise and response noise of the photomultiplier detector, Eq. (1) can be further expressed as</p>	line 121-127
4	$P_{new}(\lambda, r) = \frac{P(\lambda, r) - E(\lambda, r) - N_{back}(\lambda, r)}{C \cdot \Delta r}$	line 124	$P(\lambda, r) = C \cdot \frac{\Delta r}{r^2} \cdot \beta(\lambda, r) \cdot \exp\left[-2\int_0^r \sigma(\lambda, r) dr\right] + E(\lambda, r) + N_{back}(\lambda, r)$	line 128

5	For ground-based lidar, the echo signal at a certain height range (>15 km in this study) can be considered background and electrical noise, $N_{back}(\lambda, r)$ can be estimated with the signal within this range, and the standard deviation of the noise within the distance range is calculated:	line 125-127	where C is the system constant, which is determined by the laser energy, receiving area of the telescope, and quantum efficiency of the detector. Δr is the detection range resolution of the system. $N_{back}(\lambda, r')$ is the background noise received by the system. $E(\lambda, r)$ represents the noise introduced to the detection system by calibration. To avoid amplifying the high-level noise signals, we do not perform distance square correction in Eq. (2) but directly process it as follows:	line 129-131
6	$P_{new}(\lambda, r) = \frac{P(\lambda, r) - E(\lambda, r) - N_{back}(\lambda, r'')}{C \cdot \Delta r}$	line 124	$P_{new}(\lambda, r) = \frac{P(\lambda, r) - E(\lambda, r) - N_{back}(\lambda, r')}{C \cdot \Delta r}$	line 134
7	(>15 km in this study) can be considered background and electrical noise	line 125-126	(>15 km in this study applied to the Xi'an region) can be considered as molecular scattering	line 135-136
8	is calculated:	line 127	is calculated as follows:	Line137
9	where x is	line 129	where x denote	line 139
10	we set $k=4$ in this paper. Usually, the moving average of $P_{new}(\lambda, r)$ is performed to reduce the influence of random noise. However, the selection of the sliding window directly affects the quality of the signal. Therefore, in this paper, we use the soft-threshold wavelet denoising method to process $P_{new}(\lambda, r)$ to obtain $P_{new-s}(\lambda, r)$. To avoid atmospheric turbulence and noise interference, $P_{new-s}(\lambda, r)$ is processed in one step according to the algorithm flow in Fig. 2, and the enhanced signal $P_{new-sp}(\lambda, r)$ is obtained, as shown in Fig. 3b) and Fig. 4b). The cloud signal is prominently increased from the background noise and the aerosol signal compared to Fig. 3a) and Fig.	line 131-138	we set $k = 4$ in this study. The algorithm flow chart of detecting cloud boundary by lidar is shown in Fig. 2. Usually, the moving average of $P_{new}(\lambda, r)$ of lidar echo signal is calculated to reduce the influence of random noise. However, the selection of a sliding window directly affects the signal quality. Therefore, $P_{new}(\lambda, r)$ is denoised by wavelet transform, threshold function is a soft threshold, wavelet base is sym7, and the number of decomposition layers is 5. Using wavelet function to reduce noise can avoid too much smoothing remove sharp signal changes due to clouds, and can also avoid the improper selection of moving average window. Obtaining cloud boundaries mainly includes three parts. The first part is signal preprocessing. $P_{new-s}(\lambda, r)$ after wavelet de-noising is discretized based on the estimates of noise, and get useful	line 141-155

	4a). In this paper, we consider that the echo signal above 15 km is caused by background and electrical noise.		signals $P_{new_s1}(\lambda, r)$ and $P_{new_s2}(\lambda, r)$. The second part is to enhance the signal to make the cloud signal sharper from the background noise and aerosol signal. Average signals $P_{new_s1}(\lambda, r)$ and $P_{new_s2}(\lambda, r)$ to obtain $P_{new_sf}(\lambda, r)$. Ascending arrangement are conducted for $P_{new_sf}(\lambda, r)$ and the new sequence R_S and the corresponding index id are recorded. The maximum and minimum R_S are denoted as Ma and Mi , respectively. By building a new mapping proportion coefficient $Pe(i)$, the enhanced signal $P_{new_sp}(\lambda, r)$ is obtained. Obtain slope of baseline 1, and obtain baseline 2 based on this slope. Signals exceeding baseline 2 are regarded as candidate cloud signals as shown in Fig. 3b) and Fig. 4b). The third part is to extract cloud signal and realize boundary detection by combining the SNR of echo signal	
11	the detected cloud information is shown in Figs. 3d) and 4d).	line 149	the detected cloud information is shown in Figs. 3d) and 4d). Compared with the traditional method of finding cloud bottom and cloud top from echo signals, this method first accurately extracts cloud signals, and then obtains cloud boundaries (cloud bottom and top). This method greatly reduces the interference caused by noise and aerosol signal.	line 167-170
12	Figure 2, Figure 3 and Figure 3	line 150-157	Figure 2, Figure 3 and Figure 3	line 171-179
13	As shown in Fig. 5	line 160	As shown in Fig. 5 (Di, H., et al., 2008)	line 182
14	echo reflectivity	line 161	reflectivity factor,	line 183
15	power spectrum signal. When there is a meteorological signal in the power spectrum, the general signal has a certain SNR and the number of spectral points, while the SNR of the noise is very low or the number of continuous spectral points is small, indicating that there is no meteorological signal (Zheng et al., 2014). Accordingly, by calculating the noise and signal boundary, we count the number of continuous spectrum signal points greater than the noise and signal boundary. Set the SNR threshold and the spectral point threshold to evaluate whether each continuous data point is a cloud signal. SNR_{min} refers to the SNR of the smallest measurable cloud signal in the power	line 162-171	Doppler spectra signal. There are two parts in Fig. 6 includes two parts: recognition of cloud signals from Doppler spectra of MMCR and data quality control for MMCR. Part one is mainly to prepare for obtaining effective cloud signals. Generally, cloud signals have a certain number of continuous spectral points and SNR . With the part one of Fig. 6, we use the segmental method to calculate the noise level, and take it as the noise and signal boundary (as shown is Fig. 5). If spectral data amplitude is greater than SNR_{min} , and search for consecutive velocity bins in its spectral data and record the	line 184-192

	spectrum. When the signal is greater than SNR_{min} , it is considered to have cloud signal; otherwise, there is only noise signal. Fig. 6 shows the algorithm flow chart of MMCR inversion cloud signal recognition. Referring to the empirical formula proposed by Riddle (Riddle et al., 1989), the SNR_{min} can be calculated by Eq. (6),		number of bins. When the number is larger than N_{ts} , and the corresponding spectral signals is determined as an effective spectrum segment. Intersections of effective spectral segment and noise and signal boundary are left and right endpoints of cloud spectral, that is, the starting and end point of the spectral moment calculation.	
16	and the SNR_{min} is -17.74 dB by calculating the SNR_{min} . Adjust the SNR_{min} according to the measured data of the MMCR, and finally determine the $SNR_{min} = -20$ dB. Referring to the research results of Shupe et al. (Shupe et al., 2004), N_{ts} is set to 5. When the spectral signal meets the thresholds of SNR_{min} and N_{ts} , it is considered that there is a cloud signal in the power spectrum, and cloud feature parameter calculation is performed, flow of cloud signal recognition algorithm is shown in Fig. 6a.	Line174-178	respectively, and the SNR_{min} obtained by calculating the SNR_{min} is -17.74 dB. The SNR_{min} is adjusted according to the measured data of the MMCR and SNR_{min} is finally determined as -20 dB. Based on the research results of Shupe et al. (2008), N_{ts} is set to 7.	line 195-197
17	Figure 5	line 179-180	Figure 5	line 198-199
18	bottom atmosphere	line 181	low-level atmosphere	line 200
19	small velocity	line 182	low velocity	line 201
20	when the subjective echo intensity $Z < -20$ dBZ, the absolute value of radial velocity is less than 0.2, and the velocity spectrum width > 0.3 is used as the threshold for removing nonmeteorological information, the expected data quality control requirements can be met. Cloud boundaries are detected using data quality-controlled cloud echo reflectivity factors.	line 184-187	when the reflectivity factor $Z < -20$ dBZ, the absolute value of radial velocity < 0.2 m/s, and the velocity spectrum width > 0.3 m/s are used as the threshold of noncloud information in bin. If the characteristic parameters of each bin meet the threshold, and assign NaN to the corresponding bin in reflectivity factor. The echo signals of floating debris in reflectivity factor are eliminated by the method, and the quality-controlled for reflectivity factor is realised.	line 203-207
21	Figure 6	line 188-189	Figure 6	line 208-210
22	the meteorological signals	line 191	the cloud signals	line 212
23	The nonmeteorological signals	Line192	The noncloud signals	line 213
24	The cloud signal shown in Fig. 7d) realizes the accurate detection of the cloud boundary.	line 193-194	As shown in Fig. 7d) accurate recognition of cloud boundary is realized.	line 214

4) Modification the part of Results and discussion

Number	original manuscript	Number of line	revised manuscript	Number of line
1	1) Case one studies of a single layer cloud	line 208	1) First case study period	line 227
2	the SNR of P_{new_sf} and P_{new_sp} of the echo signal of the lidar @1064 nm are obtained time-height-indicator information (THI) and are shown in Figs. 8a) and 8b).	line 211-213	we can obtain cloud change information of time-height-indicator (THI) for SNR of P_{new_sf} and P_{new_sp} of lidar @1064nm with a duration of 7 hours, as shown in Figs. 8a) and 8b)	line 230-232
3	Combined with the thresholds of SNR	line 220	Combined with the SNR ($SNR > 5.2$ without considering the low-level saturation zone)	line 240
4	Figure 9 shows the cloud echo reflectivity factor of the MMCR at the same observation time period,	line 225	Cloud reflectivity factor of the MMCR for the same observation time period,	Line242
5	is not carried out, there are obvious nonmeteorological signals in the range of 0~2 km,	Lin227-228	is not conducted, noncloud signals in the range of 0–2 km are not prominent,	line 244
6	We can effectively eliminate the nonmeteorological signals at the bottom atmosphere and the interference signals around the clouds by using data quality control for the echo reflectivity coefficient in Fig. 9b). From the THI of the echo reflectivity of the cloud, the cloud layer starts at 03:00 and gradually develops from 7 km to 12 km (the lidar signal fails to show this detail). When rain appeared at 06:00, the cloud bottom boundary detected by the MMCR became blurred	line 230-240	We can effectively eliminate the noncloud signals at the bottom atmosphere and the interference signals around the clouds using data quality control for the reflectivity factor in Fig. 8e). According to the reflectivity factor of the MMCR, from 03:00 CST to the end of observation, the cloud layer developed deeper, the cloud bottom height gradually decreased from 7 km to 300 m, and the cloud top height developed to ~12 km (the lidar signal fails to show this detail). When rain appeared at 06:00 CST (The microwave radiometer accurately records the rainfall time, similar to the following), MMCR cannot accurately detect the cloud bottom height,	line 246-252
7	Figure 8 and Figure 9	line 222, 237	Figure 8	line 255-258
8	Although lidar cannot penetrate more clouds in this period, it can obtain an effective cloud bottom boundary.	line 243-244	Although lidar cannot penetrate more clouds during this period, it can provide an effective cloud bottom boundary. At 19:00–20:00 CST, in cloud top boundaries where the ice crystals are too small to be detected by the MMCR, lidar detects the real cloud top. This is attributable to the echo intensity of the MMCR being proportional to the 6th power of the particle diameter, and the lidar echo signal is	line 262-269

			proportional to the square of the particles. From 19:00 to 00:00 CST, cirrus cloud transition to altostratus, where size of cloud particles increases in the form of collision and finally produces precipitation. In this process, the lidar beam entering the cloud is attenuated, but MMCR has a good advantage in cloud-top detection.	
9	Figure 10	line 245-246	Figure 9	line 270-271
10	2) Case two studies of double-layer clouds	line 247	2) Second case study period	line 272
11	From March 4 to 5, 2021,	line 248	From 4 to 5 March 2021,	line 273
12	Fig. 11a) and Fig. 1b). These THIs display	line 250	Figs. 10a) and 10b). These THIs reveal	line 275
13	during the observation process.	line 251	during the observation period.	line 276
14	During the period from 17:00 to 01:00, there is a relatively weak P_{new_sp} signal	line 258	From 17:00 to 01:00 CST, there was a relatively weak P_{new_sp} signal	line 283
15	the echo reflectivity of MMCR	line 269	the reflectivity factor of the MMCR	line 295
16	From the joint observation results	line 271	The joint observation results	line 297
17	During the period from 15:00 to 01:00,	line 273	From 17:00 to 01:00 CST	line 299
18	is obviously better than	Line274	was markedly better than	line 299
19	Figure 11 and Figure 12	line 266, 284	Figure 10	line 290-292
20	Based on the cloud signals (Fig. 11c and Fig. 12b) jointly observed by the lidar and MMCR, the height distribution of the double-layer cloud boundaries is detected, as shown in Fig. 13. From the cloud boundary height distribution, it can be seen	line 286-288	The height distribution of the double-layer cloud boundaries was detected based on the cloud signals (Fig. 10c and Fig. 10e) jointly observed by lidar and MMCR, as shown in Fig. 11. The cloud boundary height distribution shows	line 309-310
21	lidar has total supremacy in detecting the information of thin clouds.	line 291	lidar is superior in detecting thin cloud information.	line 214
22	Figure 13	line 292-293	Figure 11	line 315-316
23	On March 10, 2021,	line 295	On 10 March 2021	line 318
24	the echo reflections of	line 299	reflectivity factor of	line 322
25	which makes	line 301	which simplifies	line 324
26	Figure 14	line 303-306	Figure 12	line 339-342
27	rain storage (>15 dBZ)	line 310	rain virga (> -15 dBZ)	line 329
28	while the millimeter	line 311	whereas the millimeter	line 330
29	There was a drizzle falling from 09:00 to 10:45,	line 317	A drizzle fell from 09:00 to 10:45 CST,	line 336

30	Figure 15	line 30-321	Figure 13	line 343-344
31	the beam of the lidar will be seriously attenuate	line 329	the more severely the beam of the	line 352
32	the complete cloud information at this time.	line 331	complete cloud informatio	line 354
33	4.2 Statistics and analysis of cloud boundary distribution characteristics in Xi'an	line 336	4.2 Analysis of cloud boundary distribution characteristics in Xi'an	line 359
34	In 2021, the Lidar and MMCR radar conducted cloud observation experiments at the Jinghe meteorological station, in which the MMCR accumulated 302 days of data (7248 hours in total) and the lidar observed 126 days (872.5 hours in total). Due to some unavoidable external reasons, the lidar failed to carry out the observation experiment at the same time as the MMCR. To further analyze the changes in the height distribution of cloud boundaries in Xi'an in 2021, we plan to employ MMCR data to replace the data of periods when the lidar is not running.	line 337-342	To further analyse the changes in the height distribution of cloud boundaries in Xi'an, we plan to use MMCR and lidar data for cloud boundary analysis.	line 360-361
35	principles and detection algorithms	line 348	principles and algorithms	line 367
36	Figure 16	line 350-351	Figure 14	line 369-370
37	From the above three cloud observation cases, it can be seen that MMCR has more advantages than lidar in detecting cloud top boundaries. Therefore, when calculating the cloud boundary height distribution characteristics over Xi'an in 2021, we only count the cloud top boundary height detected by MMCR and take it as the actual cloud top boundary. The statistical rules shown in Table 3 are established for the statistics of cloud bottom boundary information. The experimental data of 302 days (65 days in spring (January-March), 84 days in summer (April-June), 65 days in autumn (July-September) and 88 days in winter (October-December) observed in 2021 are classified and sorted out to ease the statistics and analysis of the variation characteristics of cloud boundary height.	line 352-358	From the above three cloud observation cases, it can be seen that MMCR has more advantages than lidar in detecting cloud-top boundaries. Therefore, when calculating the cloud boundary height distribution characteristics over Xi'an, we only counted the cloud top boundary height detected by the MMCR and considered it as the actual cloud top boundary. From December 2020 to November 2021, MMCR and lidar stored 302 d (7248 h) and 126 d (872.5 h) of observational data, respectively. During the 12-month observation period, the maximum detection altitude of the MMCR changed. From December 2020 to June 2021, the maximum detection range of MMCR is 12.6 km, and the maximum detection height is changed to 18 km. The total observation hours of MMCR and lidar for each month are shown in Fig. 15. The hours of lidar, MMCR, and simultaneous measurements are 872.5 h. In this study, the four seasons were defined as follows: spring from March to	line 371-380

			May (MAM), summer from June to August (JJA), autumn from September to November (SON), and winter from December to February (DJF).	
38			Add Figure 15	line 381-832
39	Table 3 Statistical rules of cloud bottom boundary information	line 359	Table 3 Cloud bottom height recording guideline	line 385
40	Thin cloud: bottom of MMCR is blurred Drizzle: bottom of MMCR is blurred	In table3 line 359	Geometrical thin cloud: bottom of MMCR is invalid Drizzle: bottom of MMCR is invalid	In table3 line 385
41	It is observed that the total sample size is N ,	line 362	The total sample size is N	line 367
42	Fig. 17 shows that the cloud top boundary occurrence frequency in spring and summer presents a bimodal distribution. In spring, the height of the first peak lies approximately 1.5 ~ 1.9 km, and the second peak is 7.8 ~ 8 km. The heights of the first and second peaks are approximately 1.5 ~ 3 km and 8 ~ 12 km, respectively, in summer. In autumn and winter, the frequency of cloud top boundary heights above 2 km is almost in the range of 0.3 to 0.4. For the vertical distribution characteristics of the cloud bottom boundary, there is a triple-mode pattern in four seasons. The frequency distribution characteristics of the cloud bottom boundary height in spring and summer are relatively similar. The first most obvious narrow peak < 1.5 km is the frequency change caused by boundary layer clouds, the second narrow peak is located at 3 ~ 4 km, and the third peaks in spring and summer are located at 6 ~ 8 km and 7 ~ 9 km, respectively. From the distribution characteristics of the cloud bottom boundary in summer and spring, it can be guessed that convective and cirrus clouds may be dominant in these two seasons. The frequency distribution of clouds above 8 km in autumn is the largest in the four seasons, and we can speculate that stratus clouds and cumulus clouds are mainly in this season. In winter, the height range of clouds is	line 366-378	Fig. 16 shows the vertical frequency distribution of the cloud boundary seasonally from December 2020 to November 2021. For the vertical distribution of cloud base, the first narrow peaks is the boundary layer clouds (≤ 1.5 km), the second peak is 2.5–3.5 km, and the third peak has a big range in vertical height, which is 4.7–10 km a in spring. Fig.16 (b) shows that the cloud bottom height in summer is mainly distributed at 3–9.5 km, indicating that middle and high clouds may be dominant. The distribution of cloud bottom is bimodal, the first peak is the boundary layer cloud peak, and the second peak is located at 2.7-3.7 km and 3.6–8.3 km in autumn and winter, respectively. The variation in cloud top with seasons shows a bimodal distribution, and spring and summer have a similar trend of cloud top boundary height distribution. The frequency of the cloud top boundary above 10 km was the highest, and the frequency below 2 km was the lowest in summer. The distribution characteristics of cloud top height in autumn and winter indicate that the frequency of low clouds is higher than that in the other two seasons.	line 391-400

	narrow, and the numerical range is wide, which may be mainly stratiform clouds.			
43	Figure 17	line 382-384	Figure 16	line 404-405
44	Fig. 18a) shows the monthly variation frequency distribution of clouds. The months with the largest (96%) and smallest (42%) cloud occurrence frequencies are August and December, respectively. Almost more than 34% of the clouds appear in the form of single layer clouds every month. Compared with January, February, November and December, the frequencies of double-layer clouds, triple-layer clouds and more clouds in other months are higher. It is also possible that there are some thin clouds and broken clouds in the upper layer, which are summarized as multilayer clouds by the algorithm. As shown in Fig. 18b), the normalized monthly distribution of cloud amount shows that the minimum cloud amount is 0.65 in spring and the maximum is 2.46 in summer,	line 385-391	Fig. 17 a) shows the monthly variation frequency distribution of clouds. The months with the largest and smallest cloud occurrence frequencies are August and February, respectively. Almost more than 34% of the clouds appear in the form of single layer clouds every month. Compared with January, February, November, and December, the frequencies of double-layer clouds, triple-layer clouds, and more clouds in other months are higher. To show the relative change trend of cloud cover, we calculated the total cloud cover of each month by using the total cloud cover at each time stored by the MMCR. It was found that the maximum cloud cover was in April. Therefore, the total cover of April was set to 1, and the normalized cloud cover distribution of 12 months was obtained, as shown in the Fig.17 b). It can be seen from the distribution of cloud cover in every month that the cloud cover is high in summer and the least in winter,	line 407-415
45	Figure 18	line 393-395	Figure 17	line 417-419
46	Fig. 19 shows the frequency change characteristics of the cloud boundary vertical height distribution in 2021, in which the frequency of the cloud bottom boundary below the vertical height of 1.5 km is greater than 10%, the frequency within the height range of 3.06 km and 3.6 km is approximately 3.24%, and the frequency above 8 km is less than 2%. The frequency of the cloud top boundary at vertical heights has a bimodal distribution; the first narrow peak is located at 1.5~3.1 km, and the second peak lies at approximately 7.5~10.5 km. Combined with the changing characteristics of cloud layers, it can be seen that during the observation process in Xi'an in 2021, the frequency of stratiform clouds below 3.5 km is the largest, and the frequency of high-level ice clouds or cirrus clouds above 8 km is small.	line 396-403	As Fig.18 caption says it is the frequency distribution of cloud boundaries observed over Xian from December 2020 to November 2021. Frequency of the cloud bottom boundary below the vertical height of 1.5 km is the highest, the frequency within the height range of 3.06 km and 3.6 km is approximately 0.4%, and the frequency above 8 km is less than 0.2%. The frequency of the cloud top boundary at vertical heights has a bimodal distribution, and the first narrow peak is located at 1.0~3.1 km, and the second peak lies at 6.4~10.5 km. Combined with the changing characteristics of cloud layers, it can be seen that during observation in Xi'an, the frequency of clouds below 3.5 km is the largest, and the frequency of high-level ice clouds or cirrus clouds above 8 km is small.	line 420-429
47	Figure 19	line 404-405	Figure 18	line 42-429

5) Modification the part of Conclusions

We rewrote the conclusions and replaced all the conclusions in the original manuscript.

Modification the part of Author contribution

Number	original manuscript	Number of line	revised manuscript	Number of line
1	In this paper, Yun Yuan proposed new methods and theories, processed and analyzed the data in the paper, and wrote the manuscript. Huige Di made many revisions and guidance to the manuscript, and put forward many feasible suggestions. Tao Yang collected the experimental data and maintained the equipment. Yuanyuan Liu sorted and classified the original experimental data. Qimeng Li calibrated the experimental system. Qing Yan coordinated the placement and calibration of experimental equipment. Dengxin Hua led the project, instrument development, experimental design and data analysis.	line 438-443	Conceptualization: Yun Yuan Investigation: Yun Yuan Methodology: Yun Yuan and Huige Di Software: Yun Yuan Supervision: Huige Di and Dengxin Hua Methodology and software improvement: Yuanyuan Liu, Tao Yang, Qimeng Li, Qing Yan, Wenhui Xin, and Shichun Li. Writing – original draft: Yun Yuan Writing – review & editing: Yun Yuan and Huige Di Project administration: Dengxin Hua	line 458-467

Modification the part of Financial support

Number	original manuscript	Number of line	revised manuscript	Number of line
1	This research was supported by the National Natural Science Foundation of China (NSFC): 61875163 and 42130612.	line 459-460	This research has been supported by the National Natural Science Foundation of China, Innovative Research Group Project of the National Natural Science Foundation of China (grant nos. 42130612, 41627807 and 61875162) and the Ph.D. Innovation fund projects of Xi'an University of Technology (Fund No.310-252072106).	line 471-473

Added 5 references, indicated in blue font