

# Albedo-Ice regression method for determining ice water content of Polar Mesospheric Clouds using ultraviolet observations from space

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**Abstract.** High spatial-resolution images of Polar Mesospheric Clouds (PMC) from a camera array onboard the Aeronomy of Ice in the Mesosphere Satellite have been obtained since 2007. The Cloud Imaging and Particle Size Experiment (CIPS) detects scattered ultraviolet (UV) radiance at a variety of scattering angles, allowing the scattering phase function to be measured for every image pixel. With well-established scattering theory, the mean particle size and ice water content (IWC) are derived. In the nominal mode of operation, approximately seven scattering angles are measured per cloud pixel. However, because of a change in the orbital geometry in 2016, a new mode of operation was implemented such that one, or at most two, scattering angles per pixel are now available. Thus particle size and IWC can no longer be derived from the standard CIPS algorithm. The Albedo-Ice Regression (AIR) method was devised to overcome this obstacle. Using data from both a microphysical model and from CIPS in its normal mode, we show that the AIR method provides sufficiently accurate average IWC so that PMC IWC can be retrieved from CIPS data into the future, even when albedo is not measured at multiple scattering angles. We also show from the model that 265nm UV scattering is sensitive only to ice particle sizes greater than about 20-25 nm in (effective) radius, and that the operational CIPS algorithm has an average error in retrieving IWC of  $-13\pm 17\%$ .

## 25 **1 Introduction**

26 Polar Mesospheric Clouds (known as noctilucent clouds in the ground-based literature) have been studied for over a  
27 century from high-latitude ground observations, but only since the space age have we understood their physical nature, as  
28 water-ice particles occurring in the extremely cold summertime mesopause region. Their seasonal and latitudinal variations  
29 have now been well documented (DeLand et al., 2006). Interest in these clouds ‘at the edge of space’ has been stimulated by  
30 suggestions that they are sensitive to global change in the mesosphere (Thomas et al., 1989). This expectation has been  
31 supported recently by a time series analysis of Solar Backscattered Ultraviolet measurements of PMC (Hervig et al., 2016)  
32 and by model calculations (Lübken et al., 2018).

33 The Aeronomy of Ice in the Mesosphere satellite (AIM) (Russell et al., 2009) was designed to provide a deeper  
34 understanding of the basic processes affecting PMC, through remote sensing of both the clouds and their physical  
35 environment (temperature, water vapor, and meteor smoke density, among other constituents). One of the two active  
36 experiments on board AIM is a camera array, the Cloud Imaging and Particle Size (CIPS) experiment, which provides high  
37 spatial resolution images of PMC (McClintock et al., 2009). CIPS measures scattered ultraviolet (UV) sunlight in the nadir  
38 in a spectral region centered at 265 nm, where ozone absorption allows the optically-thin ice particles to be visible above the  
39 Rayleigh scattering background issuing from the ~50-km region (Rusch et al., 2009; Bailey et al., 2009). Because of its wide  
40 field of view and 43-second image cadence, CIPS views a cloud element multiple times in its sun-synchronous orbital  
41 passage over the polar region, thus providing consecutive measurements of the same location at multiple (typically seven)  
42 scattering angles (SA). Together with scattering theory, the brightness of the cloud (albedo) at multiple angles provides  
43 constraints needed to estimate the mean ice particle size (Lumpe et al., 2013). From the particle size and albedo  
44 measurements, the ice water content is calculated for each cloud element ( $7.5 \times 7.5 \text{ km}^2$  in the most recent CIPS retrieval  
45 algorithm). However, over time, the AIM orbit plane has drifted from its nominal noon-midnight orientation to the point  
46 where the satellite is currently operating in a terminator orbit. Responding to this altered geometry and the desire to  
47 broaden the scope of AIM, new measurement sequences were implemented to provide observations of the entire sunlit  
48 hemisphere, rather than just the summertime high-latitude region. Because the total number of images per orbit is fixed by  
49 data storage limitations, a new mode (the ‘continuous imaging mode’) of observations, with a reduced three-minute image  
50 cadence, was implemented in February 2016. The present sampling in a single Level 2 pixel contains many fewer scattering  
51 angles (often only one). To maintain consistency in the study of inter-annual variations of PMC, this necessitates a revised  
52 method of retrieving ice water content (IWC) where only a single albedo measurement is available. IWC is a valuable  
53 measure of the physical properties of PMC since it largely removes the effects of scattering-angle geometry, is a convenient  
54 PMC climate variable when averaged over season, and can be used in comparing with contemporaneous measurements of  
55 PMC that use different observational techniques.

56 The Albedo-Ice Regression (AIR) method was developed to fill the need to retrieve PMC IWC with only a single  
57 measurement of albedo. Based on the simple notions that both albedo and IWC depend linearly upon the ice-particle column  
58 density, multiple linear relationships are established between IWC and cloud directional albedo, depending upon scattering  
59 angle. The regressions are derived from three data sources: (1) the Specified Dynamics version of the Whole Atmosphere  
60 Community Climate Model (SD-WACCM) combined with the Community Aerosol and Radiation Model for Atmospheres  
61 (CARMA); (2) CIPS data for the years 2007-2013, when multiple scattering angles were available to derive IWC; and (3)  
62 Solar Occultation For Ice Experiment (SOFIE), which provides IWC and particle sizes. These three sources provide many  
63 thousands of albedo-IWC-particle size combinations, from which the AIR regressions are derived. Although the AIR method  
64 may be inaccurate for a single retrieval of IWC, averages over many observations result in close agreement as the number of  
65 data points increases. The utility of AIR thus depends upon the availability of large data sets that apply to roughly the same  
66 atmospheric conditions. For example, we will show CIPS results for July and January averages for ascending and descending  
67 portions of the orbit.

68 In this paper we first describe the theoretical framework relating the scattered radiance to mesospheric ice particles. It is  
69 desirable to use model data to test the AIR method, without the complications of cloud heterogeneity and viewing geometry.  
70 We utilized a first-principles microphysical model that accurately simulates large numbers of cloud properties (number  
71 density and particle size distribution). The processes treated by the model include nucleation on meteor ‘smoke’ particles,  
72 growth, and sedimentation, occurring in a saturated environment at density and temperature conditions provided by the main  
73 global climate model (Bardeen et al., 2010). Several runs for one-day and multiple-day periods during summer solstice  
74 conditions for solar conditions applying to 1995 were analyzed. Cloud radiances (albedos) at 265 nm were calculated for the  
75 SA range encountered by the CIPS experiment. We chose a set of cloud simulations to derive a single set of two AIR  
76 coefficients through linear regression. The accuracy of the AIR approximation was then tested on the same data, and on  
77 other model runs, using averages as a function of SA, and increasing IWC threshold values. Thresholding is necessary to  
78 account for the fact that different measurement techniques have different detection sensitivities. This is not a signal/noise  
79 issue, rather the ability to discriminate PMC against a background that is usually larger than the PMC signal itself. We show  
80 in particular how seasonal means of IWC can be derived from Solar Backscatter Ultraviolet Spectrometer (SBUV) radiance  
81 data, without the need to derive particle size.

82 Having tested the technique for model data, we use the same approach with real-life PMC data collected from CIPS in the  
83 normal pre-2016 operating mode. This mode provided scattering angles needed to define an ice scattering phase function,  
84 from which mean particle size was derived based on assumed properties of the underlying size distribution (Lumpe et al.,  
85 2013). The regressions were run for a period of 40 days in each of four seasons, each comprising millions of separate cloud  
86 measurements, and from both summertime hemispheres. The results were combined into a single set of AIR coefficients, and

87 again the AIR technique was tested on monthly averages. These averages were constructed over all years of nominal  
 88 spacecraft operations (2007-2013 in the northern hemisphere, and 2007-8 through 2013-14 in the southern hemisphere).  
 89 Note that testing the accuracy of the AIR technique during the nominal mission period allows the method to be used even  
 90 during the ‘continuous imaging mode’ of CIPS operation.

91 We then employed highly-accurate data from SOFIE for ice column density and mean particle size. Since the SOFIE  
 92 technique uses near-IR solar extinction in ice-water absorption bands, the primary measurement is ice water content. As  
 93 shown in Sec 2.3, we inverted the retrieved SOFIE IWC to derive the equivalent 265-nm albedo, and then applied the  
 94 regression method described above to the results.

95 After describing the AIR method, we discuss briefly the application of the method to a third contemporaneous experiment,  
 96 the SBUV satellite experiment, which has in common the same limitations as CIPS in its continuous-imaging mode, namely  
 97 that measurements of nadir albedo are made at a single scattering angle. This has already resulted in a publication (DeLand  
 98 and Thomas, 2015) where we provided a time series of PMC IWC from the AIR method extending back to the first SBUV  
 99 experiment in 1979.

## 00 **2 Theoretical Basis**

01 Here we provide a brief overview of the theoretical basis of the IWC retrieval technique, referring to previous publications  
 02 for more detail (Thomas and McKay, 1985; Rusch et al., 2009; Bailey et al., 2009, Lumpe et al., 2013). The basic  
 03 measurement is PMC cloud radiance  $I(\Phi, \theta)$  where  $\Phi$  is the scattering angle (angle between the sun and observation vectors)  
 04 and  $\theta$  is the view angle, which is the angle subtended by the nadir and observation direction, measured from the point of  
 05 scattering. Since the ice layer is optically thin, and secondary scattering is negligible, the albedo is described by first-order  
 06 scattering. The ratio of scattered (detected) radiance to the incoming solar irradiance  $F_\lambda$  is the albedo  $A_\lambda$ , where

$$A_\lambda(\Phi, \theta) = I_\lambda(\Phi, \theta) / F_\lambda = \sec \theta \int_{z_t}^{z_b} dz' \int_{r_{\min}}^{r_{\max}} dr' \sigma_\lambda(r, \Phi) n(r', z'). \quad (1)$$

07 Here  $n(r', z') dr' dz'$  is the column density of ice particles ( $\text{cm}^{-2}$ ) in the ranges  $r', r'+dr'$  and  $z', z'+dz'$ . For CIPS  
 08 measurements, each camera has a finite bandpass, centered at 265 nm, and is characterized by a function  $R_\lambda$  with an  
 09 effective width of 10 nm (McClintock et al., 2009). The albedo  $A_\lambda^m$  derived from this instrument is given by

10 Here  $z'$  and  $r'$  are the height and particle radius variables, and  $z_b$  and  $z_t$  define the height limits of the ice layer, with the  
 11 majority of the integrand extending between 83 and 85 km.  $r_{\min}$  and  $r_{\max}$  are particle radii which span the particle size regime

12 responsible for scattering (from ~20 nm to ~150 nm). As shown by Rapp and Thomas (2006), particles with sizes < 20 nm  
 13 are not detectable by UV measurements because of their small cross-section values – hence we refer to ‘UV-visible clouds’.  
 14  $\sigma_\lambda$  is the differential scattering cross-section (cm<sup>2</sup>-sr<sup>-1</sup>) at wavelength  $\lambda$  and scattering angle  $\Phi$ .  $n(r',z')dr'dz'$  is the  
 15 number density of ice particles (cm<sup>-2</sup>) in the ranges  $r', r'+dr'$  and  $z', z'+dz'$ . For CIPS measurements, each camera has a finite  
 16 bandpass, centered at 265 nm, and is characterized by a function  $R_\lambda$  with an effective width of 10 nm (McClintock et al.,  
 17 2009). The albedo  $A_\lambda^m$  derived from this instrument is given by

$$A_\lambda^m(\Phi, \theta) = \sec \theta \int_{z_t}^{z_b} d\lambda' R_{\lambda'} \int_{r_{\min}}^{r_{\max}} dz' \int dr' \sigma_\lambda(r, \Phi) n(r', z'). \quad (2)$$

18  
 19 In the model, the ice particles are assumed spherical, but the scattering theory should take account of the non-spherical  
 20 nature of ice crystals. The best agreement of theory with near-IR mesospheric ice extinction occurs for a randomly rotating  
 21 oblate- spheroid shape, of axial ratio two (Hervig and Gordley, 2010). This shape is assumed in the calculation of the cross-  
 22 section, which is accomplished through a generalization of Mie-Debye scattering theory, the T-matrix method (Mishchenko  
 23 and Travis, 1998). The radius in the T-matrix approach is defined as the radius of the volume-equivalent sphere. In the  
 24 model calculations, we will ignore the view angle effect. In the reported CIPS data, the  $\sec\theta$  factor is applied to the reported  
 25 albedos, so that  $A$  always refers to the nadir albedo ( $\theta = 0^\circ$ ).

26  
 27 The ice water content (IWC) is the integrated mass of ice particles over a vertical column through the layer. Its definition is

$$IWC = \rho \int_{z_b}^{z_t} dz' \int_{r_{\min}}^{r_{\max}} dr' (4\pi / 3) r'^3 n(r', z'). \quad (3)$$

28  $\rho$  denotes the density of water-ice at low temperature (0.92 g-cm<sup>-3</sup>). Anticipating the results of this study that IWC is linearly  
 29 related to the column density of ice particles,  $N = \int dr' \int dz' n(r, z') dz'$ , we explore the physical basis of this result. As  
 30 pointed out by Englert and Stevens (2007) and Hultgren and Gumbel (2014) such a relationship exists for certain SA values,  
 31 for which  $\sigma \sim r^3$ , in which case it is easily seen that Eq. (2) is proportional to IWC. However, we find that a linear  
 32 approximation is valid for a much wider range of scattering angles. To understand this result, we imagine that all particles  
 33 have the same radius, so that  $n = n_c \delta(r - r_c)$ , where  $\delta$  is the Dirac  $\delta$ -function. Then Eqs. (1) and (3) ‘collapse’ to a simpler  
 34 result,

$$A_{\lambda}(\Phi, 0) = \sigma_{\lambda}(r_c, \Phi)N(r_c), IWC(r_c) = \rho V(r_c)N(r_c). \quad (4)$$

35 Here  $N(r_c) = n_c \Delta z$  where  $\Delta z$  is the effective vertical layer thickness. Eliminating the column density  $N(r_c)$ , IWC is written

$$IWC(r_c) = \rho V(r_c) A(\Phi, 0) / \sigma_{\lambda}(r_c, \Phi). \quad (5)$$

36  $V(r_c)$  denotes the particle volume. Thus in this special case,  $IWC(r_c) \sim A_{\lambda}(\Phi, 0)$ . A superposition of the effects of all  
 37 participating particle sizes will exhibit a similar proportionality. When  $IWC(r)$  is integrated over all  $r$ , the contributions from  
 38 each size are straight lines, each having different intercepts and slopes.

39

40 As previously discussed, the value of the AIR method is in evaluating *average* IWC (denoted by  $\langle IWC \rangle$ ) over many albedo  
 41 observations made at numerous scattering angles. The accuracy of the method should be assessed primarily on this basis, not  
 42 on how well an individual albedo measurement yields the correct value of IWC. However we also address the error of using  
 43 individual albedo measurements in estimating IWC. An additional issue is the differing detection thresholds for IWC among  
 44 the various experiments. In the case of the scattered-light experiments, the detection threshold depends upon how well the  
 45 cloud radiance data can be separated from the bright Rayleigh-scattered background. The CIPS experiment retrieval method  
 46 relies upon high spatial resolution over a large field of view, and the differing scattering-angle dependence of PMC and the  
 47 Rayleigh-scattering background (Lumpe et al., 2013). The SBUV retrieval relies upon differing wavelength-dependence of  
 48 PMC and background, but primarily on the PMC radiance residuals being higher (2 sigma) than fluctuations from a  
 49 smoothly- varying sky background (Thomas et al., 1991; DeLand and Thomas, 2015). The AIM SOFIE method is very  
 50 different, being a near-IR solar extinction measurement in multiple wavelength bands. SOFIE can detect much weaker  
 51 clouds with smaller effective sizes than either CIPS or SBUV. Particle radii values as small as 10 nm can be retrieved from  
 52 the SOFIE data (Hervig et al., 2009). To compare the various experiments, it is necessary to ‘threshold’ the data from more  
 53 sensitive experiments with a cutoff value of IWC.

54

55 In the next three sections, we present the AIR results from the model, CIPS and SOFIE, using averages over many cloud  
 56 occurrences. It is not our intention to compare the different thresholded data sets to one another (this task will be relegated to  
 57 a separate publication), but to illustrate how even measurements made at a single scattering angle (e.g., SBUV) can yield  
 58 averaged IWC values that are sufficiently accurate to assess variations in daily and seasonal averages. These variations are of  
 59 crucial value to determining solar cycle and long-term trends in the atmospheric variables (mainly temperature and water  
 60 vapor) that control ice properties in the cold summertime PMC region. We examine the accuracy of AIR through simulations

61 of scattered radiance from the model, and from CIPS and SOFIE data. Since these data sources yield particle radii, they can  
62 provide both the ‘actual’ and approximate values of IWC from the regression formulas. Hervig and Stevens (2014) used the  
63 spectral content of the SBUV data to provide limited information on particle size. Together with the albedos themselves,  
64 they used this information to derive seasonally-averaged ice water content. They showed that the variation of mean particle  
65 size over the 1979-2013 time period was relatively low (standard deviation of  $\pm 1$  nm). They also found a very small  
66 systematic increase with time, as discussed in Sec. 3.

67

## 68 **2.1 Model Results**

69 Using a microphysical model as a reference source of IWC ‘data’ is useful, in the following ways: 1) in contrast to the CIPS  
70 and SOFIE retrieval algorithms, no artificial assumptions are needed concerning the size distribution of ice particles; (2)  
71 limitations due to background removal are absent; (3) radiance and IWC may be calculated accurately, so that effects of  
72 cloud inhomogeneity are absent. With regard to the latter point, we describe in more detail the model calculations. The  
73 model grid is  $4^\circ$  in latitude,  $5^\circ$  in longitude and variable in the vertical. Ice particles of varying sizes fill many of these cells,  
74 but the density of particles within each cell is, by definition, constant. For a given model cloud, the integration is made  
75 through a vertical ‘stack’ of all ice-filled cells generated in a given computer run, and within each particle size grid. The total  
76 radiance is the sum of contributions from the size range 20 to 150 nm. The observation angles are always assumed to be  
77 zero, in other words, the integration is performed in the vertical only. Thus cloud ‘boundaries’ in the horizontal plane are not  
78 an issue. This contrasts with real heterogeneous clouds where these approximations would not hold. The model contains  
79 variability due to waves of various sorts, including tides and gravity waves. However, it does not capture all known details of  
80 PMC, such as double layers. Since we are dealing with integrated quantities, this should not be an important issue.  
81 Furthermore, we don’t place full reliance on the model, which is why we also use two independent data sets.

82

83 To gain insight into the accuracy of the AIR approach, it is sufficient to work with monochromatic radiance at the central  
84 wavelength of the various passbands. The integrations of Eqs. (1) and (3) were approximated by sums over variable  
85 increments of radius, and over all sub-layers within the model ice cloud (a typical ice layer is several km thick.). The model  
86 height grid is variable, so that the smallest layer thickness is 0.26 km, which resolves the narrow ice layers (see Bardeen et  
87 al., 2010 for more details). We then performed the linear regression for SA values over which CIPS observations are made.

88

89 Figure 1 displays the regressions for six scattering angles, and 2514 individual model clouds. The units of IWC are  $\text{g}\cdot\text{km}^{-2}$ ,  
90 or equivalently  $\mu\text{g}\cdot\text{m}^{-2}$ , which are commonly used in the literature. Each plot is divided into two groups according to the  
91 effective radii  $r_{eff}$  for each cloud.  $r_{eff}$  is defined in the literature (Hansen and Travis (1974) as

$$r_{eff} = \int dr' n(r') r'^3 / \int dr' n(r') r'^2. \quad (6)$$

92 Figure 1 clearly illustrates that particle size contributes to the ‘scatter’ from the linear fits. For the conditions in Fig. 1(c), the  
 93 mean error of AIR for a *single* model simulation is 19%. The error can be reduced substantially by averaging. For example,  
 94 for 100 measurements, the AIR error in the average IWC is only 2%. Figure 1 also shows the existence of a non-zero  
 95 intercept of IWC vs albedo. The non-zero intercept was at first surprising since we expected that for an albedo of zero, IWC  
 96 should also be zero. In fact, we found that the linear relationship breaks down for very small albedo, and the points in the  
 97 plot narrow down in this limit (not shown). In albedo units of  $10^{-6}$  sr<sup>-1</sup> (hereafter referred to as 1 G) this departure from  
 98 linearity occurs for  $A < 1$  G and  $IWC < 10$  g-km<sup>-2</sup>, conditions which fortunately are below the detection threshold of CIPS and  
 99 SBUV, and are a result of the very faint small particles. For more sensitive detection techniques, this limitation must be kept  
 00 in mind. A limitation of the present model (not necessarily all models) is that it does not simulate the largest particles in  
 01 PMC and the largest values of IWC, as seen in both AIM SOFIE and CIPS experiments. The largest model IWC value is 180  
 02 g-km<sup>-2</sup> and the largest effective radius is 66 nm, whereas CIPS and SOFIE find particle radii up to 100 nm and IWC up to  
 03 300 g-km<sup>-2</sup>. This limitation is irrelevant for the AIR CIPS results (to be discussed), but could limit the application of the AIR  
 04 technique to SBUV data. In Sec. 3 we will return to the issue of the AIR accuracy, as applied to SBUV data.

05

06 We chose to use averages for the entire model run, which includes different latitudes, longitudes, and UT, but the data can be  
 07 divided in many different ways. It is certainly preferable in data sets to choose a small time and space interval over which  
 08 temperature and water vapor are not expected to vary, but this is not necessary for the model. All that we ask of the model is  
 09 whether the AIR results provide an accurate estimate of  $\langle IWC \rangle$ , taken over the ensemble of model cloud albedos calculated  
 10 at a variety of scattering angles.

11

12 As discussed above, we are also interested in the accuracy of AIR in the thresholded data, that is, how AIR represents  
 13  $\langle IWC \rangle$  in comparisons of data sets with varying detection sensitivities to PMC. Figure 2 displays the error in the ensemble-  
 14 average (2488 model clouds) as a function of the IWC threshold and scattering angle. Despite the large data scatter from the  
 15 linear fit shown in Fig. 1, the averaging removes almost all the influence of the ‘random error’. In this case, the overall error  
 16 is less than 3%. The influence of particle size is of course not a random error, but acts like one in the averaging process.  
 17 However, the AIR coefficients also depend weakly upon the mean effective radius, defined in Eq. (6) for a single cloud,  
 18 which varies from one latitude to another and from year to year. The effect of variable  $r_{eff}$  on the AIR error is discussed in  
 19 Sec. 3.

## 20 **2.2 AIR Results from CIPS**



21

22 A detailed description of the Version 4.20 CIPS algorithm, together with an error analysis of individual cloud observations,  
 23 was presented in Lumpe et al. (2013). Here we describe only what is necessary to understand how IWC is derived from the  
 24 data. Even though an accurate determination of the scattering-angle dependence of radiance (often called the scattering phase  
 25 function) is obtained by seven independent measurements, this does not fully define the distribution of particle sizes. Instead,  
 26 additional constraints need to be introduced to derive the mean particle size. The particles are assumed to be the same oblate-  
 27 spheroidal shape as defined for the model calculations, and to have a Gaussian size distribution (see eq. 11 in Rapp and  
 28 Thomas, 2006). A relationship between the Gaussian width  $s$  and the mean particle radius  $r_m$  is derived from that found in  
 29 vertically-integrated lidar data (Baumgarten et al., 2010). The net result is that two parameters, the mean particle size and the  
 30 Gaussian width, are retrieved from a given scattering phase function. However, there is only one independent variable, since  
 31 the two are related by  $s(r_m)$ . Thus Eq. (3) simplifies to

$$IWC = \rho V(r_m) A(\Phi = 90^\circ, 0) / \sigma_\lambda(r_m, \Phi). \quad (7)$$

32  $V$  denotes the ice particle volume, averaged over the Gaussian distribution with a mean particle radius value  $r_m$ .  $A$  refers to  
 33 the retrieved albedo, corrected to view angle  $\theta = 0^\circ$  and interpolated to scattering angle  $\Phi = 90^\circ$ . Note the resemblance of  
 34 Eq. (7) to Eq. (5).  $A(\Phi = 90^\circ, 0)$ , along with  $r_m$  and  $IWC$  are products reported in the CIPS PMC data base, found at  
 35 (<http://lasp.colorado.edu/aim/>).  $\sigma_\lambda(r_m, \Phi = 90^\circ)$  is the mean scattering cross-section, integrated over the assumed  
 36 Gaussian distribution with mean radius  $r_m$  and distribution width  $s$ .

37

38 Before discussing the AIR results, we first apply the CIPS algorithm to the model data to test how well it works on a set of  
 39 realistic particle sizes. As mentioned earlier, UV measurements of ice particles are not sensitive to particle radii  $< 20$ - $25$  nm.  
 40 We applied the CIPS algorithm to 6672 model clouds, using seven scattering-angle points, spanning the range  $50^\circ$ - $150^\circ$  (the  
 41 results are insensitive to the values chosen). We then calculated the % difference between the exact model calculation of  
 42 IWC and the simulated CIPS retrieved IWC for every model cloud. Figure 3 shows the result as a function of  $A(\Phi=90^\circ)$ .  
 43 Assuming the microphysical model is accurate, the accuracy of the CIPS UV measurements ranges from over  $+100\%$  for  
 44 very small albedo to  $-60\%$  for high albedos. We emphasize that this is not an AIR result, but is an attempt to assess how  
 45 particles that are too small to be visible to UV measurements affect the accuracy of the CIPS IWC results. The mean  
 46 difference and standard deviation for the (albedo) bin-averages for two model days is  $-13 \pm 17\%$ . With the *caveat* that not all  
 47 ice is retrieved, a large subset of CIPS IWC data thus has an acceptable accuracy (an average of  $84\%$  of the modelled ice

48 mass is contained in particles with radii exceeding 23 nm). We note that IWC in the model used to derive the AIR  
49 approximation refers to *all* particle sizes.

50 The procedure for deriving AIR coefficients from the CIPS data is as follows: (1) Regression coefficients were derived from  
51 data pertaining to 0-40 days from summer solstice (day from solstice, DFS=0 to 40) on every third orbit. This meant that  
52 ~200 orbits per season were used. The regression analysis was performed on four years of data (2010-2013). The data were  
53 binned in 5-degree SA bins and only the best quality pixels with six or more points in the phase function were used; (2) data  
54 from each northern and southern summer season were treated separately. The coefficients and standard deviations of the fit  
55 were then interpolated to a finer SA grid from 22° to 180° in increments of 1°; (3) The coefficients from each hemisphere  
56 were averaged, and these coefficients were then used to create an AIR IWC data base, to accompany the normal CIPS  
57 products. As previously shown, the AIR data applies to the ice mass of ‘UV-visible’ clouds, not to their total ice mass.

58 We emphasize that using the AIR data is unnecessary for seasons prior to the northern summer season of 2016 – however the  
59 AIR data have great importance since that time because the observing mode was changed, resulting in measured phase  
60 functions that contain many fewer (and often only one) scattering angles. As illustrated in Fig. 4, it is trivial to infer both  
61 IWC and  $A(90^\circ)$  from a single measurement of albedo. This alternative 90-deg albedo value, ALB\_AIR, is now included  
62 along with IWC AIR in the CIPS Level 2 data files. Fig. 5 shows the AIR results for monthly-averaged IWC (July and  
63 January) compared to the same averages of the more accurate results from the operational (OP) retrieval described in Lumpe  
64 et al. (2013). The data have been separated into different hemispheres, and into ascending and descending nodes of the sun-  
65 synchronous orbit, and apply to the years of the nominal operating mode. The ALB\_AIR results are systematically higher  
66 than the operationally retrieved 90-deg albedo, whereas there is no consistent bias in the IWC (AIR) value compared to the  
67 operational product. However, for both quantities the interannual changes between the AIR and OP results agree very well.  
68 This is reflected in the very high correlation coefficients of the two sets of values. A more stringent test of the AIR method  
69 comes from daily values of CIPS IWC. Shown in Figs. 6 and 7 are polar projections of IWC (AIR) and the more accurate  
70 operational IWC data product. These ‘daily daisies’ are taken from overlapping orbit strips pertaining to 28 June of two  
71 different years. Figure 6 shows data from 2012, when CIPS was still in normal mode. The AIR result shows remarkable  
72 agreement with the operational IWC data. By 2016 (see Figure 7) CIPS is in continuous imaging mode and the standard  
73 IWC retrieval is limited due to the scarcity of pixels with three or more scattering angles. Here the AIR approach is clearly  
74 superior and does a good job of filling in the polar region where CIPS detects high-albedo clouds. The differences in patterns  
75 are due primarily to variations of particle size, rather than errors in the AIR method.

76 AIR accuracy can also be tested in the study of latitudinal variations. Figure 8 compares daily-averaged IWC from the CIPS  
77 Level 3C data, for both the standard and AIR algorithms, for the Northern Hemisphere 2011 season. It is clear that AIR is

78 adequate even for 24-h averages. For example, it is capable of defining the beginning and ending of the PMC season, a  
79 metric that has valuable scientific value (e.g., Benze et al., 2012)

### 80 **2.3 AIR Results from SOFIE**

81 A third independent data set of IWC and particle size is available from the AIM SOFIE experiment. SOFIE provides very  
82 accurate values of IWC, through precise near-IR extinction measurements, independent of particle size. It assumes the same  
83 Gaussian distribution of particle sizes as CIPS, so that the reported value of mean particle radius  $r_m$  is consistently defined.  
84 SOFIE data are useful to investigate the extent to which the AIR approximation can be applied to an independent data set. To  
85 do so, it is necessary to calculate 265-nm albedo at various SA values, given the values of  $r_m$ , ice column density  $N$  from the  
86 data base, and the mean cross-section  $\sigma_\lambda(r_m, \Phi)$ . The latter quantity is averaged over the assumed Gaussian distribution.  
87 The equation for the albedo is

$$A_\lambda(\Phi, 0) = \sigma_\lambda(r_m, \Phi)N . \quad (8)$$

88 Given  $A_\lambda(\Phi, 0)$  and IWC for each PMC measurement (one occultation per orbit), we can once again perform regressions  
89 and find AIR coefficients for the SOFIE data set. The comparison of AIR results from all three data sets is shown in Fig. 9,  
90 where the constant term  $C$  is the y-intercept and  $S$  is the slope in the AIR regression,

$$IWC(AIR) = C(\Phi) + S(\Phi) * A(\Phi, 0) . \quad (9)$$

91  
92 Figure 10 displays the results from the three data sets, expressed as contour plots of AIR-derived IWC as functions of SA  
93 and Albedo. This comparison shows that the three sets of IWC resemble one another far better than would be anticipated  
94 from the AIR coefficients in Fig. 9, where the constant coefficient differs significantly between data sets. Since the result of  
95 the regression in yielding IWC is more significant than the coefficients themselves, the comparisons of Fig. 10 are the more  
96 appropriate diagnostic. The fact that the IWC derived from AIR is more accurate than would be expected from the differing  
97 coefficients is due to the fact that the errors of the constant and slope coefficients are anti-correlated. The agreement between  
98 the three results will be even better when taken over a large data set with variable SA and albedo. The comparisons of IWC  
99 from different satellite experiments as a function of year and hemisphere will be the subject of a separate publication.

00

01 Figure 11 shows that the regressions with AIM SOFIE data obey a linear relationship between IWC and albedo for IWC  
02  $<220 \text{ g-km}^{-2}$ , but for SA values  $<90^\circ$ , AIR overestimates IWC by up to 15%, depending upon the SA. For SA= $110^\circ$  the  
03 regressions are still linear up to  $300 \text{ g-km}^{-2}$ , values above which are seldom encountered in the data.

04

#### 05 **2.4 SBUV Data**

06 The AIR coefficients from the model have been used by DeLand and Thomas (2015) to derive mean IWC from SBUV data,  
07 which spans the largest time interval of any satellite data set (1979-present). The 273 nm wavelength used in the SBUV  
08 Version 3 analysis is sufficiently close to the effective wavelength of the broader passband of the CIPS cameras (Benze et  
09 al., 2009) that the same coefficients may be applied to both data sets. The accuracy of the average IWC results was estimated  
10 by removing half the data by random sampling from an entire season and comparing the two results. For a highly-populated  
11 region (more than 1000 clouds per season at latitudes higher than  $70^\circ$ ), the differences in IWC ranged between  $\pm 3\text{--}5 \text{ g-km}^{-2}$ ,  
12 thus can be considered typical systematic errors. For a less-populated region ( $50^\circ\text{--}64^\circ$  latitude) where there were many  
13 fewer clouds ( $<50$ ), the differences were larger,  $\pm 5\text{--}10 \text{ g-km}^{-2}$ . Even the larger errors are sufficiently small for  
14 intercomparison of SBUV and contemporaneous PMC measurements. Figure 12 shows a comparison of SBUV IWC, using  
15 the model AIR coefficients, to the results of a more accurate determination of IWC derived from particle size determinations  
16 using SBUV spectral information (Hervig and Stevens, 2014). The comparison is for data residuals from July averages over  
17 the time series 1979-2017. Given the different assumptions underlying the two data sets, the agreement is very good (with an  
18 rms difference of 3% for the residuals, and 5% for the actual values of  $\langle \text{IWC} \rangle$ ).

#### 19 **3. Effects of Mean Particle Size**

20 The AIR approximation is based on the notion that particle size effects can be ignored in retrieving IWC from albedo  
21 measurements, that is they contribute in a sense to the ‘noise’ of the measurement, which can be minimized by averaging. In  
22 fact, the particle size (or more accurately, the term  $r^3$ ) is a principal ‘driver’ of  $\langle \text{IWC} \rangle$  itself, so it is not obvious that  
23 particle size effects play a minor role in deriving IWC. The dependence of albedo on column density adequately captures this  
24 part of the variability (albedo is strictly linear in column density). The AIR slope term is  $\sim r^3 / \sigma_\lambda(r, \Phi)$  averaged over a  
25 distribution of particle sizes,  $r$ . The size dependence of the cross-section varies as a power of  $r$ , within two limits, the  
26 geometric-optics limit,  $r^2$ , and the small-particle (Rayleigh) limit,  $r^6$ . In the intermediate and realistic conditions of  
27 PMC, the exponent has an intermediate value. Fortunately, there is a ‘sweet spot’ (or better, a ‘sweet region’ of the  $r$ -  
28 domain) in which the  $r$ -dependence of  $\sigma_\lambda$  is  $\sim r^3$ , so that the slope term is constant (for fixed SA). This behavior occurs for  
29 all relevant values of SA, and for the albedo values typical of CIPS. It accounts mainly for the effectiveness of the AIR  
30 method. The other aspect favorable to AIR is the steep fall-off of the particle size distribution at the largest sizes, which  
31 contributes to the sharpness of the lower boundaries in the spread of points in Fig. 1. Averaging over many values of  $r$  results

32 in the AIR slope term that, in the limit of large number, the term depends predominantly on  $\Phi$ . This is an example of  
33 “regression to the mean”, and illustrates how the approximation is designed to work for large numbers of clouds. In a  
34 fictitious case where the mean cloud particle size is larger in one year than another, but the cloud column number remains  
35 the same, the mean albedo would increase according to Eq. (8), resulting in an increase of  $\langle IWC \rangle$ . We might expect that the  
36 slope term would be different in the two cases. Our study with three different data sets shows that the regression slope itself  
37 remains almost the same among the three data sets, despite their differing in mean particle size.

38

39 In fact, Hervig and Stevens (2014) found from SBUV spectral data a small long-term trend in  $\langle IWC \rangle$  and in addition a trend  
40 in the mean particle size ( $+0.23 \pm 0.16$  nm/decade). This contributed an additional 20% to the overall long-term trend in  
41  $\langle IWC \rangle$ . The ignored dependence on mean particle size using the AIR method thus adds a systematic uncertainty in derived  
42  $\langle IWC \rangle$  trends, which can be as large as 20%, according to their analysis. This error undoubtedly varies inversely with the  
43 number of clouds in the averaging process. For example, the number of CIPS observations per PMC season greatly exceeds  
44 that of SBUV, so that the error in  $\langle IWC \rangle$  should be correspondingly smaller.

45

#### 46 **4. Conclusions**

47 We have described the theoretical basis and accuracy for an approximation for retrieving the average ice water content  
48 (IWC) of Polar Mesospheric Clouds (PMC) from measurements of UV albedo at a single scattering angle. This approach  
49 provides a continuous set of consistent CIPS measurements of IWC from year to year, regardless of the number of scattering  
50 angles for which albedo at a single location is measured. The consistent AIR IWC data base enables robust IWC  
51 comparisons throughout the AIM mission, from 2007 to the present. A comparison of IWC derived from the microphysical  
52 model and from the CIPS algorithm suggests that CIPS is capable of measuring 84% of the total ice content of PMC (for  
53 particle sizes exceeding  $\sim 23$  nm). Assuming the microphysical model is accurate, the accuracy of the CIPS UV  
54 measurements ranges from over +100% for very small albedo to -60% for high albedos. The overall accuracy of IWC  
55 (averaging over all albedo bins) is  $-13 \pm 17\%$ . The CIPS algorithm overestimates the small-particle population (20-30 nm) as  
56 a result of the Gauss approximation when the mean particle size is small, and the opposite is true when the mean size is  
57 large. These errors are a result of the CIPS approximations and the invisibility of small particles, and are irrelevant to the  
58 AIR approximation.

59

60 Distinct from the more fundamental errors due to the invisibility of very small ice particles and the Gaussian approximation,  
61 we also estimated the errors in the AIR approximation, relative to the AIM SOFIE data which apply to larger values of IWC  
62 than the model. AIR is less accurate for high IWC ( $>220$  g-km<sup>-2</sup>), but very-high mass clouds ( $IWC > 300$  g-km<sup>-2</sup>) are  
63 infrequent and do not influence seasonal averages of IWC. For the dimmer and more frequent clouds, Fig. 2 shows that the

64 error in ensemble averages is of the order of 3%. The accuracy of the AIR results for ensemble averages has a small  
65 systematic dependence on mean particle size- the error depends inversely on the size of the ensemble. The inter-annual and  
66 hemispheric variations of IWC derived from CIPS and SBUV measurements throughout an entire 11-year period (2007-  
67 2018) will provide detailed information on PMC variability over the recent solar cycle 24.

68

### 69 **Figure Captions**

70 Figure 1. Linear regressions of model PMC albedo versus model PMC ice water content. The black points represent model  
71 clouds with  $r_{\text{eff}} < 40$  nm. The red points apply to  $r_{\text{eff}} > 40$  nm. The blue line is the linear least-squares fit to all points. (a)  
72 through (f) are for different scattering angles. The lower limit for the albedo ( $SA=90^\circ$ ) is 1G, which is the detection limit of  
73 the CIPS experiment.

74

75 Figure 2. Relative errors of ensemble averages,  $\langle IWC \rangle$  using the AIR approximation, taken over all cloud model simulations  
76 for conditions of summer solstice.  $\langle IWC \rangle$  is ‘thresholded’ by the variable IWC in the vertical axis, so that  $\langle IWC \rangle$  applies to  
77 all values above IWC. Contour lines are labelled as percent errors relative to the accurate model values.

78

79 Figure 3. Differences of IWC derived from the model cloud ‘data’ and the accurate IWC from the model, plotted against the  
80 265-nm albedo (in G units, see text), evaluated at  $SA=90^\circ$ . The error bars are the standard deviations in intervals of 2G.

81 Figure 4. Illustration showing how  $IWC=98$  g-km<sup>-2</sup> (horizontal arrow) and  $A(90^\circ) = 16$  G (thick downward arrow) are  
82 derived from the AIR method from a single measurement of cloud albedo at 60 G and  $SA=50^\circ$  (upward arrow). Each  
83 straight-line plot is calculated from Eq. (9).

84

85 Figure 5. Comparison of CIPS  $A(90^\circ)$  (top) and  $\langle IWC \rangle$  (bottom) calculated from the operational (OP) and AIR algorithms.  
86 Data points correspond to July northern hemisphere (NH) and January southern hemisphere (SH) averages in a 5-degree  
87 latitude bin centered at  $70^\circ$ . Left and right panels are for ascending and descending legs data, respectively.

88

89 Figure 6. Polar projection map of IWC from CIPS, Day 180 (28 June 2012). Left and rights panels show the operational  
90 IWC product and the AIR result, respectively.

91

92 Figure 7. Same as Fig. 6 except for 28 June 2016.

93

94 Figure 8. Filled circles and dotted line: IWC (AIR) averaged over 1-deg latitude bins centered on  $70^\circ$  (green) and  $80^\circ$  (blue),  
95 and over 15 orbits (from which daily averages are derived). Solid line: standard L3C IWC averaged in the same way.

96 Figure 9. AIR coefficients for three different sources of IWC and particle size: Model (solid line with open circles), CIPS  
97 (solid line), and SOFIE (dashed line).

98

99 Figure 10. Contour plots of the AIR approximations for IWC versus cloud albedo ( $G$ ) for the three data sources: (a) model,  
00 (b) SOFIE, and (c) CIPS

01

02 Figure 11. Examples of SOFIE AIR regressions for two (specified) scattering angles, (a)  $80^\circ$  and (b)  $110^\circ$ .

03

04 Figure 12. Comparison of annually-averaged northern hemisphere July-averaged residuals ( $\langle IWC \rangle - \text{long-term mean}$ )  
05 derived by two independent methods from SBUV 273 nm albedo data. Black curve:  $\langle IWC \rangle$  derived from the AIR  
06 approximation. Blue curve:  $\langle IWC \rangle$  derived from the same SBUV albedo data, but including mean particle size variations  
07 (see text). A three- year smoothing has also been applied.

08

### 09 **Data Availability**

10 The CIPS operational PMC data, along with the AIR data, can be found at <http://lasp.colorado.edu/aim/>.

11

### 12 **Author Contribution**

13 Author G.T. formulated the AIR approximation, and derived the AIR coefficients from the microphysical model (provided  
14 by author C.B.), and from the AIM SOFIE data (<http://sofie.gats-inc.com/sofie/index.php>). Authors J.L. and C.R. calculated  
15 the AIR coefficients from the CIPS data.

16

### 17 **Disclaimer**

18 The authors declare that they have no conflict of interest.

19

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Albedo-Ice Regression, SD-WACCM/CARMA microphysical model

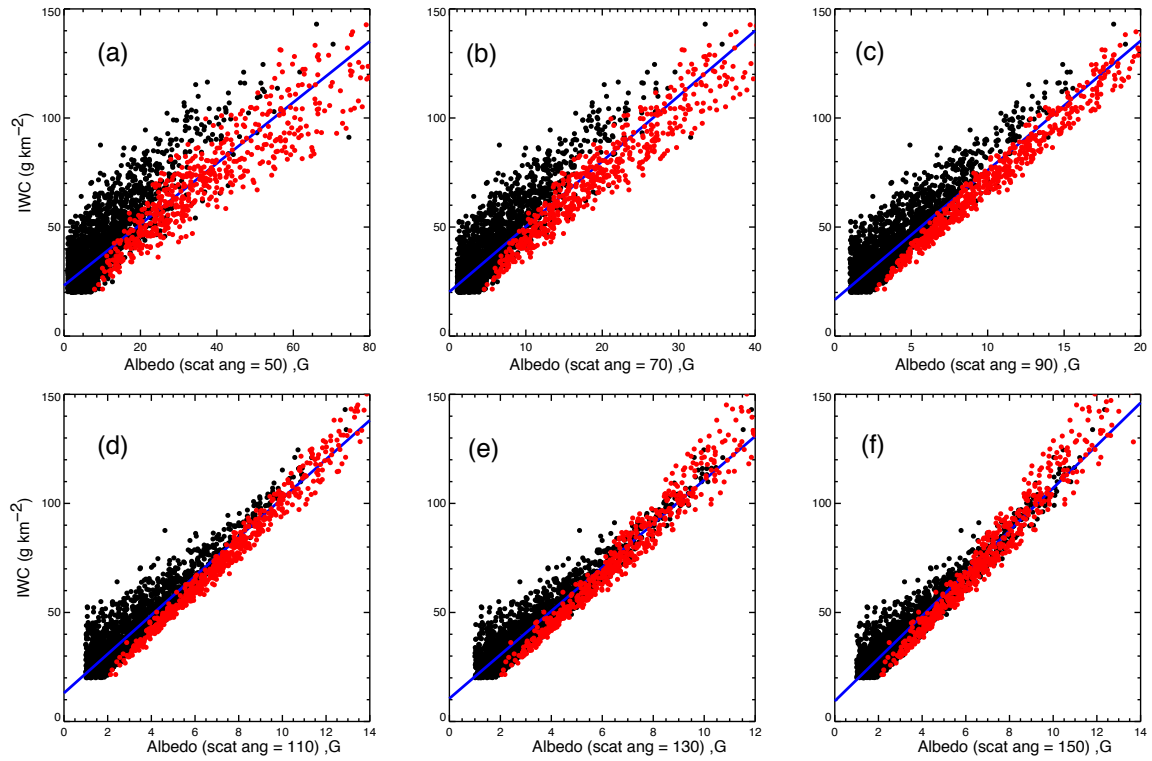


Figure 1

### AIR Error in $\langle \text{IWC} \rangle$ : Model

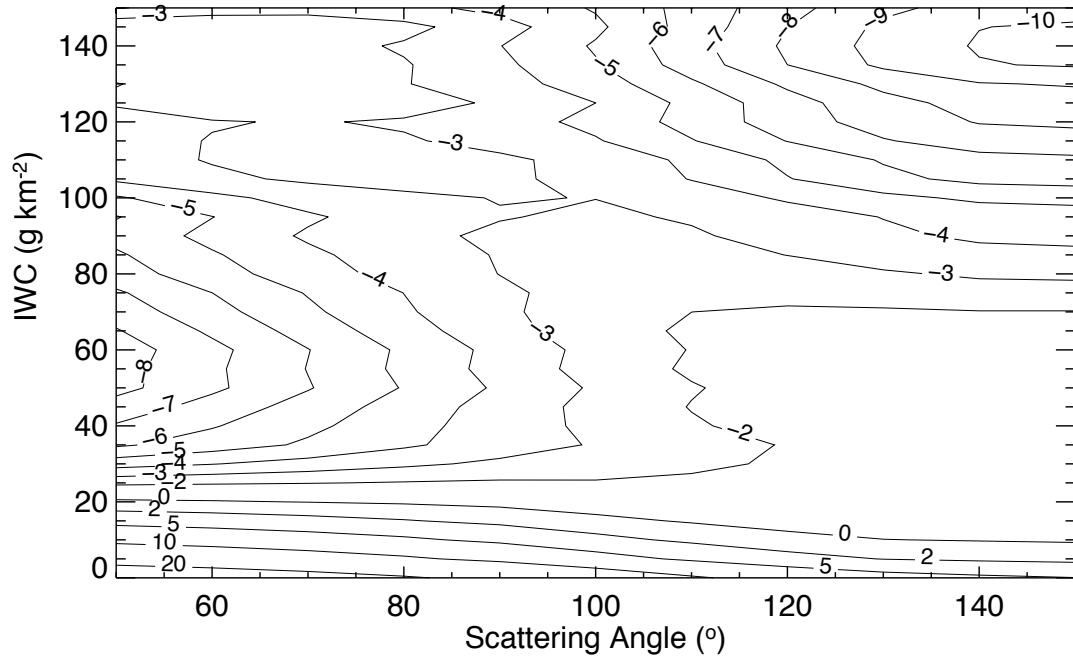


Figure 2

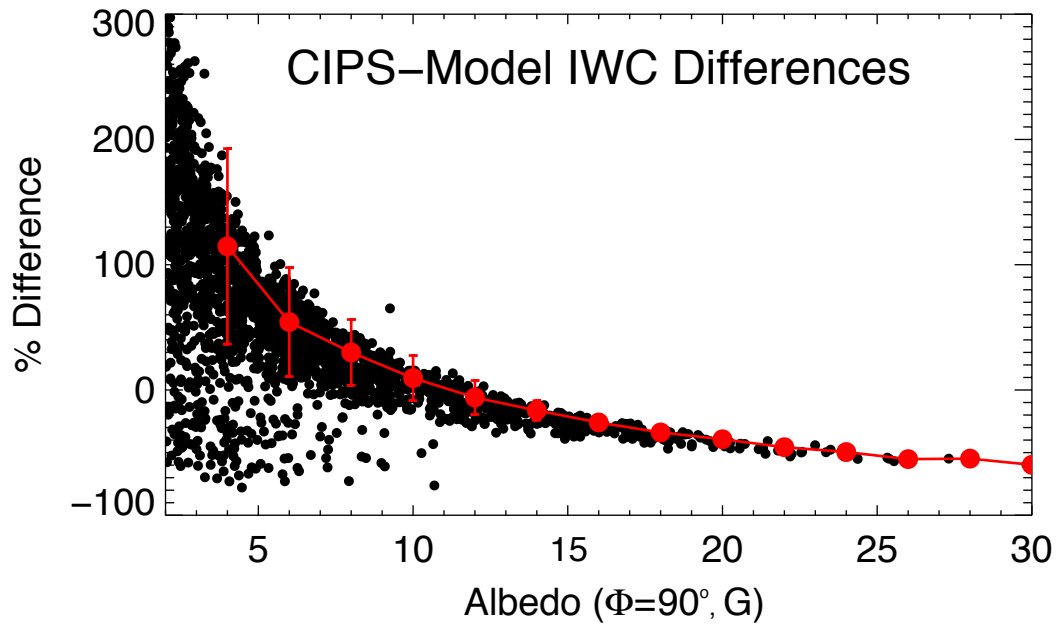


Figure 3

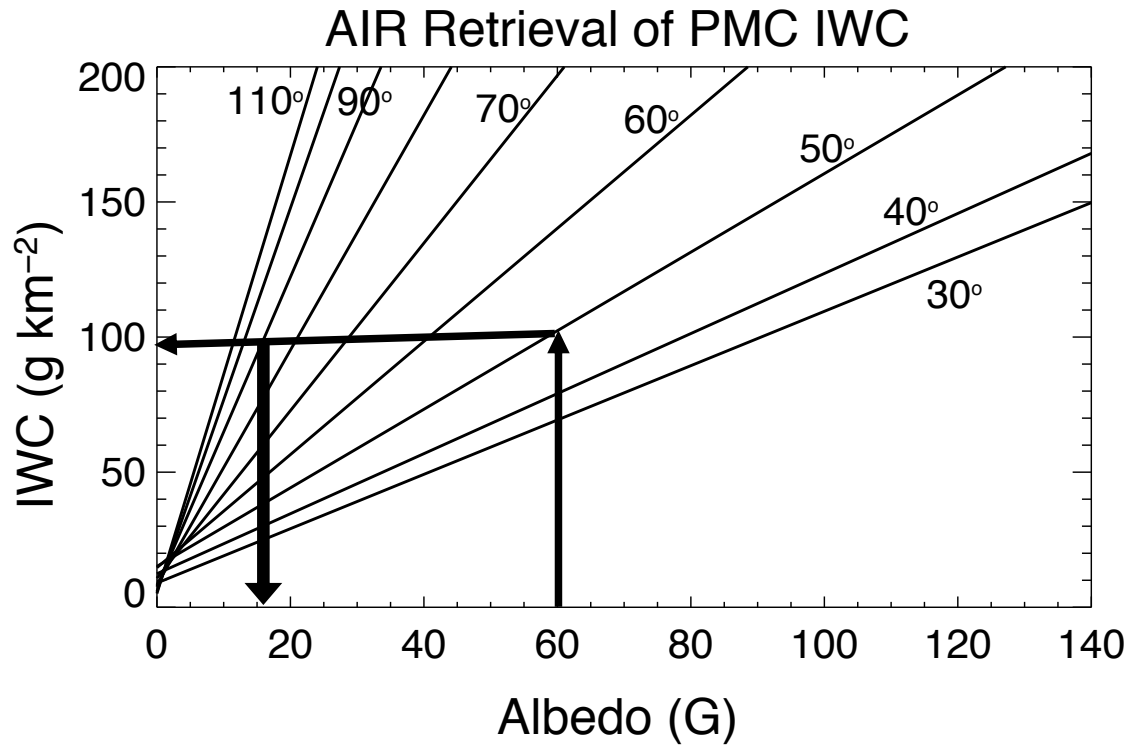


Figure 4

Comparison of OP retrieval to AIR: DFS=[10,40] ; Lat = 70°

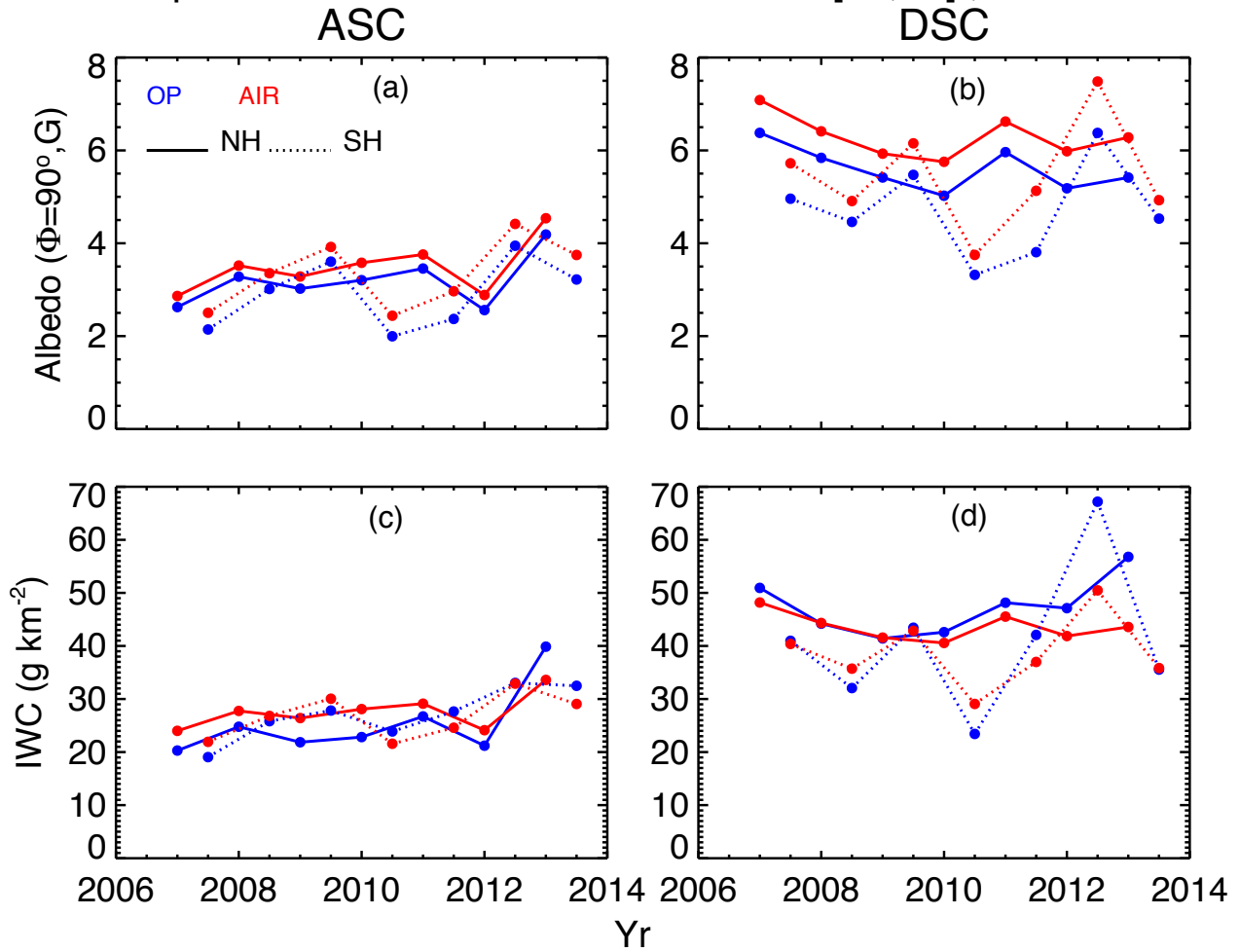
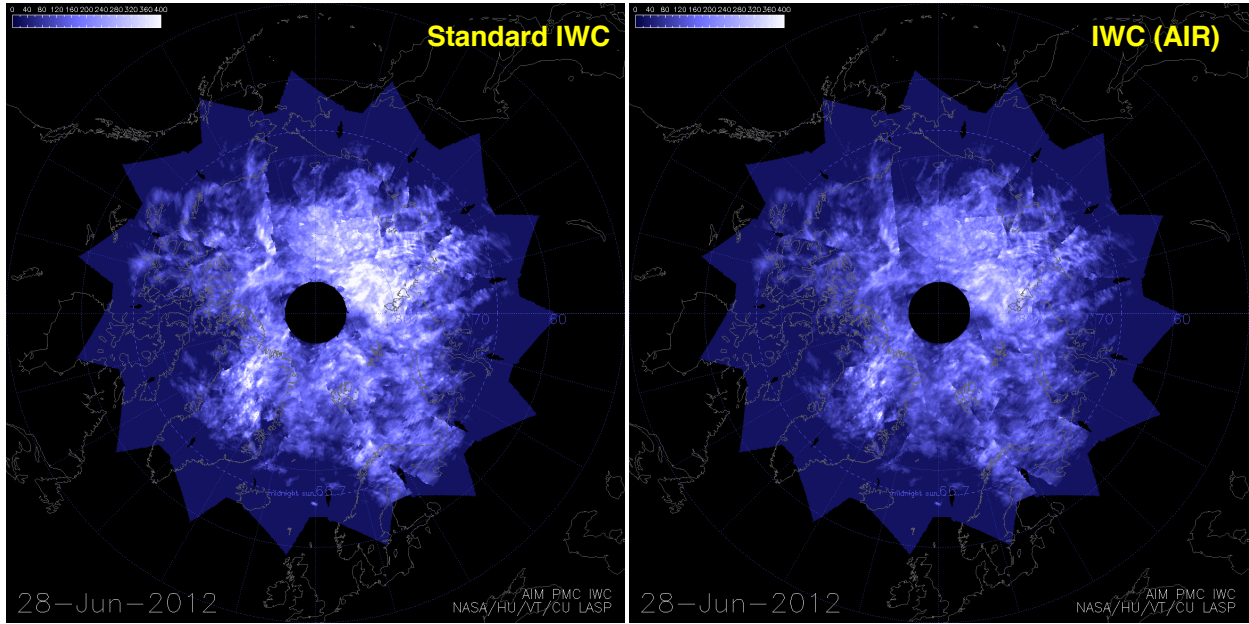


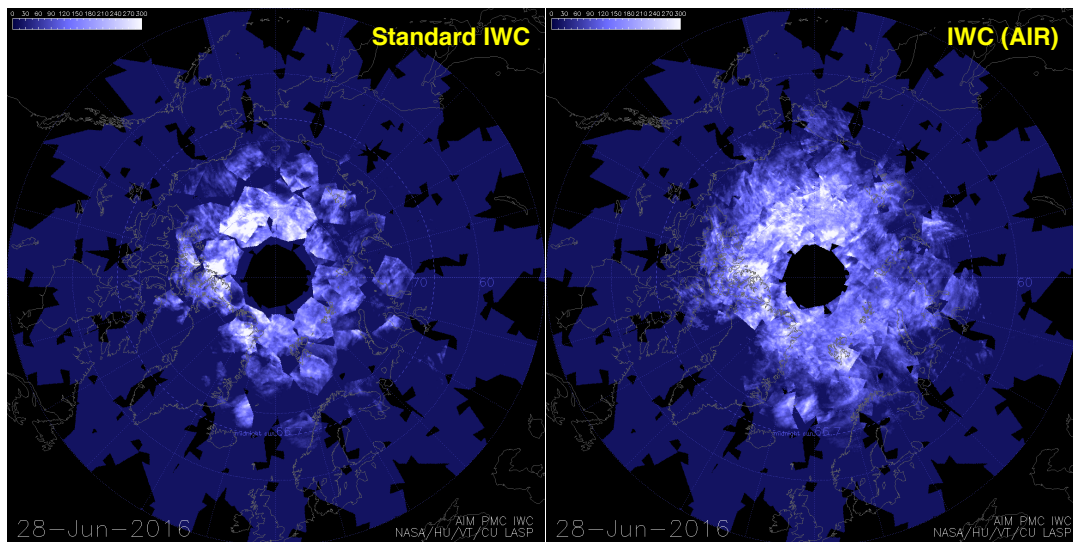
Figure 5



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Figure 6

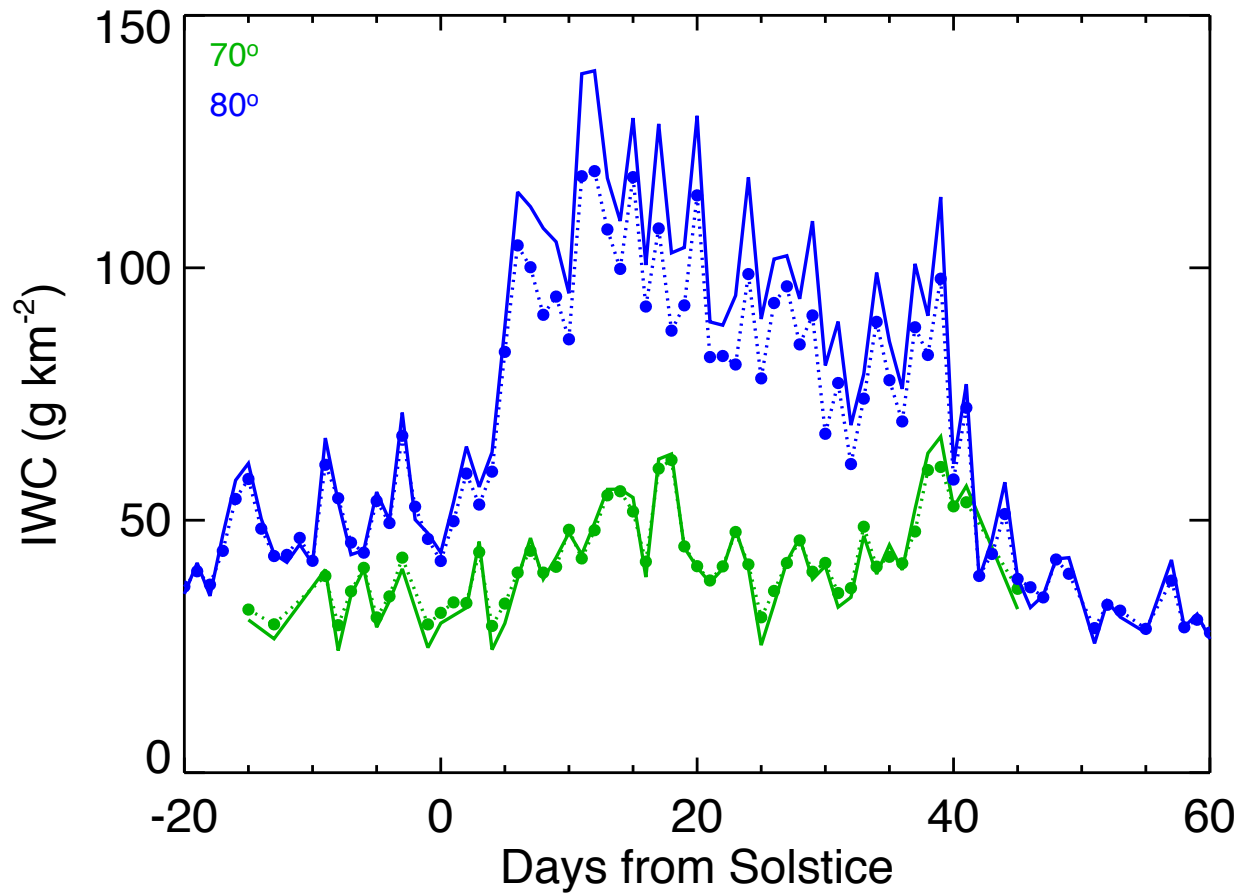




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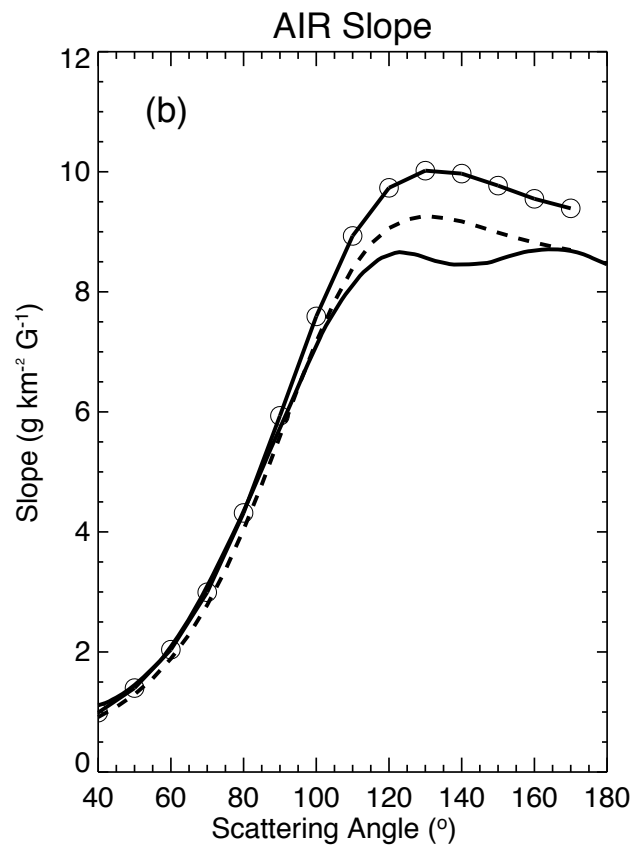
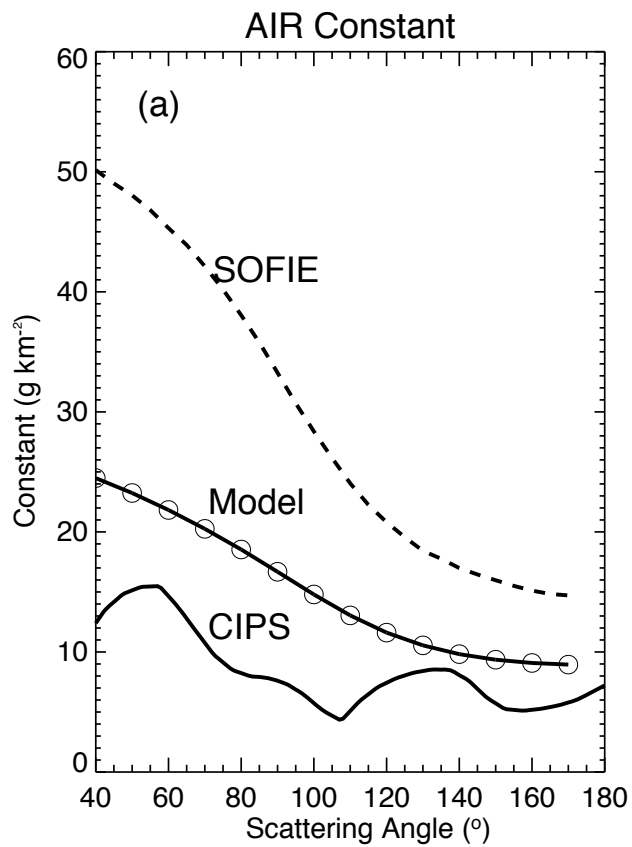
Figure 7

# CIPS Level 3C IWC - NH 2011



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Figure 8



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Figure 9

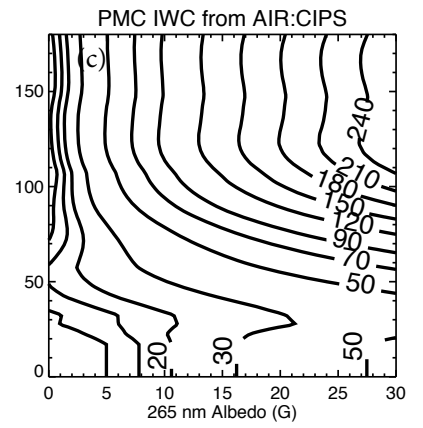
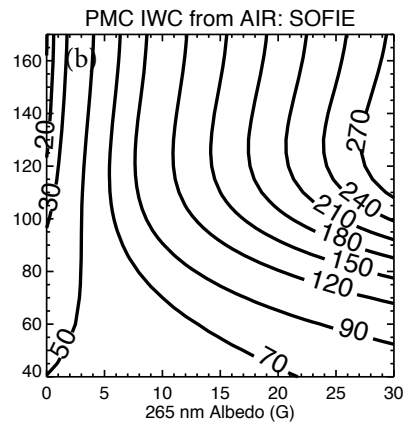
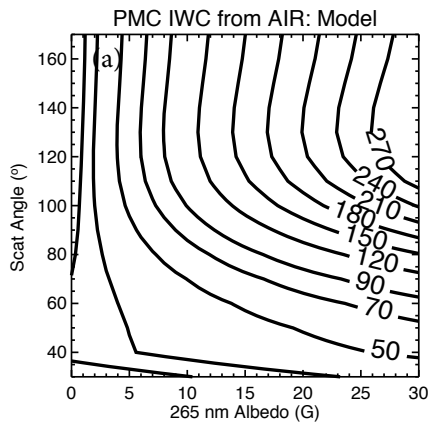


Figure 10

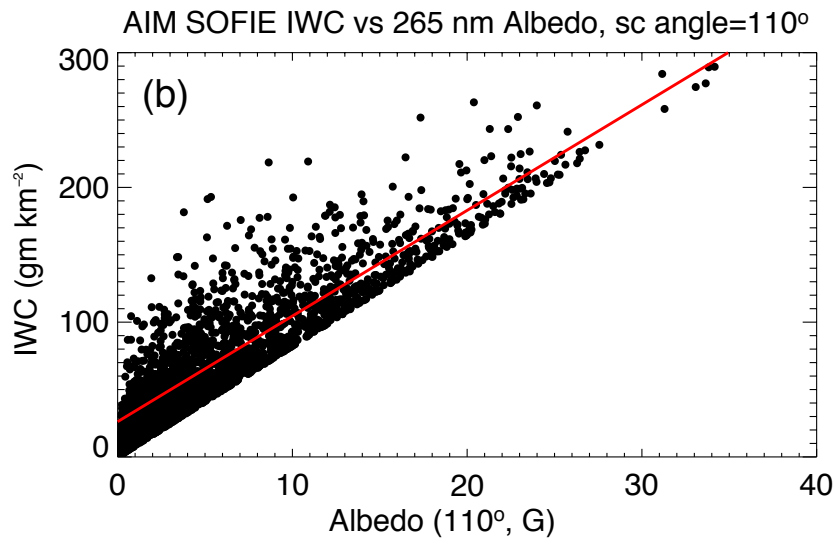
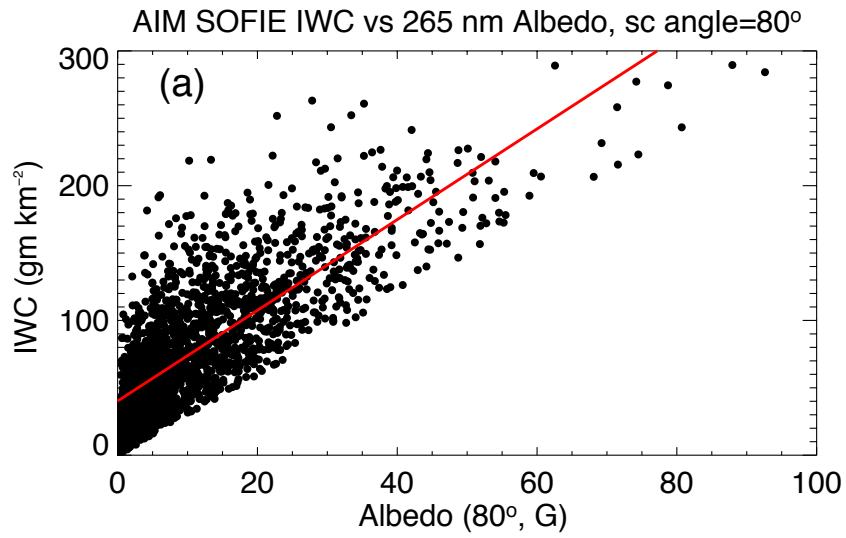
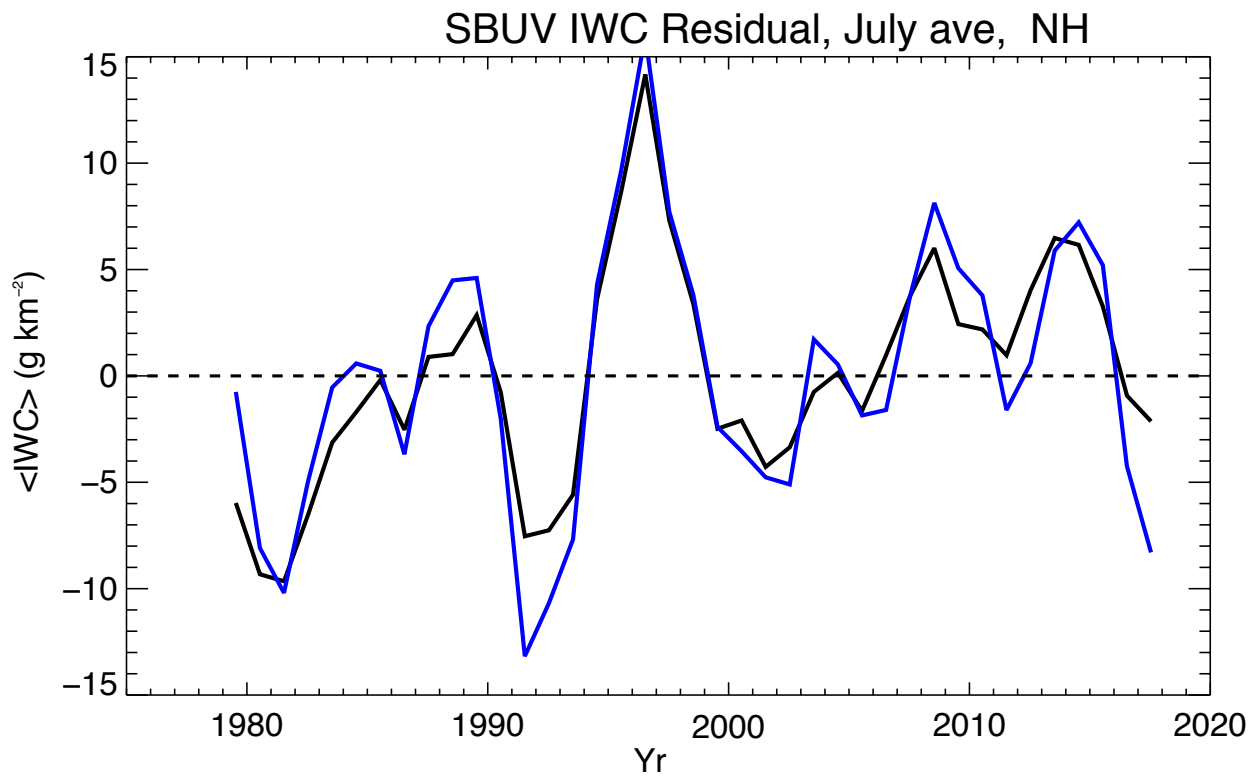


Figure 11



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Figure 12