

1 Evaluating the Consistency and Continuity of Pixel-Scale Cloud

2 Property Data Records From *Aqua* and *SNPP*

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14 Abstract

15 The *Aqua*, *SNPP*, and *JPSS* satellites carry a combination of hyperspectral infrared sounders (AIRS,
16 CrIS) and high-spatial-resolution narrowband imagers (MODIS, VIIRS). They provide an
17 opportunity to acquire high-quality long-term cloud data records and are a key component of the
18 existing Program of Record of cloud observations. By matching observations from sounders and
19 imagers across different platforms at pixel scale, this study evaluates the self-consistency and
20 continuity of cloud retrievals from *Aqua* and *SNPP* by multiple algorithms, including the AIRS
21 Version-7 retrieval algorithm and the Community Long-term Infrared Microwave Combined
22 Atmospheric Product System (CLIMCAPS) Version-2 for sounders, and the Standard
23 *Aqua*MODIS Collection-6.1 and the NASA MODIS-VIIRS continuity cloud products for imagers.
24 Metrics describing detailed statistical distributions at sounder field of view (FOV) and the joint
25 histograms of cloud properties are evaluated. These products are found highly consistent despite
26 their retrieval from different sensors using different algorithms. Differences between the two
27 sounder cloud products are mainly due to cloud clearing and treatment of clouds in scenes with
28 unsuccessful atmospheric profile retrievals. The sounder subpixel cloud heterogeneity evaluated
29 using the standard deviation of imager retrievals at sounder FOV shows good agreement between
30 the standard and continuity products from different satellites. However, impact of algorithm and
31 instrument differences between MODIS and VIIRS is revealed in cloud top pressure retrievals and
32 in the imager cloud distribution skewness. Our study presents a unique aspect to examine NASA's
33 progress toward building a continuous cloud data record with sufficient quality to investigate
34 clouds' role in global environmental change.

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39 **1. Introduction**

40 Clouds play an important role in Earth’s energy balance and hydrological cycle. They occur
41 with processes involving atmospheric radiation, thermodynamics, and dynamics at various spatial
42 and temporal scales, making clouds a crucial component of the weather and climate system. With
43 daily regional and global coverage, space observations provide a unique vantage point to monitor
44 the change of the cloud properties in the climate system across different time scales. This offers an
45 important observational basis to resolve cloud processes in the background atmospheric circulation,
46 which is widely recognized as a critical challenge within Earth Sciences (Bony et al. 2015, IPCC
47 2013). The 2017 US National Academy Decadal Survey (ESAS 2017) has noted the importance
48 of long-term and sustained observations of many key components of the Earth system, including
49 continuity measurements of clouds. Many of these observations are obtained from the existing
50 Program of Record (POR). Since the “POR forms the foundation upon which the committee’s
51 recommendations are established” (ESAS 2017), it is crucial to evaluate whether a self-consistent
52 and continuous POR for cloud-related variables is indeed available with sufficient data quality and
53 spatio-temporal coverage.

54 Cloud retrievals from the NASA’s Earth Observing System (EOS) satellites, including *Terra*
55 and *Aqua*, the joint NASA/NOAA Suomi National Polar-orbiting Partnership (*SNPP*), and
56 NOAA’s new generation of Joint Polar Satellite System (*JPSS*) series weather satellites, are a key
57 component in the POR for cloud properties. Through efforts on continuity and consistency by
58 rigorous instrument mission design and ongoing algorithm development, these satellites provide
59 high quality, long-term cloud data records derived from the Top of Atmosphere (TOA) radiances
60 observed across a wide range of the emission and reflection spectrum. Particularly, *Aqua*, *SNPP*,

61 and *JPSS-1* (now *NOAA-20*), which were launched in 2002, 2011, and 2016, respectively, carry
62 high spatial resolution narrowband imagers, hyperspectral infrared (IR) sounders, and microwave
63 (MW) sounding measurements. As a result, observations with similar spatial resolution and
64 coverage, and similar spectral resolution at analogous wavelengths are obtained from different
65 satellites. For *Aqua*, this instrument trio consists of the Atmospheric Infrared Sounder (AIRS), the
66 Advanced Microwave Sounding Unit (AMSU), and the Moderate Resolution Imaging
67 Spectroradiometer (MODIS). For *SNPP* and *JPSS*, the trio includes the Cross-track Infrared
68 Sounder (CrIS), the Advanced Technology Microwave Sounder (ATMS), and the Visible Infrared
69 Imaging Radiometer Suite (VIIRS).

70 Retrieval algorithms to maintain the continuity of the data records across these platforms have
71 been developed. For joint retrievals by IR and MW sounders such as AIRS/AMSU and CrIS/ATMS,
72 the Community Long-term Infrared Microwave Combined Atmospheric Product System
73 (CLIMCAPS; Smith and Barnet, 2019) provides cloud properties together with vertical profiles of
74 atmospheric temperature, water vapor, and trace gases, as well as surface conditions. For imagers
75 like MODIS and VIIRS, the NASA MODIS-VIIRS continuity cloud products have been developed
76 for both cloud mask (CLDMSK; Frey et al. 2020) and cloud optical properties (CLDPROP; Platnick
77 et al. 2021). These continuity algorithms have heritage with NASA operational retrieval products
78 previously developed for individual sensors and satellites, such as the AIRS Science Team retrieval
79 algorithm Version 7 (AIRS V7, Yue and Lambrigsten 2017, 2020) in the case of CLIMCAPS, and
80 the Standard *Terra/Aqua* MODIS Collection 6.1 cloud retrievals
81 (MOD35/MYD35, MOD06/MYD06; Baum et al. 2012, Platnick et al. 2017) in the case of MODIS-
82 VIIRS. However, significant differences exist between the standard and continuity algorithms, as
83 the focus of the continuity algorithms is to minimize the impact of instrument between platforms.

84 The sounder-imager combination on the same sun-synchronous polar-orbiting satellite,
85 together with the temporal coverage overlap between satellites, provides opportunities to utilizing
86 spectral and spatial capabilities from different sensors at global scale. Previous studies have shown
87 the benefits of using the combined information to intercalibrate and test radiometric consistency
88 among sensors (Tobin et al. 2006, Schreier et al. 2010, Wong et al. 2015, Gong et al. 2018);
89 crossvalidate the retrievals (Nasiri et al. 2011, Kahn et al. 2014); further improve atmospheric and
90 surface geophysical parameter retrievals (Irion et al. 2018, Yao et al. 2015); provide simultaneous
91 observations to resolve complex physical processes (Yue et al. 2013, 2016, 2019, McCoy et al.
92 2017); quantify the subpixel heterogeneity (Li et al. 2004, Kahn et al. 2015); and enhance the
93 utilization of satellite observations in numerical weather prediction and climate models (Eresmaa
94 2014). Therefore, the sounder-imager combination is an important aspect of data record continuity
95 and consistency among sensors across different platforms. This helps provide robust monitoring
96 of long-term changes in cloud properties, an important capability expected from the POR.

97 Pixel-scale analyses are an effective and unique way to investigate the consistency and
98 continuity of these data records because of the one-to-one relationships established by these
99 comparisons and their direct links to algorithm performance. This includes examining differences
100 of (1) the same physical parameters observed by different sensors or satellites but processed using
101 the same (or similar) algorithms, and (2) the same parameters obtained from the same sensor but
102 from different algorithms. Both of these differences are quantified at the pixel scale in this study.
103 The cloud properties determined by the sounder and imager pairs on board *Aqua* and *SNPP*, namely
104 AIRS/MODIS and CrIS/VIIRS, are investigated using the collocated sounder-imager fields of
105 view (FOVs) for sets of pixels obtained during Simultaneous Nadir Observations (SNOs) between
106 *Aqua*-AIRS and *SNPP*-CrIS. This approach ensures nearly identical viewing geometry by the two

107 satellites while pixel-scale cloud assessment is carried out by comparing cloud parameters
108 determined by hyperspectral IR sounders and high spatial resolution imagers at the minimum
109 spatial scale of individual instrument fields of view. Using this approach, products from both the
110 heritage NASA standard retrieval algorithms and the newly-developed continuity cloud algorithms
111 are analyzed (Table 1). This is essential for retrieval algorithm development and crossvalidation of
112 multiple sensors and products on *Aqua* and *SNPP*, and also important for data continuity extending
113 to future *JPSS* satellites. ▼

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114 This article is organized as follows. Section 2 describes various cloud products and their
115 retrieval algorithms analyzed in this study, as well as the method used to create pixel-scale
116 collocated datasets between sounders and imagers across different satellites. Section 3 shows the
117 detailed comparisons of cloud properties and their joint histograms from different algorithms and
118 sensors, and the discussions on implications on retrieval algorithm development and instrument
119 differences. A summary and set of conclusions are presented in Section 4.

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121 2. Data and Methodology

122 2.1 Cloud products and algorithms

123 Table 1 summarizes the cloud parameters analyzed in this study from various Level 2 (L2)
124 retrieval products derived from the sounders and imagers aboard *Aqua* and *SNPP*. For AIRS and
125 MODIS, both the standard operational and continuity products are evaluated: the AIRS V7 and
126 CLIMCAPS-*Aqua* Version 2 (V2) retrievals for AIRS, and the Collection 6.1 *Aqua* MODIS
127 Atmosphere Level 2 Cloud Product (MYD06) and Version 1.1 NASA *Aqua* MODIS Continuity
128 Cloud Property Products (CLDPROP_MODIS). For *SNPP*-CrIS and -VIIRS, only the continuity
129 products are evaluated, which are the V2 CLIMCAPS-*SNPP* and Version 1.1 *SNPP*-VIIRS

Continuity Cloud Property Products (CLDPROP_VIIRS). The CLIMCAPS-*SNPP* products were produced using Version 2 of the CrIS Level-1B product in Nominal Spectral Resolution (NSR) and Full Spectral Resolution (FSR), which differ in the spectral resolution of the shortwave and mid-IR CrIS observations transmitted from *SNPP* (Monarrez et al. 2020). The spectral resolution differences cause subtle differences between the CLIMCAPS FSR and NSR retrievals, especially in the upper tropospheric humidity and trace gases (Wang et al. 2021).

In both the AIRS V7 and CLIMCAPS algorithms for AIRS and CrIS, the radiatively effective cloud amount (effective cloud fraction, ECF) and cloud top pressure (CTP) are retrieved by matching the calculated cloudy radiances with the observed radiances for a set of channels that are sensitive to clouds. Then the cloud top temperature (CTT) is derived as the atmospheric temperature matching the retrieved CTP. In this process, best estimates of surface and atmospheric parameters are used to calculate the cloudy radiances, either from the *a priori* state or from the physical retrieval after the cloud clearing step (Susskind et al. 2003, Susskind et al. 2006, Smith and Barnett 2019). The cloud clearing approach (Chahine 1974) is applied in both the AIRS Science Team algorithms and CLIMCAPS. It predicts a single cloud cleared radiance at one AMSU or ATMS field of regard (FOR) using *a priori* temperature, water vapor, and surface information and a linear combination of IR radiances from nine AIRS or CrIS FOVs that are co-registered with one AMSU or ATMS FOR (Susskind et al. 2003). The cloud cleared radiances are subsequently used to retrieve surface and atmospheric parameters. Flowcharts of the retrieval steps and differences in these two sounder retrieval systems are given in Thrastarson et al. (2021).

The ECF is the product of cloud areal fraction and the IR cloud emissivity, the latter of which is assumed to be spectrally flat in the retrieval of ECF (Susskind et al. 2003). Previous studies show

154 that the AIRS ECF is consistent with the cloud properties such as the cloud frequency and cloud
155 optical depth measured by CloudSat and MODIS (Yue et al. 2011, Kahn et al. 2014). The AIRS
156 and CrIS retrievals of ECF and cloud top properties (CTT and CTP) are reported for up to two
157 cloud layers in each IR sounder FOV (~13.5 km spatial resolution at nadir).

158 There are distinct differences between the AIRS V7 and CLIMCAPS V2 algorithms regarding
159 cloud retrievals, summarized here. The first major difference is how cloud clearing is iterated in
160 the retrieval flow. The second major algorithm difference is quality control (QC) procedures when
161 1) the physical retrieval of atmosphere and surface is not successful, and 2) the final-stage cloud
162 clearing is not successful (Susskind et al. 2014). The third major difference is the choice of the
163 prior states for the two algorithms. The AIRS Science Team algorithms, including both V6 and V7,
164 iterate cloud clearing multiple times, and cloud parameters are determined after the last iteration
165 of cloud clearing using the retrieved surface and atmospheric conditions (Fetzer et al. 2020). In
166 contrast, CLIMCAPS V2 performs a single cloud clearing pass and cloud properties are retrieved
167 using the surface and atmospheric parameters from successful retrievals of surface and atmospheric
168 properties (Smith and Barnet 2019, Thrastarson et al. 2021). The QC procedure used in the two
169 sounder cloud retrievals are also different. AIRS V7 produces case-by-case QC indicators for each
170 retrieved variable; while CLIMCAPS V2 derives one QC value based on the cloud clearing and
171 retrieval status of temperature and water vapor, and the same QC value is assigned to all retrieved
172 variables for the given FOV, including the cloud parameters. Particularly, in AIRS V7 cloud
173 retrieval process, the final stage of cloud clearing and cloud retrievals uses the surface and
174 atmospheric variable retrievals, except for cases over ocean when the retrieved surface temperature
175 differs from the first guess by more than 5 K. For these cases, the surface temperature and surface
176 emissivity from the *a priori* are used instead, and cloud properties retrieved under this condition

177 are flagged as valid with QC=1, indicating successful cloud retrievals but potentially higher
178 uncertainty than QC=0. This surface test effectively filters out cases when the cloud top is
179 misidentified as surface and causes extremely small ECF values for overcast cloudy conditions
180 over ocean. For ~1% of cases the final cloud retrieval step does not complete successfully, and a
181 QC=2 flag is assigned to cloud parameters to indicate invalid retrievals. As a result, the AIRS V7
182 cloud retrievals produce a much higher percentage of cases with successful cloud retrievals (cloud
183 variable QC=0 or QC=1) than its temperature and water vapor profile products. For CLIMCAPS
184 V2, cloud clearing is not iterated and cloud parameters follow the QC procedure in the physical
185 atmospheric state retrievals. As a result, a much larger number of cases with QC=2 cloud retrievals
186 are reported by CLIMCAPS V2 compared to AIRS V7, especially for cloudier conditions or cases
187 with large cloud clearing errors, typically those FORs with low cloud contrast between associated
188 FOVs. Different *a priori* in the two retrieval systems impact their cloud retrievals. AIRS V7 uses
189 the Stochastic Cloud Clearing / Neural Network (SCCNN) solution as *a priori* on atmospheric
190 temperature and water vapor profiles and surface temperature trained using a few months of
191 AIRS/AMSU radiances and European Center for Medium-Range Weather Forecasting (ECMWF)
192 Integrated Forecast System (IFS) 3-hourly forecast fields that are collocated to AIRS observations
193 (including updates since Version CY31R1: [https://www.ecmwf.int/en/forecasts/documentation-](https://www.ecmwf.int/en/forecasts/documentation-and-support/changes-ecmwf-model)
194 [and-support/changes-ecmwf-model](https://www.ecmwf.int/en/forecasts/documentation-and-support/changes-ecmwf-model)) (Milstein and Blackwell 2016). For land and sea ice surface
195 emissivity prior estimates, AIRS V7 uses the University of Wisconsin – Madison Baseline Fit
196 Emissivity database (Seemann et al. 2008), which is based on the monthly climatology of MODIS
197 land surface emissivity product (MOD11) in 2008 (Thrastarson et al. 2021). The CLIMCAPS
198 system (Smith and Barnet 2020, Smith et al. 2021), instead, uses concurrent fields from the Version
199 2 ModernEra Retrospective analysis for Research and Application (MERRA-2, Gelaro et al. 2017)

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204 as the *a priori* and implements the Combined ASTER (Advanced Spaceborne Thermal Emission
205 and Reflection Radiometer) and MODIS Emissivity database for land surface (Hook 2019, [Borbas
206 et al., 2018, Feltz et al. 2018](#)). Over ocean, both systems use the Masuda IR sea surface emissivity
207 model (Masuda et al., 1988) as modified by Wu and Smith (1997). Since the *a priori* temperature,
208 water vapor, and surface properties are used in the cloud clearing step, differences in the *a priori*
209 contribute to the differences between the retrieval products, including cloud properties (Yue and
210 Lambrigtsen 2020, Yue et al. 2021). Cloud clearing plays an important role in both retrieval
211 systems, and physical retrievals of surface and atmospheric parameters are obtained from the cloud
212 cleared radiances, which, in turn, impact the determination of cloud properties. ▲

213 In addition to these major differences, the two sounder retrieval systems differ in the prior
214 estimates used for ECF and CTP. CLIMCAPS starts the cloud retrieval with background estimates
215 of 0.5 and 0.25 ECF at 350 hPa and 800 hPa CTP for the upper and lower cloud layers, respectively.
216 AIRS V7 uses 1/6 ECF at 350 hPa for the upper layer, and 1/3 ECF at 850 hPa (or 100 hPa above
217 surface in elevated terrain) for the lower cloud layer. However, since the final cloud retrievals of
218 both systems are shown to diverge significantly from their prior (Yue and Lambrigtsen 2020, Yue
219 et al. 2021), it is unlikely that different cloud prior estimates are a main contributor to the sounder
220 cloud retrieval product differences.

221 Although their spectral resolution is coarser than that of AIRS and CrIS, instruments like
222 MODIS and VIIRS provide high spatial-resolution cloud properties through information in
223 multiple narrowband channels covering the visible and IR spectral regions. However, significant
224 differences exist between the two imagers. MODIS measures the reflectance or radiance in 36
225 spectral bands, while VIIRS has an analogous subset of these bands (20 channels) plus a day/night
226 visible channel (Oudrari et al. 2015). The lack of near-IR and IR water vapor and CO₂ absorption

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227 channels in VIIRS has important implications on the available information content for clouds with
228 respect to MODIS. This impacts the determination of clouds, especially the detection of multilayer
229 clouds and clear sky in polar night conditions, and the determination of cloud thermodynamic
230 phase. It also impacts the retrieval of cloud-top properties, especially for high thin clouds.
231 Moreover, the difference of spectral location of the VIIRS 2.25 μm channel compared to the
232 analogous 2.13 μm MODIS channel has implications on the retrievals of cloud particle size, optical
233 depth, and thermodynamic phase (Platnick et al. 2020), which will be briefly summarized in the
234 following discussions. On the other hand, VIIRS provides a higher spatial resolution of 750 m at
235 nadir in cloud property retrievals, compared to the 1-km resolution in the Collection 6.1 MYD06
236 and cloud mask products. In addition, VIIRS has an onboard detector aggregation scheme that
237 limits the across-swath pixel growth. VIIRS edge of scan pixel size is roughly 1.625 km x 1.625
238 km versus roughly 2km x 4.9 km for MODIS (Platnick et al. 2021). The MYD06 products have
239 been shown to provide stable and well characterized cloud data records since 2002 (e.g. Yue et al.
240 2017). Given these instrument differences between MODIS and VIIRS, and a need to develop a
241 continuous data record extending beyond the MODIS era, the MODISVIIRS CLDMSK cloud
242 mask (Frey et al. 2020) and CLDPROP cloud-top and optical property (Platnick et al. 2021)
243 continuity algorithms were developed. By applying common algorithms to a subset of channels
244 available on both instruments, the continuity algorithms accommodate the detailed channel
245 differences between the two instruments while maximizing the information content on cloud
246 parameters.

247 The continuity CLDPROP products use only spectral channels common to both MODIS and
248 VIIRS. The algorithm has direct heritage with the Collection 6.1 MODIS atmosphere cloud
249 retrievals (MYD06), with cloud-top property datasets provided by the CLOUDS from AVHRR (the

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251 Advanced Very High Resolution Radiometer) - Extended (CLAVR-x) processing system
 252 (Heidinger et al. 2012, 2014) to account for more limited information for cloud-top property
 253 retrieval. CLAVR-x produces cloud phase reported as Cloud_Phase_Cloud_Top_Properties in the
 254 MODIS-VIIRS continuity cloud products. Since VIIRS does not have IR channels in the 13 μ m
 255 CO₂ absorption band, the MODIS CO₂ slicing solution for cloud top pressure retrievals for cold
 256 clouds is replaced with an IR_window channel optimal estimation approach coupled with a Cloud-
 257 Aerosol Lidar and Infrared Pathfinder Satellite Observations (CALIPSO)-derived *a priori*,
 258 (Heidinger et al. 2019). This in turn affects the optical property cloud phase algorithm (reported as
 259 Cloud_Phase_Optical_Properties in CLDPROP products), which removes the cold cloud sanity
 260 check applied in the MOD06/MYD06 that is based on the CO₂-slicing solution. The spectral
 261 mismatch of the MODIS 2.13 μ m and VIIRS 2.25 μ m channels also bring further changes to the
 262 Cloud_Phase_Optical_Properties retrieval by modifying the spectral cloud effective radius (Re)
 263 test approach. In the Version 1.1 MODIS-VIIRS continuity cloud product used in this study, the
 264 2.25 μ m test is omitted and the 1.61 μ m test is duplicated. Moreover, this channel spectral
 265 differences compel changes in the look-up tables (LUT) of spectral liquid cloud reflectance used
 266 in the retrieval, which include the use of an updated liquid water imaginary index of refraction
 267 dataset in the shortwave infrared region (Kuo et al. 1993) and an updated complex index of
 268 refraction dataset for 3.7 μ m (Wagner et al. 2005). Such differences in LUTs result in changes of
 269 cloud effective particle size (Re) (Platnick et al. 2020) that, along with cloud optical depth (COD),
 270 are used to derive cloud water path. Differences with the Collection 6.1 MODIS cloud retrieval
 271 algorithms, as well as inter-sensor differences between MODIS and VIIRS, have been reported in
 272 detail in recent studies such as Frey et al. (2020) and Platnick et al. (2021), which are based on
 273 granule comparisons and long-term mean statistics.

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286 2.2 Simultaneous Nadir Observations (SNOs) of collocated satellites

287 The pixel-scale comparisons will use SNOs between *Aqua*-AIRS and *SNPP*-CrIS. These SNOs
288 contain pixel pairs of observations from the two instruments when they observe the same location
289 at approximately the same scan angle and time. The AIRS-CrIS SNOs used herein were originally
290 developed by the JPL Sounder Science Investigator Processing System (SIPS) for inter-calibration
291 of two sounders (Manning and Aumann 2015). In order to ensure a close match between the
292 instruments, the following criteria are used to identify candidate SNOs:

- 293 • FOV centers between *Aqua*-AIRS and *SNPP*-CrIS are within 8 km;
- 294 • Observations are made within 10 minutes;
- 295 • Both instruments observe within 3.3° of nadir, which corresponds with ± 1 FOR
296 of AMSU for AIRS or ATMS for CrIS.

297

298 2.3 Pixel-scale collocations of imagers and sounders:

299 Utilizing the multi-sensor capability at the pixel scale requires accurate and computationally
300 efficient collocation of sounder and imager measurements. Various collocation methods exist
301 (Schreier et al. 2010, Nagle and Holz 2009, Yue et al. 2013). In this study, the method developed
302 by Wang et al. (2016) is applied by matching the instantaneous multi-sensor observations directly
303 based on line-of-sight (LOS) pointing vectors, defined as the vector from the satellite position to
304 the Earth surface pixel location. The details of this method and its accuracy are discussed at length
305 in Wang et al. (2016).

306 In this study, the same collocation method is applied to both *Aqua* and *SNPP* to match the finer
307 resolution imager pixels (MODIS and VIIRS) within a given sounder FOV (AIRS and CrIS). The
308 LOS vectors are calculated using the geolocation datasets for different sensors, which contain
309 latitude, longitude, satellite range, satellite azimuth and zenith angles. Collocation is performed
310 using the criterion that the angular difference between the LOS vectors for sounder and imager
311 should be less than half of the sounder FOV size angle. The CrIS FOV is treated as a 0.963° circle
312 which corresponds to ~41% of the peak response and collects ~98% of total radiation falling on
313 the detector (Wang et al. 2013). AIRS has a FOV half-power width of 1.1° (Fishbein et al. 2001).
314 However, 0.963° is used for both AIRS and CrIS in the collocation. After obtaining collocation
315 indices, the L2 cloud properties from both the imagers and sounders are populated accordingly.
316 The high spatial resolution information from MODIS and VIIRS is retained using higher statistical
317 moments and frequency distributions of cloud properties retrieved by imagers within collocated
318 sounder FOV. These statistical metrics include the mean, standard deviation, skewness and kurtosis
319 of MODIS and VIIRS cloud properties, the occurrence frequency of cloud types and cloud phase
320 reported by the cloud mask and cloud thermodynamic phase variables, and joint histograms on the
321 COD and CTP two-dimensional space following the convention of the International Satellite Cloud
322 Climatology Project (ISCCP, Rossow and Schiffer 1999). In addition to summarizing fine imager
323 spatial information over a coarser resolution sounder instrument, these statistical metrics physically
324 describe a variety of cloud processes at both regional and global scales for a range of cloud types
325 in different climate regimes, which are particularly relevant to sub-grid cloud parameterization in
326 numerical models (e.g. Zhu and Zuidema 2009, Kawai and Teixeira 2010 and 2012, Kahn et al.
327 2017). The ISCCP-type of joint histograms have been widely used to dissect the uncertainty of the

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cloud radiative forcing (e.g. Pincus et al. 2012) and climate feedback (e.g. Zelinka et al. 2012, Yue et al. 2016 and 2019) by cloud regimes (e.g. Oreopoulos et al. 2016).

By combining the SNOs and the sounder-imager collocated datasets, a multi-sensor multisatellite investigation is conducted to evaluate, at pixel scale, the self-consistency of cloud properties, to benchmark data continuity from the US polar-orbiting operational environmental satellites.

3. Results

Both *Aqua* and *SNPP* are in the 1:30 PM local equatorial crossing time sun-synchronous polar orbits, but at different altitudes. This altitude difference gives a ~ 2.667 day repeating pattern for AIRS and *SNPP*-CrIS observations at the same location. Accordingly, the number of SNOs between these two IR sensors varies with time and a large fraction are located at the high latitudes. In this study, seven focus days in January 2016 are selected for their large numbers of SNO pairs and the full operation for all four instruments. Table 2 lists the focus days and gives the number of observations obtained on each day. Figure 1 shows the latitudinal distribution of the focus day SNOs (black bars, y-axis on the left, Table 2). A significant number of observations ($>2,500$) are available at all latitudes, including the midlatitudes and tropics where SNOs are harder to obtain.

Fig. 2 shows the latitudinal variations of cloud frequency and zonal mean ECF and COD based on the data from the seven focus days. To determine the detection of clouds in the sounder FOV, two threshold values of ECF are used: 0.05 (solid lines) and 0.01 (dash lines). For MODIS and VIIRS, frequency of Cloudy, Uncertain cases as reported by the cloud mask variable is shown for MYD06 (black), MODIS continuity (red), and VIIRS continuity (blue) cloud products. Although it is difficult to directly compare the mean cloud properties retrieved by imagers and sounders,

AIRS V7 produces similar general patterns of latitudinal variation of cloud frequency with the imager products, which shows peaks of cloud occurrence in the tropics and midlatitude storm tracks, and troughs in the subtropics. However, CLIMCAPS V2 cloud retrievals do not show these variations, and its mean ECF values are much lower than AIRS V7 at all latitudes. A higher percentage of cloud frequency in the low latitude regions is reported by AIRS V7 than by imagers, consistent with previous findings showing higher sensitivity of hyperspectral IR sounders to optically thin clouds (Kahn et al. 2014, Yue et al. 2016). An increase of COD with latitude at mid to high latitude regions is detected by imagers, compared to a nearly flat or even decreasing mean ECF retrieved by the sounders. These differences will be further assessed in the following discussions.

3.1 Clouds retrieved by hyperspectral IR sounders

In Fig. 1, overlapped with the SNO count histograms are the occurrence frequency of sounder FOVs (colored lines, y-axis on the right) for four composites that satisfy the following four conditions, respectively: $ECF > 0.01$ (general cloudy condition), $ECF \leq 0.01$ (clear or very thin clouds), $ECF > 0.8$ (overcast or very thick clouds), and cases with successful CTP retrievals (QC for CTP is 0 or 1). These ECF values are selected based on the relationships between clouds and the IR sounder spectral information, as well as the retrieval uncertainty. The fraction of the highest quality atmospheric state retrievals below clouds, obtained from IR spectral information, decreases with higher ECF (Fetzer et al. 2006). The combination of IR and MW radiances can facilitate the retrieval of vertically resolved temperature and humidity profiles up to ECF of 0.7~0.8 (Yue et al. 2011, Yue and Lambrigtsen 2020, Yue et al. 2021). The ECF of 0.01 is often used as the threshold of cloud detection by IR sounders (e.g. Kahn et al. 2014). Moreover, it has

374 been shown that AIRS V7 cloud retrievals present higher uncertainty on thin, broken clouds and
375 cloud edges when $ECF < 0.01$ (Yue and Lambrigtsen 2020).

376 For each composite, the occurrence frequency is calculated as the percentage of AIRS or
377 CrIS FOVs with successful cloud retrievals that satisfy the composite condition relative to the
378 total number of FOVs in each latitudinal bin. The QC flags for each cloud parameter are reported
379 in the L2 products and used to determine whether the algorithm reports a successful cloud
380 retrieval (when $QC = 0$ or 1). Different colors are used to indicate retrieval algorithms for the two
381 sounders. Since AIRS V7 and CLIMCAPS retrieve cloud properties up to two cloud layers over
382 each IR sounder FOV, an effective CTP is calculated as the weighted mean CTP by the
383 ECF reported at each cloud layer.

384 These results show large differences between the AIRS V7 clouds with those from CLIMCAPS.
385 AIRS V7 produces a much larger number of cloudy observations (solid pink line in Fig. 1) and a
386 higher yield for CTP retrievals (dash dotted line, Fig. 1), except in the Antarctic region. The
387 magnitude of this difference reaches up to 30% over the Southern Hemisphere and the tropics.
388 Furthermore, AIRS V7 produces much more overcast or very thick clouds (dash lines, Fig. 1) but
389 fewer clear or very thin cloudy cases (dotted lines, Fig. 1) than CLIMCAPS, which is consistent
390 with smaller mean ECF and lower cloud frequency in the tropics and midlatitude storm track
391 regions by CLIMCAPS V2 in Fig. 2. As discussed previously, this is related to the differences
392 between the two algorithms for AIRS in cloud clearing and cloud retrieval QC, as well as the use
393 of different *a priori*. These differences are further evaluated in the following sections using the
394 imager observations.

395 Despite the differences of sensors, satellites, and spectral resolutions, the three CLIMCAPS

Version 2 retrievals evaluated in this study present similar latitudinal distributions of the cloud property distribution and cloud detection. As seen from Fig. 1, CLIMCAPS-*Aqua* (green dotted line) reports a higher percentage of clear or very thin cloudy cases than those for *SNPP* (yellow dotted line for CLIMCAPS-*SNPP* FSR and purple for CLIMCAPS-*SNPP* NSR), especially in the midlatitude region. Among the three CLIMCAPS products, CLIMCAPS-*Aqua* (green solid line) reports fewer cloudy cases than CLIMCAPS-*SNPP* (yellow and purple solid lines) in midlatitudes, but more cloudy cases in the tropics. The finer spectral resolution for CLIMCAPS-*SNPP* FSR retrievals produces a higher percentage of cloudy FOVs than the coarser spectral resolution radiances used by the NSR retrieval.

Figure 3 further characterizes the four IR sounder cloud retrievals using the joint distributions of observations among different algorithms. It is known that larger uncertainty of both sounder and imager retrievals exists over snow and ice covered surfaces (Chan and Comiso, 2013, Yue and Lambrigtsen 2020), so in this comparison the data points located in regions poleward of 60° are excluded. Cases are only included if both data products in the comparison (indicated by x- and yaxes of the plot) report valid retrievals. The three CLIMCAPS retrievals (x-axes) are compared with AIRS V7 (y-axes) for both ECF and CTP. The generally good agreement among the algorithms and sensors, especially for CTP, is encouraging, which shows the robustness of these products and consistency of information for clouds in hyperspectral IR sounders. However, CLIMCAPS reports a large number of cases with ECFs between 0 and 0.1, for which AIRS V7 reports ECFs ranging from 0 (clear sky) and 1 (completely cloudy). This issue is further illustrated in Fig. 4. For cases where CLIMCAPS-*Aqua* V2 retrieved ECF is less than 0.1, AIRS V7 (the magenta line) shows two peaks in the ECF occurrence frequency. The first peak is located at V7 ECF < 0.1, indicating the two algorithms agree with each other in cloud amount detection. The

larger second peak shows that more than 25% of cases with CLIMCAPS ECF < 0.1 have AIRS V7 ECF values of 0.8–0.9. As a result, the correlation coefficient (r) between ECF retrievals from AIRS V7 and CLIMCAPS V2 is only 0.27, which increases to 0.79 when neglecting ECF < 0.1 observations. Further separating the sounder FOVs into ice- and liquid-cloud-only categories shows that such inconsistency in cloud amount detection between the sounder algorithms exist in both categories as illustrated in Fig. S1 and S2. The sounder FOV is determined as ice/liquid-cloud-only when over 80% of collocated cloudy MODIS pixels are in ice/liquid thermodynamic phase in the MYD06 optical property cloud phase retrievals. Better agreements between sounder cloud products are found for ice-cloud-only FOVs.

A tighter agreement between CLIMCAPS V2 and AIRS V7 is seen for CTP retrievals as shown by points densely located along the identity line in Fig. 3. The correlation coefficients between CLIMCAPS-Aqua and AIRS V7 CTP are 0.69 for all cases and 0.92 for ECF > 0.1, respectively. High cloud cases (AIRS V7 CTP < 440hPa) show a much higher CTP correlation ($r = 0.87$) than for low clouds (AIRS V7 CTP > 600 hPa, $r = 0.43$). When both algorithms identify low clouds in the FOV, CLIMCAPS reports a slightly lower cloud top (larger CTP) than AIRS V7, with a median value difference of 12 hPa; whereas for high clouds, CLIMCAPS V2 reports a higher cloud top with its median CTP 13 hPa smaller than the one by AIRS V7.

In the next section, these differences among the various sounder cloud retrieval products are further evaluated using the cloud parameters determined by collocated MODIS and VIIRS data.

3.2 Comparison of sounder cloud properties and collocated imager measurements

Figures 5 and 6 compare the cloud properties retrieved from various sounder algorithms with the collocated imager cloud retrievals in the MYD06 and CLDPROP_MODIS products,

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442 respectively. Comparisons with CLDPROP_VIIRS are similar to those using CLDPROP_MODIS
443 and hence are not shown in these figures. The cloud properties from MODIS pixels are averaged
444 within the collocated sounder FOV before this comparison.

445 The IR sounder retrieved ECF is positively correlated with the imager observed COD in the
446 top rows of Figs. 5 and 6, showing the consistency of cloud amount determined using different
447 sensors. However, two main differences are noticed. First, it is clear that the CLIMCAPS V2 (for
448 both *Aqua* and *SNPP*) misidentifies a significant number of cloudy cases as clear or thin clouds.
449 As shown in Fig. 4, more than 50% of these cases are optically thick clouds with large cloud
450 amount ($ECF > 0.7$) reported by AIRS V7 and COD values ranging from 2 to 10 by MODIS and
451 VIIRS. Secondly, the comparisons between CLIMCAPS and imager cloud products do not have
452 the cluster corresponding to cases with both high ECF and large COD values, as in the comparison
453 between AIRS V7 and imagers. As discussed previously, this is related to misidentification of
454 cloudy cases as clear or thin cloud conditions by CLIMCAPS. However, another main cause is that
455 CLIMCAPS cloud retrievals have the same QC flags as the physical atmospheric state retrievals;
456 as a result, cases with large cloud amount are filtered out. In general, AIRS V7 products exhibit
457 better agreement with MODIS and VIIRS in detecting cloud amount and occurrence. CLIMCAPS
458 V2 cloud retrievals could be further improved with better cloud clearing flow and more careful
459 treatment when retrieving clouds with unsuccessful atmosphere physical retrievals.

460 The sounder and imager CTP retrievals are also compared in the bottom rows of Fig. 5 and 6.
461 Despite instrument and algorithm differences, when both sounder and imager detect high clouds
462 ($CTP < 440$ hPa, including $ECF < 0.1$ cases), CTP retrievals agree with each other well. The
463 correlation coefficients with MYD06 CTP are 0.77, 0.52, and 0.62 for AIRS V7, CLIMCAPS*Aqua*,
464 and CLIMCAPS-*SNPP*-FSR, respectively. When imagers detect low clouds ($CTP > 680$ hPa), IR

465 sounders determine the majority of cases as low clouds but with a tail toward CTP values
466 corresponding to high and mid-level clouds (middle row). The disagreement mainly occurs when
467 sounder retrieved ECF is less than 0.1 as shown by the magenta contour lines. These are cases
468 when larger uncertainty in infrared cloud retrieval exists, as discussed previously. After removing
469 these cases, the sounder-imager discrepancy in the low cloud conditions is reduced greatly (bottom
470 row), especially for AIRS V7. These differences are consistent with the known limitation of
471 imagers such as MODIS, which tend to miss high and thin cloud layers (Holz et al. 2008) when
472 compared with AIRS (Kahn et al. 2014). However, the analysis presented here cannot completely
473 rule out the impact of uncertainty in the IR sounder cloud retrievals. When both hyperspectral
474 sounders and narrowband imagers detect low clouds, sounders tend to retrieve smaller CTP than
475 imager. For AIRS V7, the median difference in this condition is -65, -77, and -80 hPa with MYD06,
476 CLDPROP_MODIS, and CLDPROP_VIIRS products, respectively. The results are further
477 analyzed for ice- and liquid-cloud-only sounder FOVs (Fig. S3-S6), which are determined using
478 the same criteria as in the previous section. It is clear that such disagreements between the sounder
479 and imager CTP are mainly originated from the liquid-cloud-only sounder FOVs (Fig. S5 and S6),
480 while good agreements between sounders and imagers are seen for ice-cloud-only conditions (Fig.
481 S3 and S4).

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483 3.3 Clouds retrieved by imagers

484 Figure 7 compares COD, CTP, and Re retrieved by different MODIS and VIIRS cloud
485 algorithms, with mean imager cloud properties over corresponding sounder FOVs are shown. Very
486 good agreement between MODIS and VIIRS, and between the MYD06 and continuity products is

488 seen. All correlation coefficients are greater than 0.8. For the three cloud parameters, correlation
 489 is always the highest between products derived from the same instrument (MYD06 and
 490 CLDPROP_MODIS), and the lowest between MYD06 and CLDPROP_VIIRS (but still reaching
 491 0.81, 0.88, and 0.81 for COD, CTP, and Re, respectively) when both instrument and algorithm are
 492 different. From the same instrument MODIS but different algorithms, the correlation is lowest for
 493 CTP retrievals ($r = 0.89$) compared to COD ($r = 0.97$) and Re ($r = 0.97$). This is because MYD06
 494 and the continuity cloud algorithm uses different methods and spectral channels to determine CTP,
 495 especially for cold clouds as shown in Fig. S7, where the correlation coefficients for CTPs from
 496 different imager cloud retrievals are less than 0.52 for ice-cloud-only conditions (Fig. S7) but larger
 497 than 0.79 for liquid-cloud-only cases (Fig. S8). However, a relationship near one-to-one is still
 498 seen clearly, indicating the consistency between the operational and continuity cloud products from
 499 MODIS, at least for the cloud properties averaged at the sounder resolution ($\sim 13.5\text{km}$).
 500 Correlations between MODIS and VIIRS cloud products are lower than those from MODIS alone
 501 (with different algorithms), even when both products are derived from the same continuity
 502 algorithm. The degradation of agreement is larger for COD and Re than for CTP (Fig. 6).
 503 Separating results into ice- and liquid-cloud-only conditions, the COD (Re) correlation coefficients
 504 between the MODIS and VIIRS continuity cloud products are 0.84 (0.70) and 0.82 (0.75) for ice-
 505 and liquid-cloud-only conditions, respectively, as shown in Fig. S7 and S8. Although such good
 506 agreements between the two imagers are encouraging, the correlation for Re from the two
 507 CLDPROP products is lower than that for COD, with a weaker correlation on the ice cloud Re
 508 retrievals. This reflects the effect of spectral channel and spatial resolution differences between
 509 MODIS and VIIRS, as well as the related adjustments made to the continuity algorithms, such as
 510 the liquid phase LUT for cloud microphysical retrievals, especially the impact of weaker ice crystal

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absorption at 2.25 μm (VIIRS) than at 2.13 μm (MODIS). Another possible factor is the collocation error existing in the SNOs, but this is ruled out since results with more conservative collocation criteria remain largely the same (not shown).

To further analyze the differences between the imager cloud products and the subpixel cloud heterogeneity over the sounder FOVs, the standard deviation and skewness of the imager cloud property distributions over the sounder FOVs are shown in Fig. 8 and 9, respectively. Correlations are weaker in these higher statistical moments, yet for standard deviation they remain larger than 0.6. Similar to comparisons for mean values, tight one-to-one relationships are seen for standard deviation at the sounder FOV scale between the two MODIS cloud products. Similar to mean value comparisons, the CTP standard deviation has the lowest correlation coefficient ($r = 0.63$) compared to the ones for COD ($r = 0.96$) and Re ($r = 0.87$), with a much lower correlation on CTP ($r = 0.44$) for ice-cloud-only conditions (Fig. S9) but a high correlation ($r = 0.71$) for liquid-cloud-only FOVs (Fig. S10). However, skewness only shows significant correlations for COD ($r = 0.78$) and Re ($r = 0.70$) between the two MODIS datasets, but poor correlations ($r < 0.3$) for CTP. The impact from the differences in CTP algorithms thus shows up more strongly on the higher statistical moments and on cold cloud scenes. When evaluating data from different sensors, no correlation is seen for skewness of any of the cloud parameters even with the same retrieval algorithms (Fig. 9, middle and right columns), different from the comparisons using mean value and standard deviation (Figs. 7 and 8, middle and right columns).

3.4 Joint histograms, cloud types, and cloud thermodynamic phase

3.4.1 Cloud type by cloud property joint histograms

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Figs. 10-13 show the two-dimensional cloud histograms calculated using SNOs from the focus days over different surface types and regions, including the tropics (30°N~30°S), over ocean (land fraction < 0.1, 60°N~60°S), over land (land fraction > 0.9, 60°N~60°S), and over ice and snow covered surfaces (frozen surfaces), respectively. The land fraction and surface classes are obtained from the AIRS V7 L2 product under variable names of landFrac and SurfClass, respectively. For MODIS and VIIRS, the ISCCP type of CTP-COD joint histograms are generated by summing the joint distributions over individual AIRS and CrIS FOV, with no averaging over sounder FOV. For AIRS and CrIS, joint distributions are calculated on the CTP and ECF space.

Consistent with results in previous sections, AIRS V7 shows peaks of both thin and thick clouds while CLIMCAPS V2 products show a single peak distribution of thin clouds. Better consistency of AIRS V7 with imager cloud products is also shown by the joint histograms. For example, in the tropics (Fig. 10) clusters corresponding to optically thick high clouds, thin cirrus, and broken or optically thin low clouds are seen in the AIRS V7 CTP-ECF histogram, consistent with the patterns in the MODIS and VIIRS CTP-COD histograms. Agreement between AIRS V7 and imager clouds is also found for mid-level and low cloud clusters over ocean (Fig. 11) and for high and mid-level clouds over land (Fig. 12). Over frozen surfaces (Fig. 13), the sounder clouds show optically thin and high clouds, especially in CLIMCAPS V2; a large percentage of mid-level clouds with medium to large ECF values are seen in AIRS V7, more consistent with the cloud histograms from imager observations. However, MODIS and VIIRS cloud detection and retrievals suffer a higher uncertainty over frozen surfaces (Chan and Comiso, 2013), and the small atmospheric thermal contrast with frozen surfaces presents additional challenges for hyperspectral

IR sounder retrievals (Yue and Lambrigtsen 2020). Therefore, more accurate cloud measurements from in-situ or active space-borne instruments are needed to further quantify the quality of these imager and sounder cloud retrieval products in snow- and ice-covered regions.

Because of its long temporal coverage since 1999 when *Terra* MODIS began operating, high quality, and the distinct physical characteristics of different cloud types, the MODIS cloud data record, especially the CTP-COD joint histograms, have been widely used in different aspects of climate studies. These include detailed analyses on the radiative effect of different cloud types (Yue et al. 2016, Oreopoulos et al. 2016), evaluation of climate model simulations of clouds (Pincus et al. 2012), quantification of the cloud feedback by different cloud types (Zhou et al. 2014, Yue et al. 2019), and investigations of cloud impacts on hydrological cycle and the global circulation (Su et al. 2017), especially in the tropics. Therefore, the differences of the cloud frequency histograms from various imager retrieval products in the tropics are further analyzed here. In Fig. 14, the MODIS continuity product (depicted in Fig. 10) is used as the common base to evaluate the differences caused by algorithms and sensors: 1) between current NASA standard MODIS retrievals and the MODIS continuity algorithms, and 2) between the MODIS and VIIRS continuity cloud data records. The magnitude of joint frequency histogram differences is within $\pm 5\%$ using the focus day observations. MYD06 shows more clouds with CTP < 180 hPa but fewer low clouds with CTP > 800 hPa than the continuity product, consistent with findings in Platnick et al. (2021). VIIRS continuity cloud retrievals produce higher frequencies of clouds with COD between 9.4 and 60, but fewer high clouds with COD < 9.4. Whether and how these differences will impact the long-term trend and short-term variability of clouds as seen by the imagers warrants further study.

3.4.2 Cloud thermodynamic phase

Both MYD06 and continuity cloud products provide cloud thermodynamic phases (Table 1), given by the optical property retrieval (Cloud_Phase_Optical_Properties, in both MYD06 and continuity products) and the CLAVR-x processing system (Cloud_Phase_Cloud_Top_Properties, continuity products only). The Cloud_Phase_Cloud_Top_Properties variable reports flags determining pixels to be cloud free, water cloud, ice cloud, mixed phase cloud, or undetermined phase. The Cloud_Phase_Optical_Properties flags indicate cloud mask not determined for pixel, clear sky, liquid water cloud, ice cloud, or undetermined phase, the last of which includes mixed phase clouds (Marchant et al. 2016). AIRS thermodynamic cloud phase, which is available in the AIRS V6 and V7 Level 2 Support product, is based on a set of brightness temperature difference and threshold tests using the channels in 960, 1231, 930, and 1227 cm^{-1} (Nasiri and Kahn 2008, Kahn et al. 2014). These tests are applied to AIRS FOVs where ECF > 0.01, and classify the AIRS FOV as containing liquid, ice, or unknown cloud phases. Detailed comparisons of AIRS cloud phases with CALIPSO indicate good agreement with CALIPSO on ice phase detection, and conservative liquid phase determination (Jin and Nasiri 2014, Peterson et al. 2020). These studies also show that the unknown class of AIRS cloud phase corresponds to scenes containing both ice and liquid particles, and low-level liquid clouds, especially in the trade-wind cumulus cloud regime.

Figs. 15-18 show the histograms of cloud thermodynamic phase (solid color bars for imagers and magenta symbols for AIRS) for the same set of focus-day SNOs. Similar to joint histograms in Fig. 10-13, each figure shows results over the four types of surfaces and regions: tropics, ocean, land, and frozen surfaces. MODIS and VIIRS cloud mask histograms (hollow color bars) are also shown in the figures, together with the frequency of clear sky detected by IR sounders (ECF < 0.01, colored solid circles). Note that for MODIS and VIIRS, the mixed-phase or undetermined phase category is shown with the y-axis on the right due to their much smaller frequency of occurrence.

603 For clear sky detection, the cloud-mask clear frequencies from all the imager products are similar
604 except over the frozen surfaces, where VIIRS cloud mask shows 10% higher frequency than
605 MODIS. For IR sounders, AIRS V7 produces significantly lower clear-sky frequency than
606 CLIMCAPS and imager cloud products over non-frozen surfaces. Over frozen surfaces, more
607 frequent clear conditions are reported by AIRS V7 than CLIMCAPS, although AIRS V7 is more
608 consistent with the clear frequency from MODIS and VIIRS data.

609 The frequencies of liquid or ice phase clouds are highly consistent between two cloud phase
610 variables in various imager cloud products, except for ice phase determination over frozen surfaces.
611 This is supported by the low uncertainty range of ice and liquid phase for these four conditions as
612 shown in Table 3. Here uncertainty is roughly characterized by the standard deviation of estimates
613 from different products and variables. The Cloud_Phase_Cloud_Top_Properties reports higher
614 percentage of liquid phase than Cloud_Phase_Optical_Properties. In particular, the VIIRS cloud
615 top cloud phase product always reports the highest frequency of liquid clouds. From both cloud
616 phase variables, MODIS reports more ice and fewer liquid clouds than VIIRS. When looking at
617 Cloud_Phase_Optical_Properties for MODIS, ice (liquid) cloud frequency is higher (lower) in
618 MYD06 than in the CLDPROP_MODIS products. The undetermined phase by the
619 Cloud_Phase_Optical_Properties includes both mixed and uncertain phases (Baum et al. 2012).
620 Except in tropics, MYD06 has the higher frequency of undetermined cases than the continuity
621 cloud products, and this is most prominent over the frozen surfaces with MYD06 reporting ~2.8%.

622 AIRS cloud phase retrievals report a higher frequency of ice clouds than imagers under all
623 conditions, especially in the tropics (Fig. 15) and over land (Fig. 17). However, a much lower
624 frequency of liquid clouds is retrieved by AIRS, which is consistent with a more conservative
625 liquid phase determination approach applied by AIRS cloud phase algorithm (Kahn et al. 2014).

The unknown phase of AIRS ranges from ~15% over the frozen surfaces to ~45% over ocean and in the tropics, which corresponds with broken and thin low clouds and scenes with both ice and liquid cloud particles (Jin and Nasiri 2014).

4. Summary

In this study, the pixel-scale collocation between the hyperspectral infrared (IR) sounders (AIRS and CrIS) and high spatial resolution imagers (MODIS and VIIRS) is performed on the pairs of Simultaneous Nadir Observations (SNOs) between *Aqua*-AIRS and *SNPP*-CrIS. Using this approach, the cloud parameters retrieved by various algorithms for IR sounders and imagers from different platforms are evaluated at the pixel level. Quantifying uncertainty in the cloud observational data records is important for constraining the high uncertainty of clouds in weather and climate research. This is also crucial in improving the retrieval of atmospheric, surface, and radiation properties since satellite observations are highly subject to uncertainties and limitations associated with cloud conditions in the instrument field of view (FOV) (e.g. Yue et al. 2013, Wong et al. 2015, Tian et al, 2020). Moreover, narrowband imagers and hyperspectral sounders provide important components of the long-term sustained observations of cloud properties in the Program of Record (POR), as noted by the 2017 US National Academy Decadal Survey (ESAS 2017). The analyses presented here will help to assess the capability of the POR, thus to identify potential gaps existed in the POR for cloud properties.

Both the NASA standard and continuity retrieval algorithms for sounders and imagers are investigated here in order to quantify the differences among the retrieval products, and to examine the consistency and continuity of the data products from multiple sensors across different satellites. This is essential to the goal of building a continuous record of satellite data using the *Terra*, *Aqua*,

649 *SNPP*, and *JPSS* series satellites, with sufficient quality to detect and quantify global
650 environmental change.

651 Multiple cloud parameters are analyzed (Table 1). Comparisons are made by investigating the
652 mean cloud parameters, and higher statistical moments of cloud property distributions measured
653 by MODIS and VIIRS over the corresponding AIRS and CrIS FOV. Cloud types indicated by the
654 joint histograms of cloud properties and cloud thermodynamic phases are included. Through these
655 comparisons, good agreement is found between the sounder and imager retrieved cloud products,
656 yet with distinct differences likely arising from algorithm and sensor differences. For IR sounders,
657 cloud top pressure (CTP) retrieved by AIRS Version 7 (V7) and CLIMCAPS (*-Aqua* and *-SNPP*)
658 Version 2 (V2) agree, as shown by correlation coefficients of 0.69 for all cases and 0.92 for cases
659 with effective cloud fraction (ECF) greater than 0.1, respectively. Compared to AIRS V7,
660 CLIMCAPS tends to produce a lower cloud top (CTP 12 hPa larger) for low clouds, but higher
661 cloud top (CTP 13 hPa smaller) for high clouds. However, CLIMCAPS V2 significantly
662 overestimates the frequency of clear and optically thin cloud ($ECF < 0.1$), relative to AIRS V7 and
663 imager products from both MODIS and VIIRS. This is due to the algorithmic differences between
664 CLIMCAPS V2 and AIRS V7 cloud retrieval algorithms. These differences include whether
665 iteration of cloud clearing is performed, the surface/atmospheric states used in the cloud retrieval,
666 the quality control procedures used, and different *a-priori* states used by AIRS V7 and CLIMCAPS.
667 How these differences affect the downstream atmospheric and surface retrievals in the two
668 algorithms, and the attribution of impacts from each factor, is beyond the scope of this study and
669 warrants further investigation.

670 High consistency is seen among different imager cloud products, especially in the mean and
671 standard deviation of cloud properties from the MODIS atmosphere cloud property retrieval

672 (MYD06) and the MODIS-VIIRS continuity cloud products (CLDPROP). The magnitude of the
673 correlation coefficients closely reflects the impact of algorithm differences and instrument spectral
674 and resolution differences, with highest correlations obtained between two MODIS products (same
675 sensor but different algorithms) and lowest between MYD06 and CLDPROP_VIIRS (different
676 sensors, different algorithms). The correlation coefficients are always higher for cloud optical
677 depth (COD) and particle effective radius (Re) than for CTP. For mean cloud properties, they are
678 as large as 0.97 between MYD06 and CLDPROP_MODIS, and 0.89 for CTP. For standard
679 deviations within the sounder FOV, the correlations are smaller than those for mean cloud
680 properties, ranging from 0.77 to 0.96 for COD, 0.66 to 0.97 for Re, but only 0.60 to 0.63 for CTP.
681 This is likely due to the fact that completely different CTP retrieval methods are used in the MODIS
682 operational and continuity cloud algorithms to accommodate the lack of near-IR and IR water
683 vapor and CO₂ absorption channels in VIIRS. Such algorithm and instrument impacts are more
684 apparent in the higher moment statistics of cloud properties such as skewness. The correlations of
685 COD and Re skewness between MYD06 and CLDPROP_MODIS drop to 0.78 and 0.70,
686 respectively. They are further reduced to below 0.4 when comparing MODIS and VIIRS cloud
687 products. For CTP skewness, the correlation coefficients are less than 0.3.

688 Two different cloud thermodynamic phase retrievals are available from imager observations,
689 which are obtained by the optical property retrieval (Cloud_Phase_Optical_Properties, in both
690 MYD06 and MODIS-VIIRS continuity products) and the CLAVR-x processing system
691 (Cloud_Phase_Cloud_Top_Properties, continuity products only). The frequencies of liquid or ice
692 phase clouds are very consistent between two cloud phase variables in different imager cloud
693 products, with uncertainty usually generally less than 4%. The largest uncertainty is reported for
694 ice phase determination over snow and ice covered surfaces. MODIS retrievals report more ice and

fewer liquid clouds than VIIRS, consistent with findings by Platnick et al. (2020). Comparing the two different cloud phase retrievals, the Cloud_Phase_Cloud_Top_Properties reports higher percentages of liquid phase than Cloud_Phase_Optical_Properties, and the Cloud_Phase_Optical_Properties in MYD06 detects higher (lower) frequencies of ice (liquid) clouds than that in the CLDPROP_MODIS products.

The general consistency of cloud observations among different sensors aboard *Aqua* and *SNPP* from various algorithms is encouraging, especially for achieving a continuous multi-decadal climate data record of clouds that can extend beyond the A-Train era and well into the 2030s with the *JPSS* series. The quantification of algorithm differences has important implications for future retrieval algorithm developments, and will further improve the capability and accuracy of such climate data records.

Data and Code Availability:

MODIS (MYD06 10.5067/MODIS/MYD06_L2.061; MYD35 10.5067/MODIS/MYD35_L2.061; CLDPROP-MODIS 10.5067/VIIRS/CLDPROP_L2_MODIS_Aqua.011; CLDMSK-MODIS 10.5067/MODIS/CLDMSK_L2_MODIS_Aqua.001) and VIIRS data (CLDPROP-VIIRS 10.5067/VIIRS/CLDPROP_L2_VIIRS_SNPP.011; CLDMSK-VIIRS 10.5067/VIIRS/CLDMSK_L2_VIIRS_SNPP.001) were obtained through the Level-1 Atmosphere Archive and Distribution System (LAADS; <http://ladsweb.nascom.nasa.gov/>). AIRS (AIRS V7 Level 2 Support Product [10.5067/APJ6EEN0PD0Z](https://doi.org/10.5067/APJ6EEN0PD0Z); CLIMCAPS-Aqua Version 2 Level 2 [10.5067/JZMYK5SMYM86](https://doi.org/10.5067/JZMYK5SMYM86)) and CrIS data (CLIMCAPS-SNPP Version 2 FSR

717 [10.5067/62SPJFQW5Q9B](https://doi.org/10.5067/62SPJFQW5Q9B); CLIMCAPS-SNPP Version 2 NSR [10.5067/8RUZ11F8U1UX](https://doi.org/10.5067/8RUZ11F8U1UX)) were
718 obtained from the NASA Goddard Earth Sciences Data Information and Services Center
719 (GESDISC) and could be accessed at <https://earthdata.nasa.gov/>. The collocation code is publicly
720 available from https://github.com/wanglikun1973/CrIS_VIIRS_collocation. The data used to
721 generate the figures and tables in this study can be obtained by contacting the corresponding author.

722

723 **Author Contribution:**

724 QY conceptualized the study, developed the methodology and datasets, carried out the formal
725 analyses, and contributed to the writing of the manuscript. EF, BK, NS, JB, and BL contributed
726 to the data curation, validation, investigation, and the writing of the manuscript. LW, IT, MM,
727 and KM contributed to the data curation and software.

728

729 **Competing Interests:**

730 The authors declare that they have no conflict of interest

731

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993 *Res. Lett.*, 36, L05807, doi:[10.1029/2008GL036817](https://doi.org/10.1029/2008GL036817), 2009.

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916 Table 1: The satellite cloud parameters examined in this study, and the retrieval algorithms 917 and products from which these parameters are obtained.

Satellite	Sensor	Retrieval Algorithm / Product (Nadir Spatial Resolution in km)	Cloud Parameters	Formatted Table
Aqua	AIRS	AIRS Version 7 Level 2 Standard and Support Product (13 km)	<ul style="list-style-type: none"> Effective Cloud Fraction (ECF) Cloud Top Pressure (CTP) Cloud Thermodynamic Phase 	Formatted: Right: 0.06", Space After: 0 pt, Bulleted + Level: 1 + Aligned at: 0.05" + Indent at: 0.25" Deleted: ¶
		Version 2 CLIMCAPS-Aqua Level 2 Infrared and Microwave Combined Retrieval (13 km)	<ul style="list-style-type: none"> Effective Cloud Fraction (ECF) Cloud Top Pressure (CTP) 	Formatted: Bulleted + Level: 1 + Aligned at: 0.05" + Indent at: 0.25" Formatted: Right: 0.06", Space After: 0 pt, Bulleted + Level: 1 + Aligned at: 0.05" + Indent at: 0.3" Deleted: ¶
	MODIS	Collection 6.1 Aqua MODIS Atmosphere Level 2 Cloud Product (MYD35, MYD06) (1 km)	<ul style="list-style-type: none"> Cloud Mask Cloud Top Pressure (CTP) Cloud Optical Depth (COD) Cloud Effective Radius (Re) Cloud Phase Optical Properties 	Formatted: Bulleted + Level: 1 + Aligned at: 0.05" + Indent at: 0.3" Formatted: Left Formatted: Bulleted + Level: 1 + Aligned at: 0.05" + Indent at: 0.3"
		Version 1.1 NASA MODIS Continuity Cloud Mask and Cloud Property Products (CLDMSK/CLDPROP_MODIS), (1 km)	<ul style="list-style-type: none"> Cloud Mask Cloud Top Pressure (CTP) Cloud Optical Depth (COD) Cloud Effective Radius (Re) Cloud Phase Optical Properties Cloud Phase Cloud Top Properties 	Formatted: Bulleted + Level: 1 + Aligned at: 0.05" + Indent at: 0.3" Deleted: Formatted: Left
SNPP	CrIS	Version 2 CLIMCAPS-SNPP Full Spectral Resolution (FSR) Level 2 Retrieval (13 km)	<ul style="list-style-type: none"> Effective Cloud Fraction (ECF) Cloud Top Pressure (CTP) 	Formatted: Right: 0.06", Space After: 0 pt, Bulleted + Level: 1 + Aligned at: 0.05" + Indent at: 0.3" Deleted: ¶
		Version 2 CLIMCAPS-SNPP Nominal Spectral Resolution (NSR) Level 2 Retrieval (13 km)	<ul style="list-style-type: none"> Effective Cloud Fraction (ECF) Cloud Top Pressure (CTP) 	Formatted: Left, Bulleted + Level: 1 + Aligned at: 0.05" + Indent at: 0.3" Formatted: Right: 0.06", Space After: 0 pt, Bulleted + Level: 1 + Aligned at: 0.05" + Indent at: 0.3" Deleted: ¶
	VIIRS	Version 1.1 NASA VIIRS Continuity Cloud Mask and Cloud Property Products (CLDMSK/CLDPROP_VIIRS) (0.75 km)	<ul style="list-style-type: none"> Cloud Mask Cloud Top Pressure (CTP) Cloud Optical Depth (COD) Cloud Effective Radius (Re) Cloud Phase Optical Properties Cloud Phase Cloud Top Properties 	Formatted: Left, Bulleted + Level: 1 + Aligned at: 0.05" + Indent at: 0.3" Formatted: Bulleted + Level: 1 + Aligned at: 0.05" + Indent at: 0.3"
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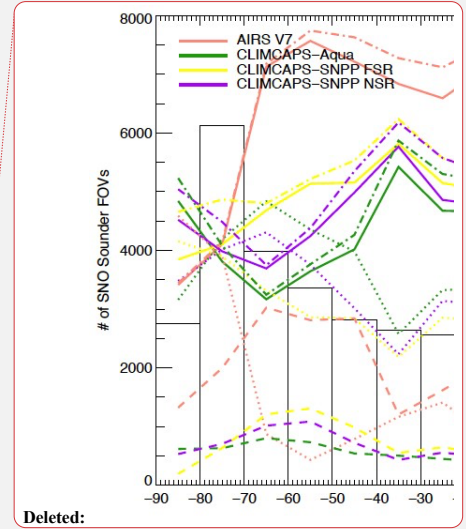
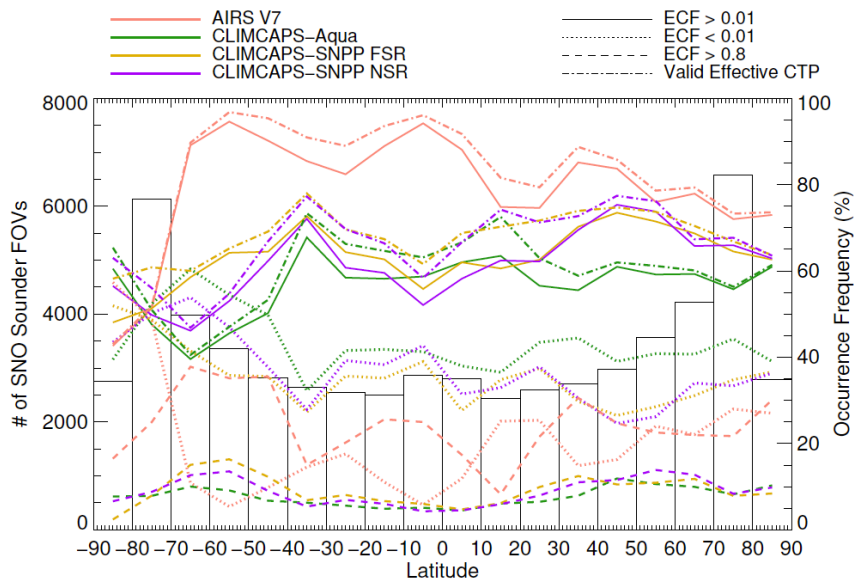
Table 2 Number of SNOs between *Aqua-AIRS* and *SNPP-CrIS* on the seven focus days used in this study.

Focus Day	Jan. 01, 2016	Jan. 03, 2016	Jan 04, 2016	Jan 09, 2016	Jan 11, 2016	Jan 14, 2016	Jan 17, 2016
# of SNOs	10,000	10,000	1372	10,000	10,000	10,000	8,903

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Table 3. The mean value and uncertainty range of the occurrence frequencies of ice and liquid phase clouds based on the cloud thermodynamic phase variables from the three imager cloud retrievals: MYD06, CLDRPOP_MODIS, and CLDPROP_VIIRS. Results over the five types of surfaces and regions are shown respectively for tropics, ocean, land, frozen surfaces, and global. For each condition, five estimates of cloud phase frequencies are available based on two types of imager-derived cloud thermodynamic phase: Cloud_Phase_Optical_Properties determined by the optical property retrieval (provided in both MYD06 and the two continuity products), and Cloud_Phase_Cloud_Top_Properties obtained through the CLAVR-x processing system applied in the continuity cloud algorithm (provided in the CLDPROP-MODIS and -VIIRS cloud products). The uncertainty range is characterized by the standard deviation of the five estimates obtained in each region.

Frequency (%)	Tropics (30°N~30°S)	60°N~60°S Ocean	60°N~60°S Land	Frozen Surfaces	Global, All Cases
Liquid Phase	37.64±3.21	53.94±3.50	35.16±2.81	14.03±1.10	44.27±2.79
Ice Phase	26.36±1.96	21.32±2.59	23.37±1.03	14.28±4.38	20.43±3.02



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Figure 1. The latitudinal distribution of the SNO pairs for *Aqua*-AIRS and *SNPP*-CrIS (black bars) and the occurrence frequencies of various sounder retrieved cloud parameters (right yaxis, %) for four composites that satisfy the following four conditions, respectively: $ECF > 0.01$ (solid lines, general cloudy condition), $ECF \leq 0.01$ (dotted lines, clear or very thin clouds), $ECF > 0.8$ (dash lines, overcast or very thick clouds), and cases with successful CTP retrievals (dash dotted lines, QC for CTP is 0 or 1). Data from the seven focus days are used (see Table 2) and binned by latitude of the sounder FOVs in 10° latitude bins. Four different sounder retrieval products are shown by colored lines: AIRS Version 7 (AIRS V7, pink), CLIMCAPS-*Aqua* (green), CLIMCAPS-*SNPP* FSR (yellow), and CLIMCAPS-*SNPP* NSR (purple). Occurrence frequency is calculated as the percentage of AIRS or CrIS FOVs with successful cloud retrievals (quality control indicator = 0 or 1) satisfying the aforementioned four conditions to the total number of FOVs in each latitudinal bin.

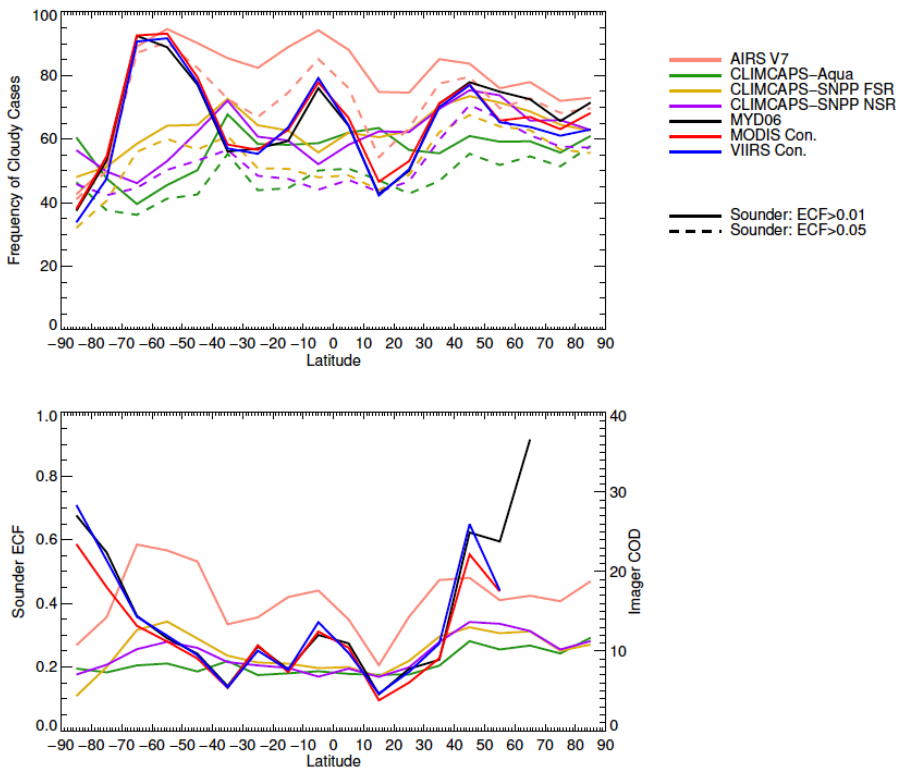
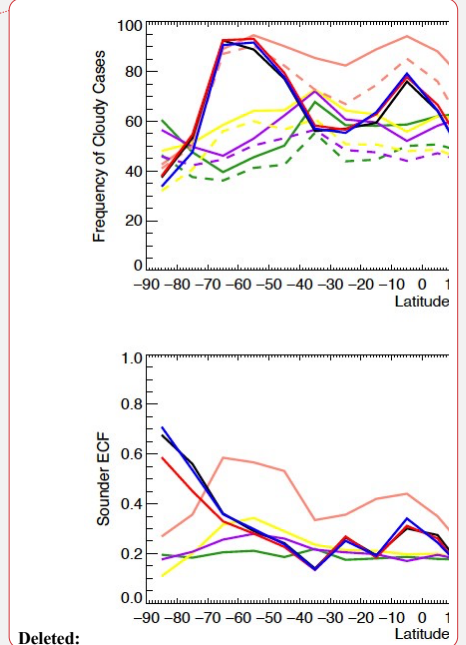


Figure 2. a) Zonal mean frequency of cloudy cases as observed by hyperspectral sounders and imagers. For MODIS and VIIRS, frequency of Cloudy, Uncertain cases as reported by cloud mask is shown for MYD06 (black), MODIS continuity (red), and VIIRS continuity (blue) cloud



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products. For AIRS and CrIS, solid and dash lines show frequencies of sounder FOVs with $ECF > 0.01$ and $ECF > 0.05$, respectively. Results for AIRS Version 7 (AIRS V7, pink), CLIMCAPS-*Aqua* (green), CLIMCAPS-*SNPP* FSR (yellow), and CLIMCAPS-*SNPP* NSR (purple) are shown for sounder cloud products. b) Zonal mean values of sounder ECFs (left y axis) and imager COD (right y axis) from these retrieval algorithms.

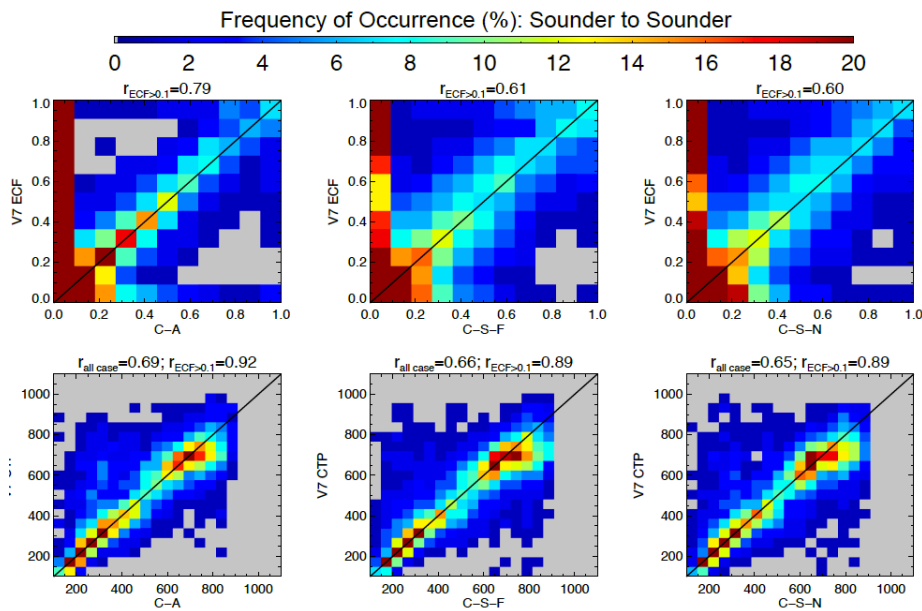


Figure 3. Comparisons of ECF (top row) and effective CTP (bottom row) derived from different sounder retrieval algorithms. Linear correlation coefficients are calculated for cloud properties obtained from retrieval products indicated on the axes and are given on top of the each plot. From left to right, results comparing AIRS Version 7 with CLIMCAPS-*Aqua* (C-A),

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Frequency of Occurrence (%): Sounder to Sounder

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CLIMCAPS-*SNPP* FSR (C-S-F), and CLIMCAPS-*SNPP* NSR (C-S-N) are shown using joint distributions of frequency of occurrence (%). The data points located in regions poleward of 60° are excluded. Cases are included only when both retrievals in comparison (x- and y-axes of the plot) report valid retrievals.

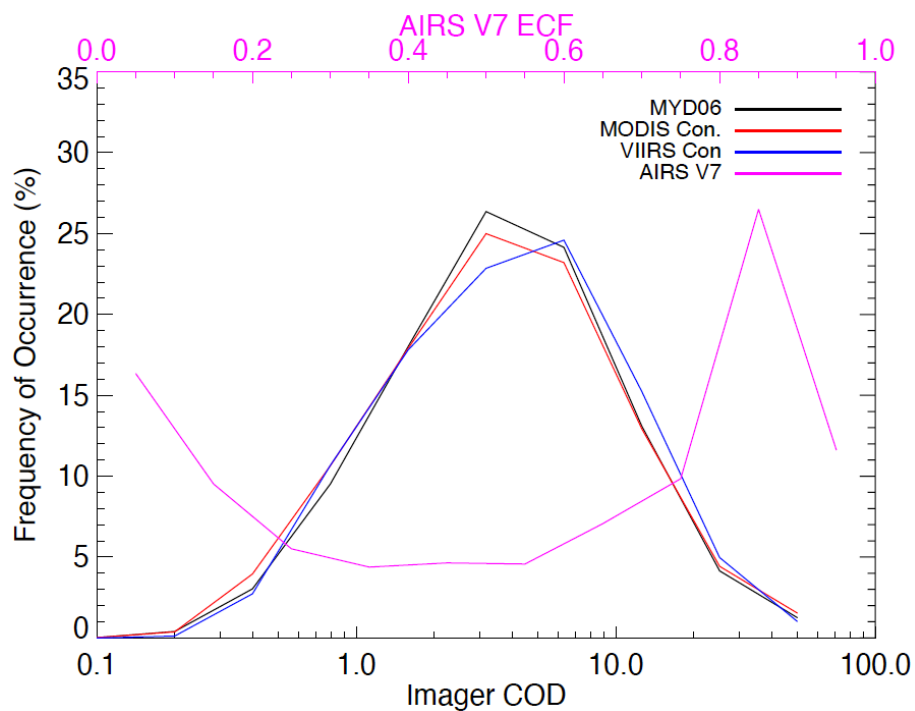
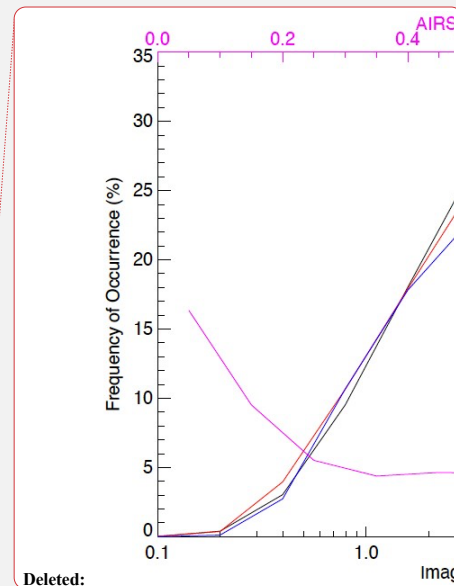
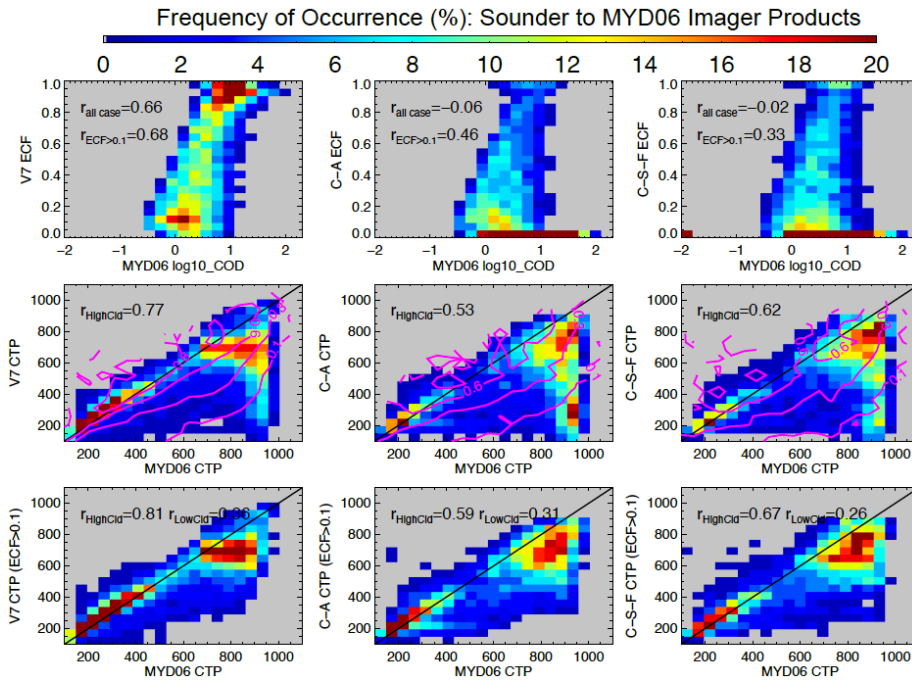


Figure 4. Frequency histograms showing the density distributions of imager cloud optical depth (COD, bottom x-axis) and AIRS V7 ECF (magenta, upper x-axis) for cases where V2 CLIMCAPS-*Aqua* retrieves an ECF value less than 0.1. Different imager cloud products are



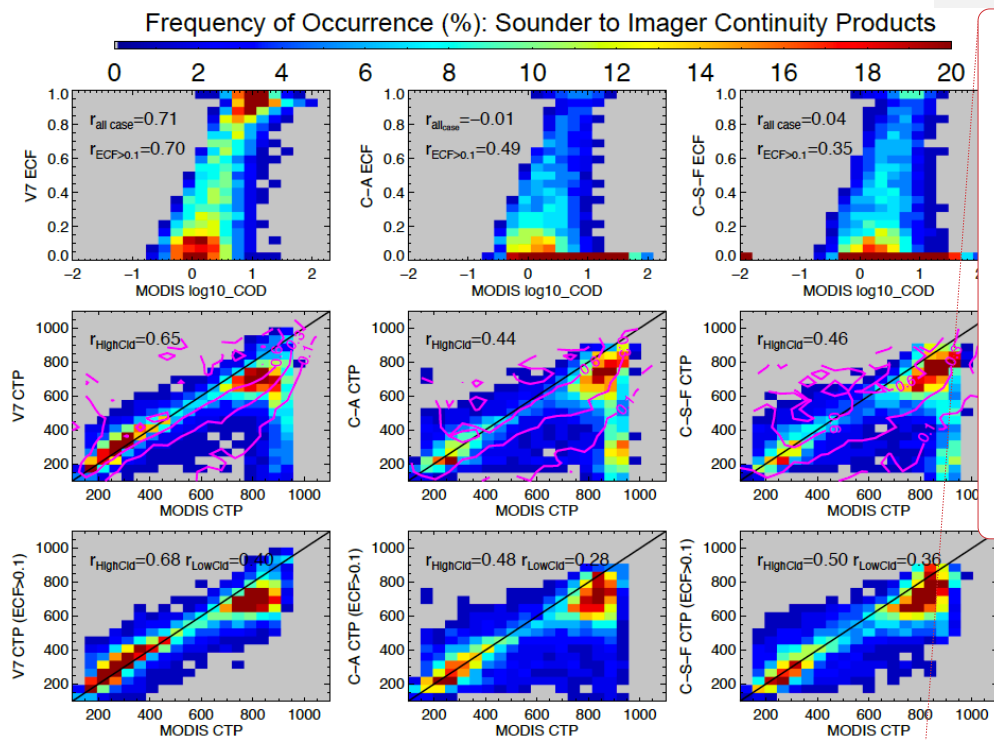
990 included: MYD06 (black), *Aqua*-MODIS continuity cloud products (MODIS Con., red), and
991 *SNPP*-VIIRS continuity cloud products (VIIRS Con., blue).
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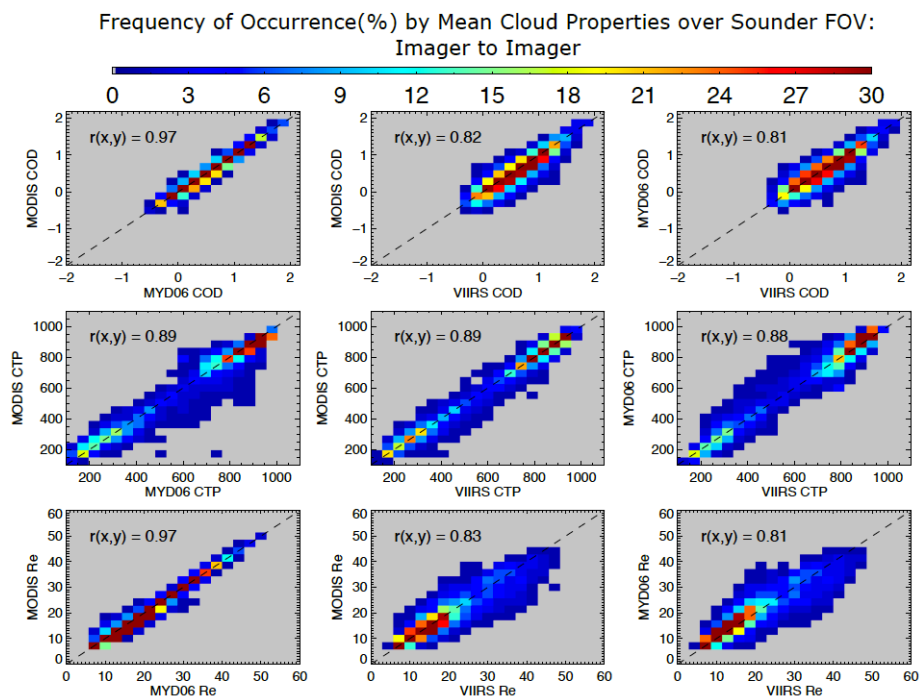
1015 pixels collocated within the same sounder FOV are averaged before comparing with the IR
1016 sounder data. Linear correlation coefficients between the variables on x- and y-axes for different
1017 conditions are given in each plot.
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Figure 6. Similar to Fig. 5, except using the MODIS continuity cloud product (CLDPROP_MODIS).

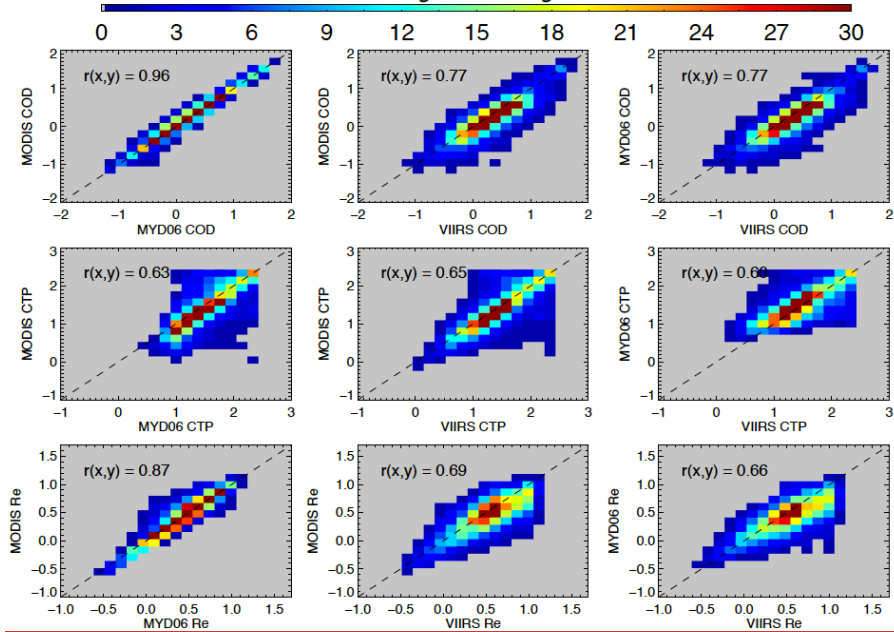


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Figure 7. Comparison of cloud optical depth (COD, in log10 scale), cloud top pressure (CTP, hPa), and effective particle size (Re, μm) retrieved by MODIS and VIIRS cloud algorithms. The mean imager cloud properties over corresponding sounder FOVs are compared over the SNOs. From left to right show the results of following comparisons: *Aqua* MODIS continuity cloud

1013 products (CLDPROP_MODIS) with MYD06, CLDPROP_MODIS with *SNPP*-VIIRS continuity
1014 cloud products (CLDPROP_VIIRS), and MYD06 with CLDPROP_VIIRS, respectively. Linear
1015 correlation coefficients between the variables on x- and y-axes are given in each plot.
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Frequency of Occurrence (%) by Cloud Property Standard Deviation over Sounder FOV:
Imager to Imager

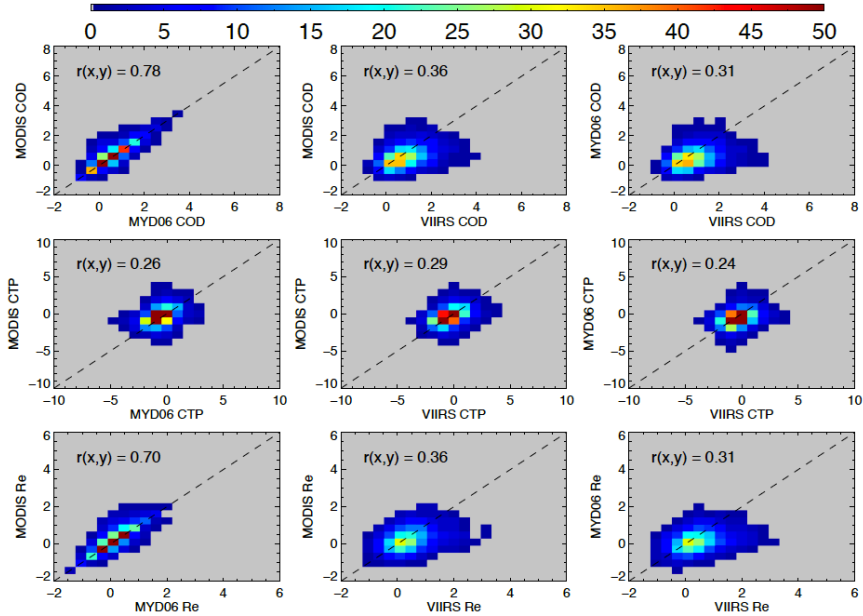


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Figure 8. Similar to Fig. 7, except showing comparisons of standard deviation of cloud properties over the sounder FOV, which are calculated using the finer resolution imager observations collocated with the same sounder FOV. All the results are presented on log10 scale. Linear correlation coefficients between the variables on x- and y- axes are given in each plot.

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Frequency of Occurrence (%) by Cloud Property Distribution Skewness over Sounder FOV:
Imager to Imager



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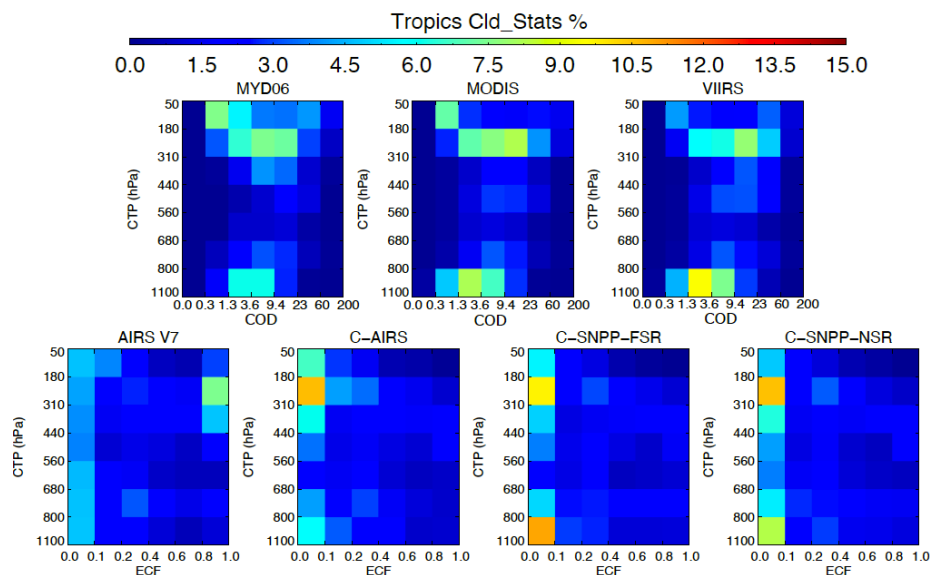
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Figure 9. Similar to Figs. 8 and 7, except cloud property skewness over sounder FOV is used in the comparison. Results are shown in linear scale. Linear correlation coefficients between the variables on x- and y-axes are given in each plot.

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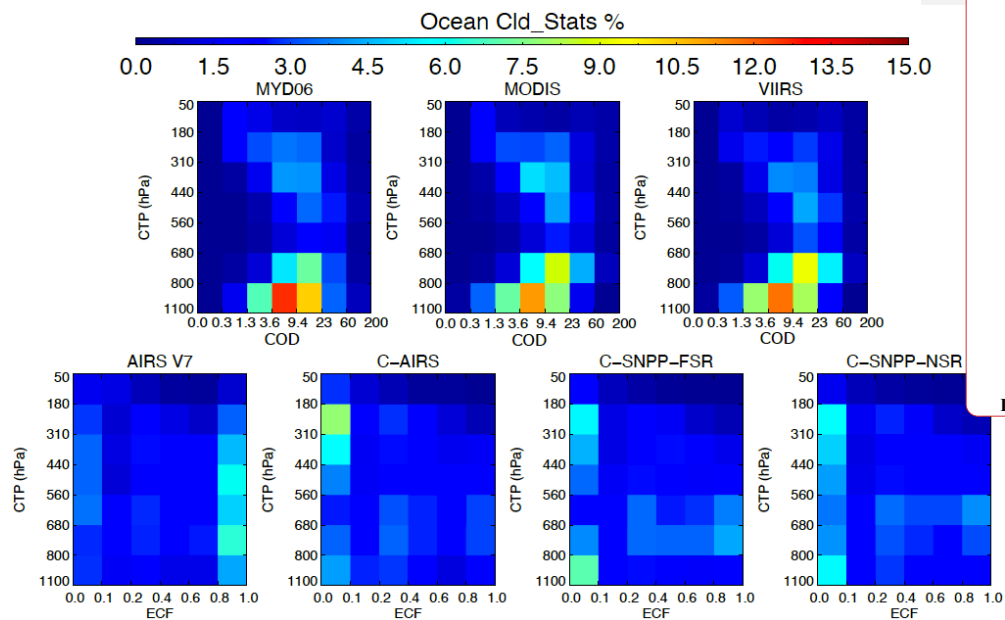
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1038 Figure 10. The 2-dimensional histograms calculated using SNOs from the focus days in the
 1039 tropics (30°N~30°S). The top row shows results for MODIS and VIIRS, for which the ISCCP
 1040 type of COD-CTP joint histograms are presented by summarizing the histograms over individual
 1041 AIRS and CrIS FOV. Note that no averaging over sounder FOV is taken for this comparison.
 1042 From left to right show results of MYD06, *Aqua*-MODIS continuity, and *SNPP*-VIIRS
 1043 continuity cloud products. The bottom row shows results for AIRS and CrIS, and joint
 1044 distributions are calculated on the imager effective CTP and ECF space. From left to right, data

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1046 from AIRS Version 7 (AIRS V7), CLIMCAPS-*Aqua* (C-AIRS), CLIMCAPS-*SNPP* FSR
1047 (C*SNPP*-FSR), and CLIMCAPS-*SNPP* NSR (C-*SNPP*-NSR) are used in the calculation.
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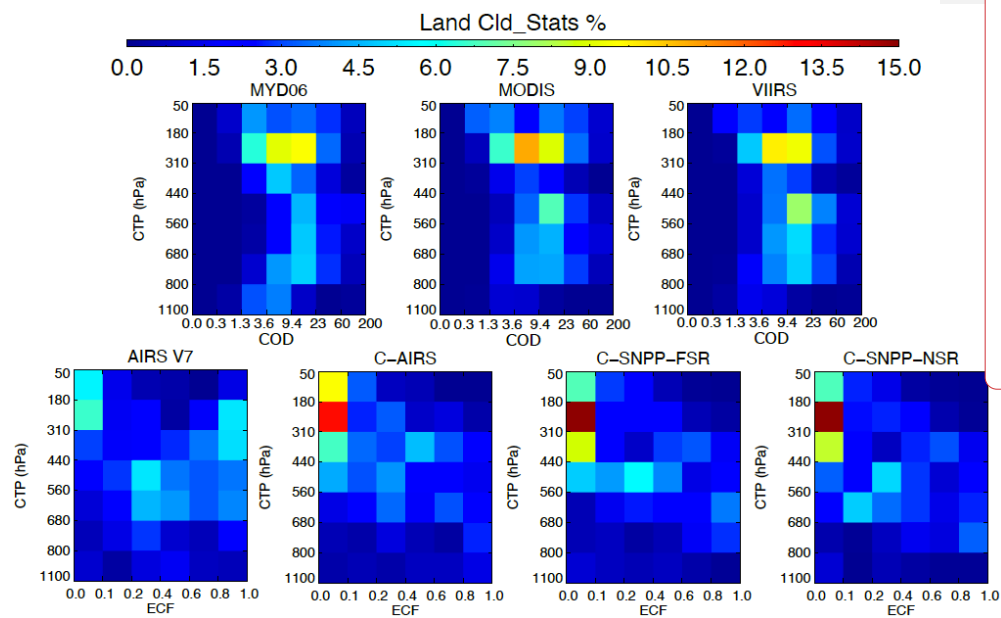
Figure 11. Similar to Fig. 10, except showing results calculated using data over 60°N~60°S ocean. Sounder land fraction < 0.1 is used to determine ocean surfaces.

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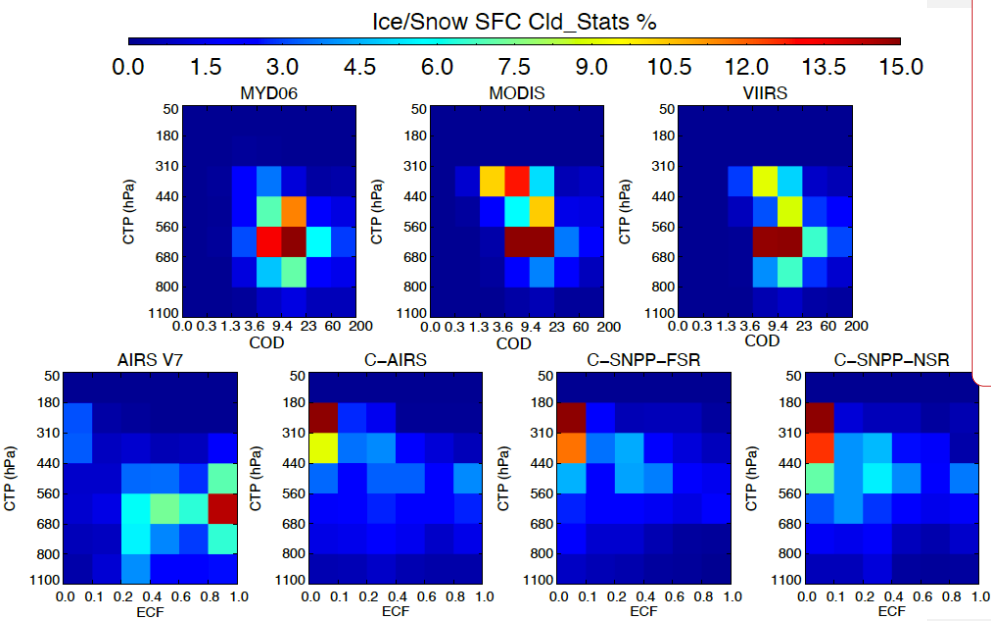
Figure 12. Similar to Figs. 11 and 10, except showing results calculated using data over 60°N~60°S land. Sounder land fraction > 0.9 is used to determine land surfaces.

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Figure 13. Similar to Fig. 10-12, except showing results calculated using data over snow and ice covered surfaces. Sounder retrieved surface classes are used to identify cases.

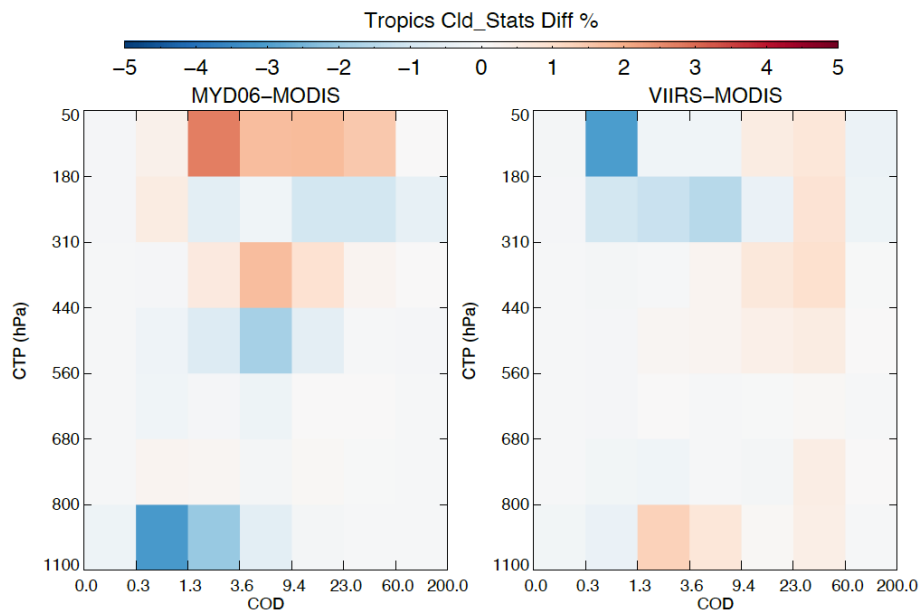
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Figure 14. Differences of the imager CTP-COD cloud histograms in the tropics: between the MYD06 and *Aqua*-MODIS continuity products (left), and between the *Aqua*-MODIS and *SNPP*VIIRS continuity cloud products (right).

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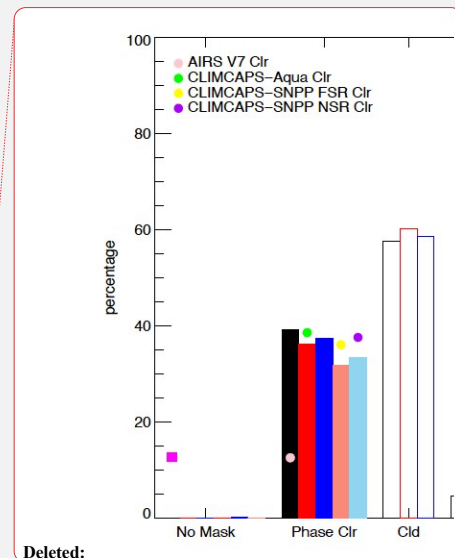
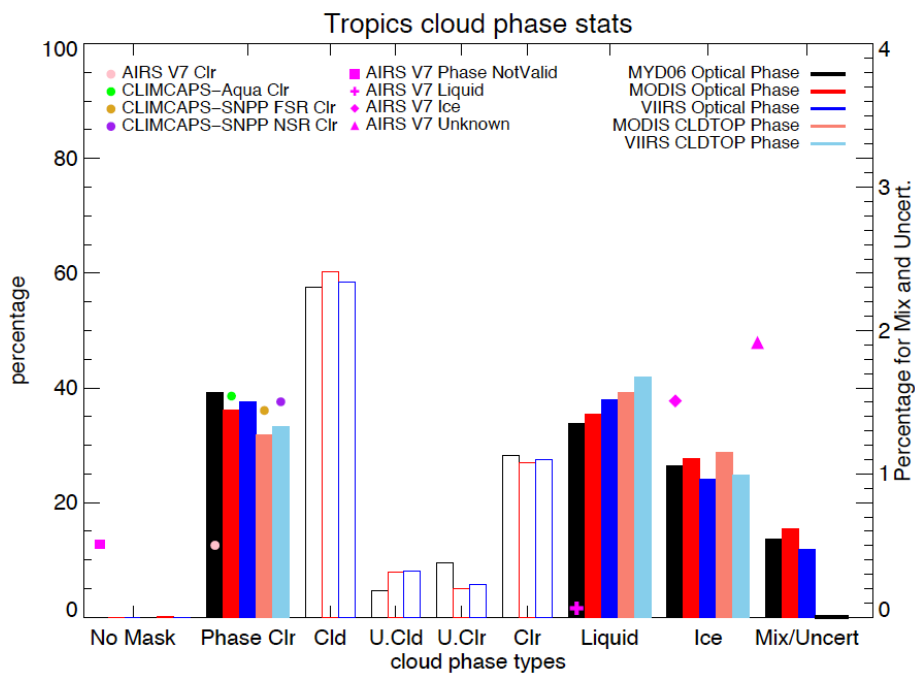
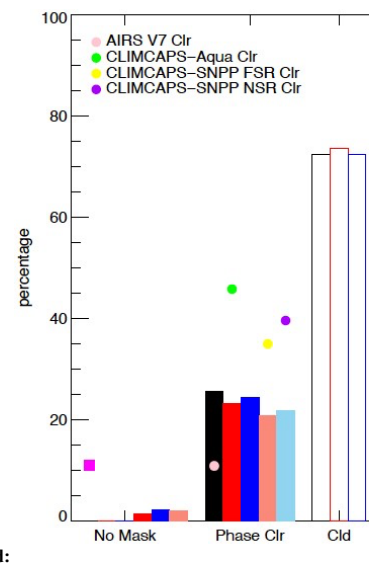
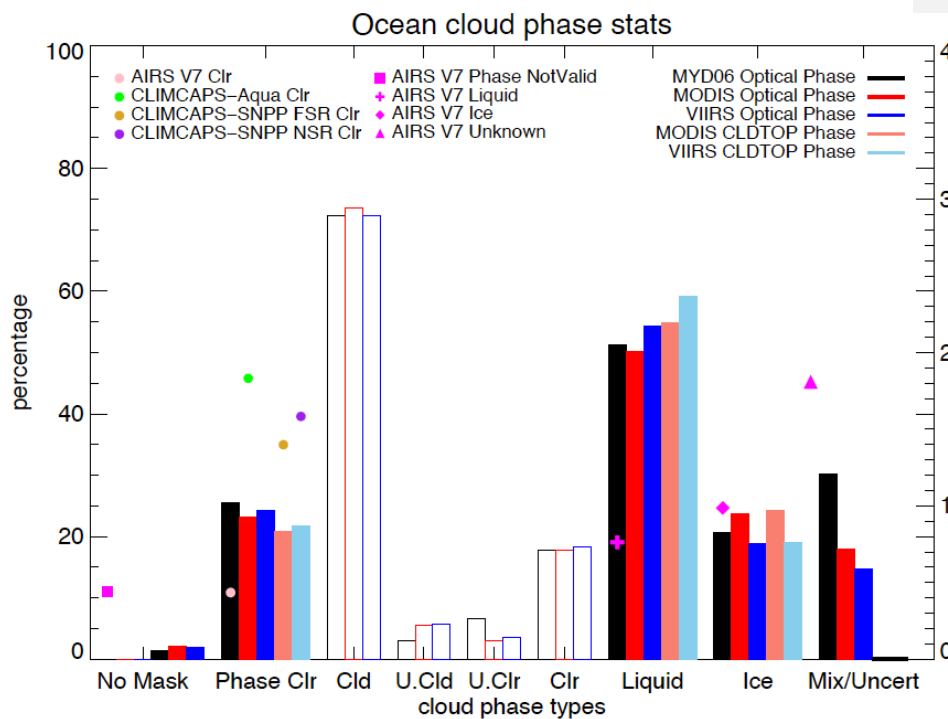


Figure 15. The histograms of cloud thermodynamic phases (solid color bars) and cloud mask (hollow color bars) in the tropics (30°N~30°S) from the imager cloud products calculated using retrievals on SNOs from the seven focus days. The frequency of clear sky detected by IR sounders using thresholds of $ECF < 0.01$ is also shown by colored solid circles. AIRS Version 7 cloud thermodynamic phase is shown by magenta symbols. Color of the bars corresponds with different imager cloud retrievals for cloud mask and cloud thermodynamic phase determined in the optical property retrieval (Cloud_Phase_Optical_Properties): black for MYD06, red for Aqua MODIS continuity products (CLDPROP_MODIS), and blue for SNPP VIIRS continuity products (CLDPROP_VIIRS), respectively. Cloud_Phase_Optical_Properties reports flags indicating cloud mask not determined for pixel (no mask), clear sky (Phase Clr), liquid water cloud (Liquid), ice cloud (ICE), or undetermined phase (Mix/Uncert). Cloud phases reported by Cloud_Phase_Cloud_Top_Properties in the MODIS-VIIRS continuity cloud products are also evaluated and results are shown with pink (MODIS) and light blue (VIIRS) bars, which shows flags indicating cloud free (Phase Clr), water cloud (Liquid), ice cloud (ICE), mixed phase cloud or undetermined phase (Mix/Uncert). Note that the Mix/Uncert phase category for imager products is shown with the y-axis on the right due to its much smaller frequency of occurrence. Cloud mask histograms of Not determined (No Mask), Cloudy (Cld), Uncertain (U. Cld), Probably Clear (U. Clr), and Confident Clear (Clr) are shown in the figure following this color convention but using hollow bars. For IR sounder clear sky frequency, results from AIRS V7 (pink), CLIMCAPS-AIRS (green), CLIMCAPS-SNPP FSR (yellow), and CLIMCAPS-SNPP

1087 NSR (purple) are overlaid on top of the Phase Clr histograms for sounder-imager clear sky
1088 detection comparison.▼
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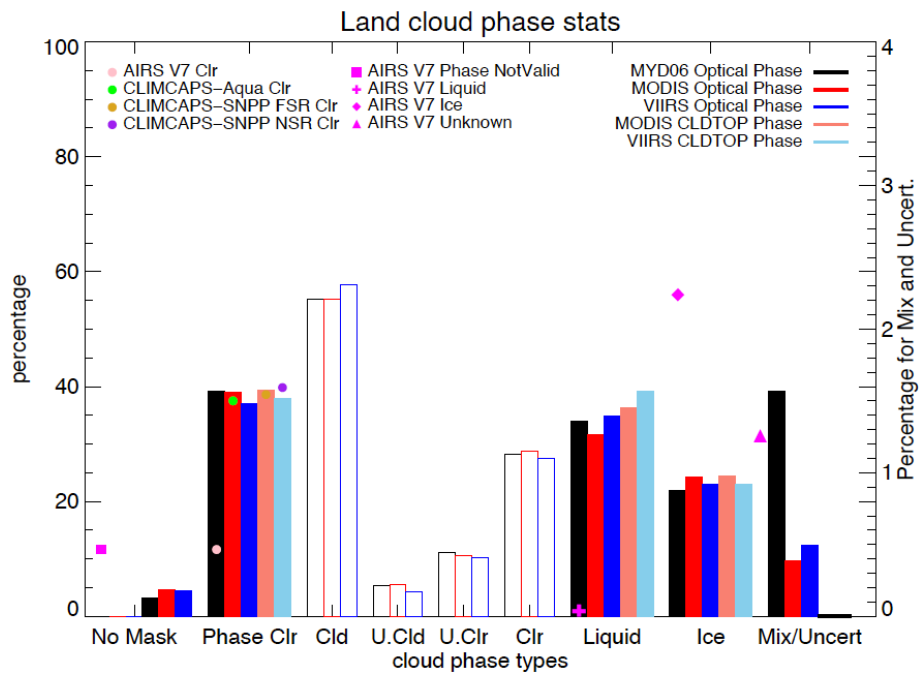
Figure 16. Similar to Fig. 15, except showing results calculated using data over 60°N~60°S ocean. Sounder land fraction < 0.1 is used to determine ocean surfaces.

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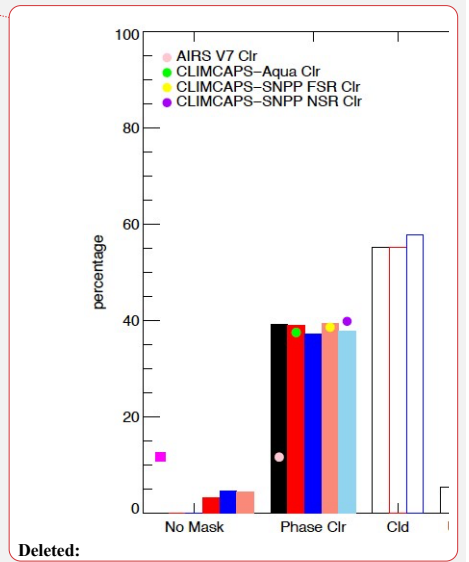
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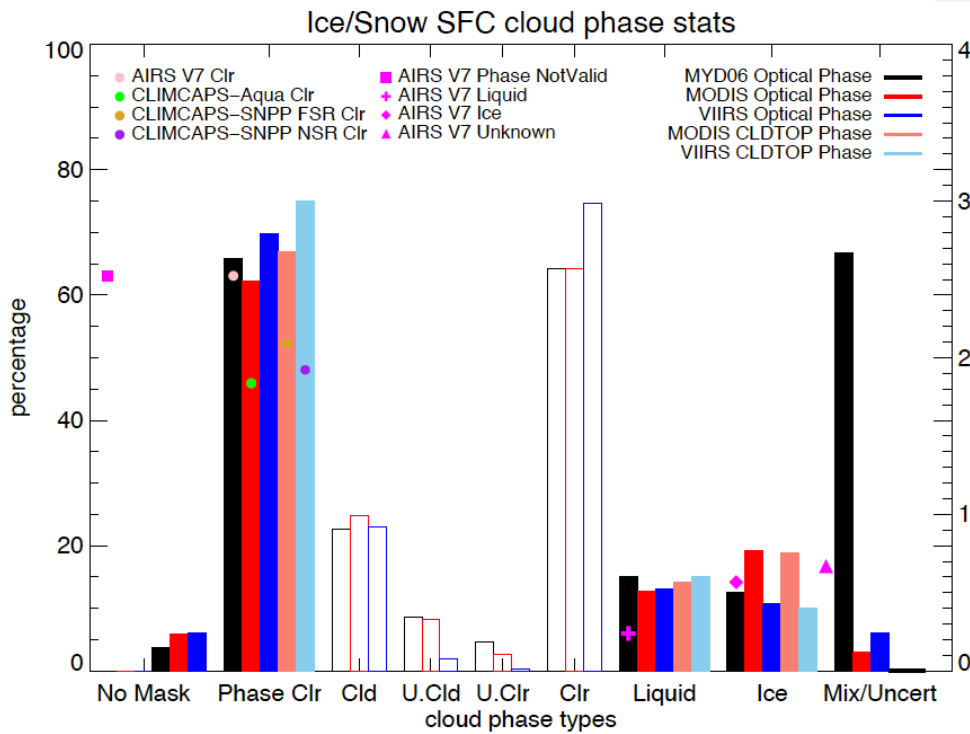
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Figure 17. Similar to Figs. 16 and 15, except showing results calculated using data over 60°N~60°S land. Sounder land fraction > 0.9 is used to determine land surfaces.



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Figure 18. Similar to Figs. 15-17, except showing results calculated using data over snow and ice covered surfaces. Sounder retrieved surface classes are used to identify cases.

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