Satellite imagery analysis for nighttime temperature inversion clouds


Abstract

Clouds play important roles in the climate system. Their optical and microphysical properties, which largely determine their radiative property, need to be investigated. Among several measurement means, satellite remote sensing seems to be the most promising. Since most of the cloud algorithms proposed so far are daytime use which utilizes solar radiation, Minnis et al. (1998) developed a nighttime use one using 3.7-, 11- and 12-μm channels. Their algorithm, however, has a drawback that is not able to treat temperature inversion cases. We update their algorithm, incorporating new parameterization by Arduini et al. (1999) which is valid for temperature inversion cases. This updated algorithm has been applied to GOES satellite data and reasonable retrieval results were obtained.

1. Introduction

Regardless of the importance of clouds in the formation of the Earth's climate, its quantitative understanding is yet to be done. One of the important properties of clouds is the radiative property, which is determined by the optical depth, droplet size, top temperature and so on. To determine these cloud physical properties accurately is crucial, and it would lead to contribution of improvement for cloud processes, which are not fully resolved, in numerical models such as GCM (General Circulation Model).

Determination of cloud physical parameters using remote sensing techniques has been conducted actively during the last decade. (e.g., Nakajima et al. 1991) In particular satellite remote sensing is promising in view of large spatial coverage and constant temporal resolution. (e.g., Han et al. 1994) Most cloud retrieval algorithms have been constructed only for daytime and require solar spectral channels. Except for CO2-slicing
methods, techniques applicable to nighttime are rare because fewer wavelengths are available. Since a consistent 24-hour retrieval of cloud properties is indispensable for studies of cloud diurnal variability, the hydrological cycle and cloud-radiation interactions, a nighttime cloud algorithm using common infrared spectral channels has been developed for application to both research and operational meteorological satellites. Our method makes use of brightness temperature difference (BTD) for determining the optical depth, droplet size and temperature of clouds (Inoue, 1985, Minnis et al. 1995). Minnis et al. (1998) developed parameterization for reflectance and effective emittance of clouds in order to treat huge satellite data set efficiently. Their parameterizations are, however, only valid under the condition of the brightness temperature at cloud base, $T_s$ minus the cloud temperature, $T_c > 4K$. Then Arduini et al. (1999) constructed a parameterization for brightness temperature (BT) of clouds under the condition of $T_s-T_c < 4K$. This temperature range covers fog or temperature inversion clouds, which are particularly common in polar regions. In this article, we will report a case study with our nighttime cloud algorithm which incorporated Arduini et al's parameterization. In section 2, retrieval principles of our method and Arduini et al's parameterization will be described, and in section 3 satellite data and some other sounding data will be presented. We will summarize results in the final section.

2. Method

1) Retrieval principles

The Solar-infrared Infrared Split-window Technique (SIST) uses 3.7-, 11- and 12-μm channels to invert three cloud parameters, that is, the optical depth $\tau$, effective water droplet radius $r_e$ or effective ice particle diameter $D_e$ and temperature $T_c$ of both water and ice clouds. Because of little sensitivity of the particle size, we limit our analysis to clouds whose optical thicknesses less than 6. In principle, these parameters can be determined by matching BTDs between observation and model calculation as closely as possible. The parameterization to calculate modeled BT will be presented in the next subsection. Meteorological data such as temperature, isobaric height and relative humidity are taken from the nearest sounding data. Water vapor absorption is based on the correlated k-distribution by Kratz (1995).
The analysis flow is as follows. In one particle model, $\tau$ is determined at a given $T_c$ so as to generate the modeled BT at 11-$\mu$m as closely as possible to the observed counterpart, and then error between observed and modeled BTDs which $\tau$ and $T_c$ produce is calculated. We search $T_c$ which produces the minimum BTD error by bisection method between BT at 11-$\mu$m and the maximum sounding temperature. This procedure is done for both water and ice models, and then the pair of $\tau$, $re$ or $De$ and $T_c$ which results in smaller BTD error between both phases is taken as solution. Phase selection is, however, taken into account in some particular conditions. For example, ice will be selected if $T_c$ (ice) < 243K, or if both $T_c$ (water) and $T_c$ (ice) < 253K, or if $T_c$ (water) < 273K and $re$ < 2.5$\mu$m. Water will be selected if $T_c$ (water) > 270K and $T_c$ (ice) > 265K.

2) cloud-top BT parameterization for temperature inversion situations

To model the brightness temperatures (BT) of clouds, Arduini et al. (1999) developed its parameterization under the condition of $T_s-T_c < 4$K, that is, the cloud temperature is warmer than or similar to that of the underlying layer. It is functions of $\tau$, $re$ and $De$, viewing zenith angle, $T_c$ and $T_s$, and $\tau$ covers from 0.25 to 32, $T_c$ ranges from 240K to 300K and viewing zenith angle is up to 78 degrees. Water-droplet clouds were represented by using modified Gamma distribution of spherical droplets having $re$ from 2$\mu$m to 32$\mu$m. On the other hand, ice-crystal clouds were assumed to be made up of distributions of hexagonal ice crystals whose optical properties were based on the ray-tracing results of Takano and Liou (1989) and the spheroidal approximation of Takano et al. (1992). $De$ ranges from 6$\mu$m to 135$\mu$m.

An adding-doubling radiative transfer code was used to simulate cloud-top BT, and then a standard linear least squares multiple regression technique, which is similar to that of Minnis et al. (1998), was taken to analyze the model results and to determine the parameterization. Results show that this parameterization successfully describes the variation in the temperature with the root mean square of the residuals well within 0.5K.

3. Data
In this work, we use GOES (Geostationary Operational Environmental Satellite) data over Oklahoma on Oct. 26 1995. Well-defined fog developed in the northern Oklahoma on that day. This area is suitable for validating the present algorithm, since some ground measurements were conducted at ARM (Atmospheric Radiation Measurement) site (36.6N, 97.5W).

Figure 1 shows the GOES IR (11-μm) imagery. Area enclosed by square, where is darker than surroundings (it means warmer than nearby ordinary clouds), corresponds the fog involved in this paper.

4. Result and concluding remarks

We analyzed the area indicated in Fig. 1 with the present algorithm. Figure 2 shows the parameterization result for BTD between 3.7-μm and 11-μm as a function of BT at 11-μm with regard to several values of τ and re, together with observation points superimposed. In this example, we set skin temperature as 281K taken from the nearest measurement point. The cloud height is retrieved about 0.8km, τ is retrieved about 0.4 to 1.2, and re is inferred 4 to 12μm. As for the cloud height, lidar at ARM site indicates similar values to ours. Air temperature and dew-point temperature vertical profiles from the sounding also support our retrievals as shown in figure 3. Such a type of cloud occurs near the temperature inversion height in general, and humidity is also found to be enhanced. As for re, these values seem to be consistent with those of continental low clouds. In addition, surface observers reported no drizzle which usually associated with larger droplets.

In particular, it might be very sensitive to input temperatures for remote sensing techniques using infrared channels like the present algorithm. We present another computation using the parameterization with different skin temperature input in figure 4. We set it as 275K in this case. Then re would be retrieved as much larger and we would get totally wrong results. It clearly shows how it is important to use accurate temperatures.

In this paper, we have incorporated cloud BT parameterization for temperature inversion cases by Arduini et al. (1999) to the nighttime cloud analysis framework by Minnis et al. (1995). This new cloud analysis system has been applied to low-level clouds found in GOES satellite imagery over Oklahoma. Our retrieval for the cloud height is consistent with the lidar result and sounding. As for re, the values seem to be reasonable. This
algorithm would be a powerful tool for global analysis and statistical study of clouds, together with the one with Minnis et al. (1998).

References


Nakajima, T., M. D. King, J. D. Spinhirne and L. F. Radke, 1991: Determination of the optical thickness and effective radius of clouds from


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fig. 2
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Fig. 4