

1 **Evaluation of globally gridded SST products from NOAA, CMC and UKMeto**
2 **using AIRS and CrIS SST measurements.**

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8
9 **Abstract**

10
11 We compare the daily Sea Surface skin Temperature (SST) derived from calibrated radiances
12 from AIRS (Atmospheric Infrared Sounder) on EOS Aqua and CrIS (Crosstrack Interferometer
13 Sounder) on SNPP between 2013 and 2019 to each other and to three major gridded daily Sea
14 Surface Temperature (SST) products: The NOAA Real Time Global (RTG) SST, the Canadian
15 Meteorological Centre (CMC) SST, and the OSTIA SST from the UK MetOffice. The mean and
16 standard deviation of SST models relative to independent observation with the SST derived from
17 AIRS or CrIS under clear conditions are an effective way of evaluating the fidelity and trends of
18 SST models. AIRS and CrIS show very similar patterns relative to the models, dominated by
19 difference between the skin and the buoy temperatures, but CrIS is on average 0.2 K warmer
20 than AIRS. The daily standard deviation of the difference between the AIRS, the CMC and the
21 OSTIA is 0.4 K, a level previously only seen in SST models relative to the Argo buoys. Between
22 2013 and 2019, all three models show artifacts related to version changes, which limit the
23 interpretation of zonal trend from the 2013-2019 time period at the 3 mK/yr level. The 5 year
24 anomaly trend in (AIRS-CrIS) double difference is relatively latitude zone and day/night
25 consistent at between +2 mK/yr and -1 mK/yr, although this results can't be assumed to be valid
26 for all channels and the full dynamic range of the scenes encountered by AIRS and CrIS.

27
28 **1. Introduction**

29
30 The Sea Surface Temperature (SST) is a key component of climate research, and a potential
31 resource for the evaluation of the accuracy and stability of the SST derived from space-borne
32 sensors. We focus on three daily gridded SST products: The NOAA Real Time Global (RTG)
33 SST (Thiebault et al. 2003), the Canadian Meteorological Centre (CMC) SST (Brasnett 2008)

34 and the Operational Sea Surface Temperature and Sea Ice Analysis (OSTIA) SST (Donlon et al.
35 2012). Between 2002 and 2019 the RTG production used AVHRR data from NOAA 16-19, and
36 more recently AVHRR equivalent VIIRS data from NOAA 20. The CMC SST has been produced
37 on a 0.2 degree grid since 1991 (v2.0), since 2016 (v3.0) it has been available on a 0.1 degree
38 grid. Sensors used in the production of the CMC SST include the AVHRR from NOAA-18, 19,
39 the European Meteorological Operational-A (METOP-A) and Operational-B (METOP-B), and
40 microwave data from the Advanced Microwave Scanning Radiometer 2 (AMSR2) onboard the
41 GCOM-W satellite. The version 2.0 of the OSTIA used in our analysis has been produced on a 5
42 km grid by the UK Meteorological Office since 2013, using AVHRR, the Spinning Enhanced
43 Visible and Infrared Imager (SEVIRI), the Geostationary Operational Environmental Satellite
44 (GOES) imager, the Infrared Atmospheric Sounding Interferometer (IASI, Blumstein et al.
45 2008), the Tropical Rainfall Measuring Mission Microwave Imager (TMI). All three products
46 also use in situ observations of the SST from drifting and moored buoys and ships. Common to
47 these models is that they refer to a daily mean buoy equivalent temperature at 2 meter below the
48 surface. All data used in the production of the SST models have to be converted to buoy
49 equivalent temperatures using a local time, latitude, season and data source dependent model.
50 The resulting model temperatures thus become sensitive to changes in the input data, potentially
51 creating trend artifacts.

52

53 Finding globally distributed and independent truth data to evaluate the stability of these SST
54 models for the global open oceans is difficult. Fiedler et al. 2019 (Table 4) stated a trend of -
55 1.9 ± 1.0 mK/yr (2 sigma) for the CMC, $+10.6 \pm 3.5$ mK/yr (2 sigma) for the OSTIA v1.0 SST
56 relative to the Global Tropical Moored Buoy Array (GT MBA) for the August 2002 - December
57 2007 period. This trend is valid only within about 6 degrees of the equator. The Argo buoys are
58 more uniformly distributed with latitude, but their number increased between 2002 and 2009 by
59 a factor of six, making trends relative to SST models difficult to interpret. This absence of a
60 stable and independent reference provides an opportunity for the use of Atmospheric Infrared
61 Sounder (AIRS, Aumann et al. 2003) and Cross-track Interferometer Sounder (CrIS) data, since
62 neither has been used in the RTG, the CMC and the OSTIA production. The objective of our
63 paper is to use AIRS and CrIS radiance derived SSTs in the evaluation of these three SST
64 products and to each other.

65

66 **2. Data**

67

68 For AIRS we used the near-real-time L1b v5 calibrated data available from the GSFC/DISC
69 since September 2002. For CrIS SNPP we used the retrospective calibration based on algorithms
70 developed in 2018 at the UW-Madison Space Science and Engineering Center: Hank
71 Revercomb, and at the UMBC Atmospheric Spectroscopy Laboratory: Larrabee Strow (2018),
72 Suomi NPP CrIS Level 1B Normal Spectral Resolution V2, Greenbelt, MD, Goddard Earth
73 Sciences Data and Information Services Center (GES DISC), for data between August 2012 and
74 March 2019. The AIRS and CrIS data used the clear footprints saved as part of the ACDS (AIRS
75 Calibration Data Subset) and the equivalent CCDS (CrIS Calibration Data Subset). AIRS and
76 CrIS used the same clear filter concept based on a 3x3 footprint Spatial Coherence Test (SCT,
77 Aumann et al. 2019). Samples where the SCT is less than a 0.5K threshold are identified as SCT
78 clear, typically more than 30,000 each day. While AIRS and CrIS both are in 1:30AM ascending
79 node orbits, their altitudes are 703 km and 825 km, respectively. As a result different locations
80 are identified as clear by AIRS and CrIS. Between May 2013 and March 2019 data from AIRS
81 and CrIS and the three previously identified SST models were available.

82

83 We evaluated the SST models using the atmospheric window channels at 1231.3 cm^{-1} for AIRS,
84 with brightness temperature bt_{1231} , and at 1232.5 cm^{-1} for CrIS, bt_{1232} . The SST derived from
85 bt_{1231} and bt_{1232} are referred to as sst_{1231} and sst_{1232} , or AIRS SST and CrIS SST, depending
86 on the context. The difference between the model SST and the AIRS measured bt_{1231} is
87 typically 1K related to surface emissivity plus 2.5K due to water vapor absorption. For the CrIS
88 bt_{1232} the difference is 1K plus 3.5K. Both sst_{1231} and sst_{1232} use a split window type water
89 vapor correction based on the brightness temperature difference between the 1231.3 cm^{-1} and
90 1227.7 cm^{-1} channels for AIRS, 1232.5 cm^{-1} and 1227.5 cm^{-1} for CrIS. The water vapor
91 correction was regression trained on 1403 open ocean profiles from ECMWF (Aumann et al.
92 2019). For the calculations of the clear AIRS spectra we used SARTA. SARTA is the Radiative
93 Transfer Model (RTM) without clouds developed for AIRS (Strow et al. 2006). SARTA is based
94 on HITRAN2008, updated with the current version of the CKHD3.2 water continuum. For CrIS
95 we used the Community Radiative Transfer Model (CRTM) 2.1.3. CRTM is based on LBLRTM

96 v12.2, with the water continuum absorption based on MT_CKD, version 2.4.

97

98 For each day we calculated the sst1231 and sst1232 for the clear footprints and matched them by
99 their longitude and latitude to the nearest grid point of the models. We then separated the data
100 into eight groups, day and night data, and four latitude bands: 30N-50N, 30N to 0, 0 to 30S, and
101 30S-50S. For each group we calculated the daily mean $=\sum(\text{SST-model})/N$, and standard
102 deviation $\text{stddev}=\sqrt{(\sum(\text{SST-mean-model})^2/N)}$, where N is the daily number of the SST matchups.

103

104 **3. Results**

105

106 We first discuss the results in terms of the comparisons of (AIRS-model) and (CrIS-model) and
107 then as the direct difference (AIRS-CrIS).

108

109 1. (AIRS-model) and (CrIS-model)

110

111 Figure 1 illustrates results in terms of the time series of the daily mean(sst1231-model) for AIRS
112 and CrIS for the night overpasses of the combined 30S-0, 0-30N latitude zone of the oceans,
113 using on average 8700 clear footprint each day from AIRS, 7500 clear footprints per day from
114 CrIS. The patterns in the time series with AIRS and CrIS are very similar, dominated by a peak-
115 to-peak 0.25K seasonal modulation. While AIRS is 0.61K colder than the CMC, CrIS is 0.43 K
116 colder, i.e. CrIS is 0.18K warmer than AIRS in the night 30S-30N zone. This difference will be
117 discussed later.

118

119 AIRS and CrIS SST are expected to be colder at night than the models: The skin is on average
120 0.2K colder than the buoys (Donlon et al. 2002) with a small wind speed dependence
121 (Gentemann & Minnett 2008, Figure 4). In addition, the models refer to the daily mean
122 temperature at about 2 meter depth, while the AIRS SST refer to the skin temperature at the
123 overpass time. On average, the daily mean buoy temperature is 0.21K warmer for the 1:30 PM,
124 0.11K colder for the 1:30 AM overpasses of the 30S-30N zone (Kennedy et al. 2007), but there
125 is a considerable seasonal and zonal variability. It is this seasonal variability which dominates the
126 difference between the AIRS and CrIS surface SST and the SST at 2 meter below the surface in

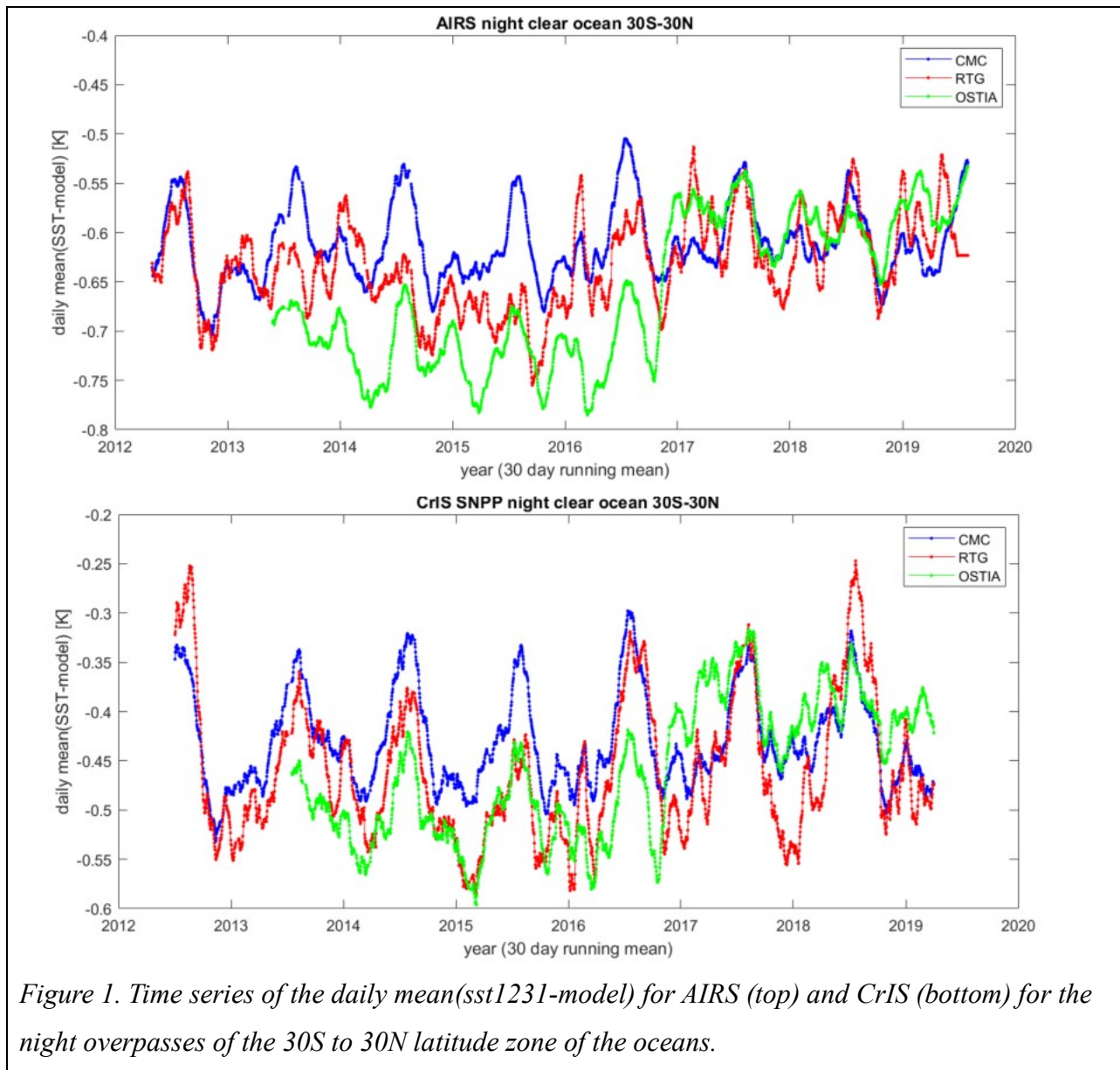


Figure 1. Time series of the daily mean(*sst1231-model*) for AIRS (top) and CrIS (bottom) for the night overpasses of the 30S to 30N latitude zone of the oceans.

128

129 However, the amplitude of the modulation changed between the model and over the years due to
 130 version changes related to the changes in the input data and other changes to the algorithms.

131 Between 2013 and 2019 the peak-to-peak (p-p) amplitude relative to AIRS was about 0.1K for
 132 the OSTIA, 0.15K p-p for the CMC, decreasing to 0.1K in 2016. The RTG modulation was less
 133 regular. The seasonal modulation decreased in the CMC (blue) in 2017, with a 15 mK shift in the
 134 bias, likely associated with a version change. For the RTG (red) the seasonal modulation doubled
 135 starting in 2016, with little change in the mean bias. For the OSTIA (green) the bias after 2017

136 shifted by about 100 mK with little change in the seasonal modulation.

137

138 Figure 2 (top) shows the time series of $\text{stddev}(\text{AIRS-model})$ for the night overpasses of the 30S

139 to 30N latitude zone of the oceans, smoothed with a 30 day running mean. The $\text{stddev}(\text{AIRS-}$

140 OSTIA) (green) and $\text{stddev}(\text{AIRS-CMC})$ are very similar, with neither showing the prominent

141 seasonal pattern in the mean. Figure 2 (bottom) shows the time series of $\text{stddev}(\text{SST-model})$ for

142 the CrIS night overpasses of the 30S to 30N latitude zone of the oceans. The $\text{stddev}(\text{CrIS-}$

143 CMC), averaged for the 5 years, is 0.51K, compared to 0.44K for AIRS.

144

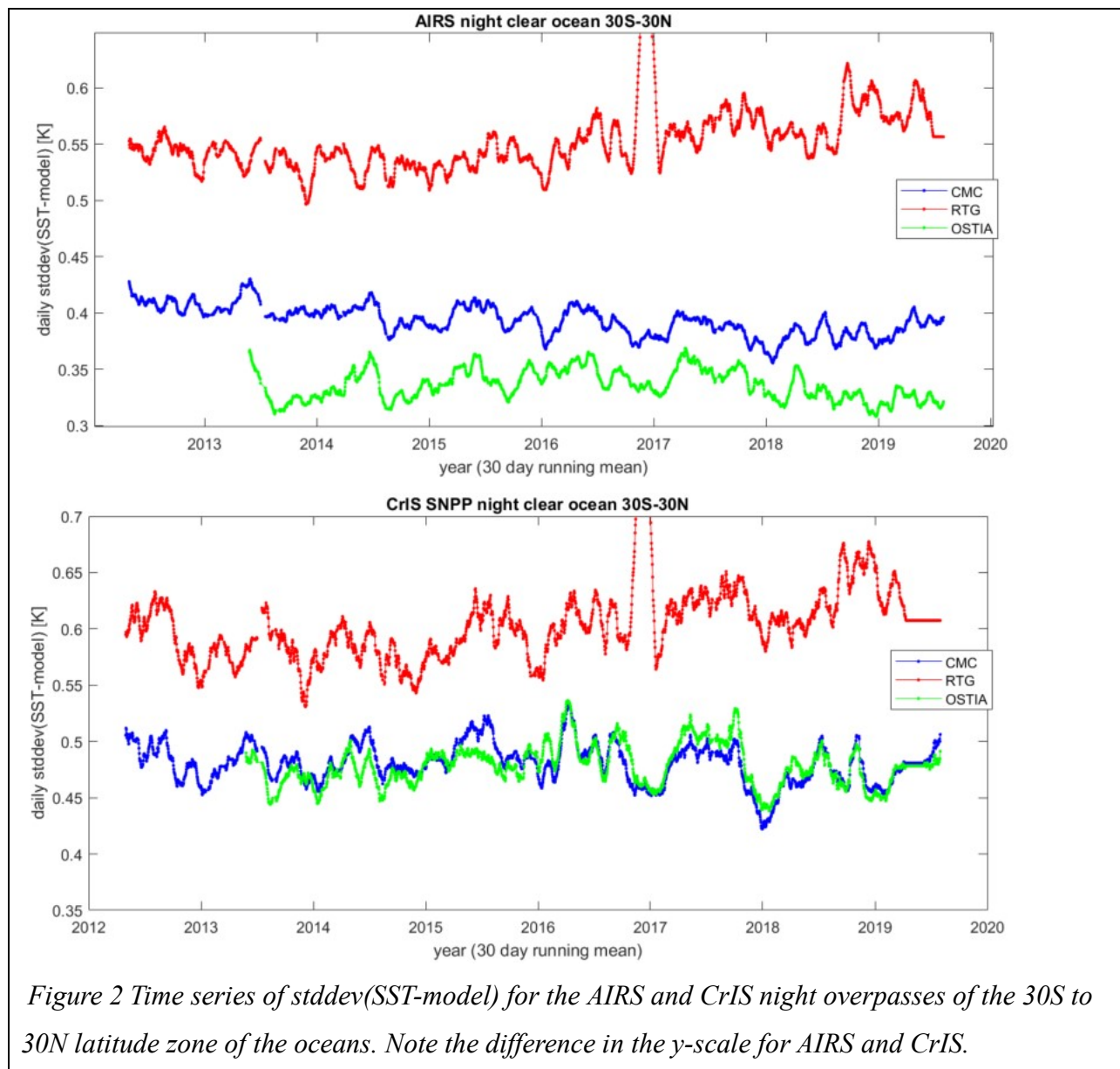
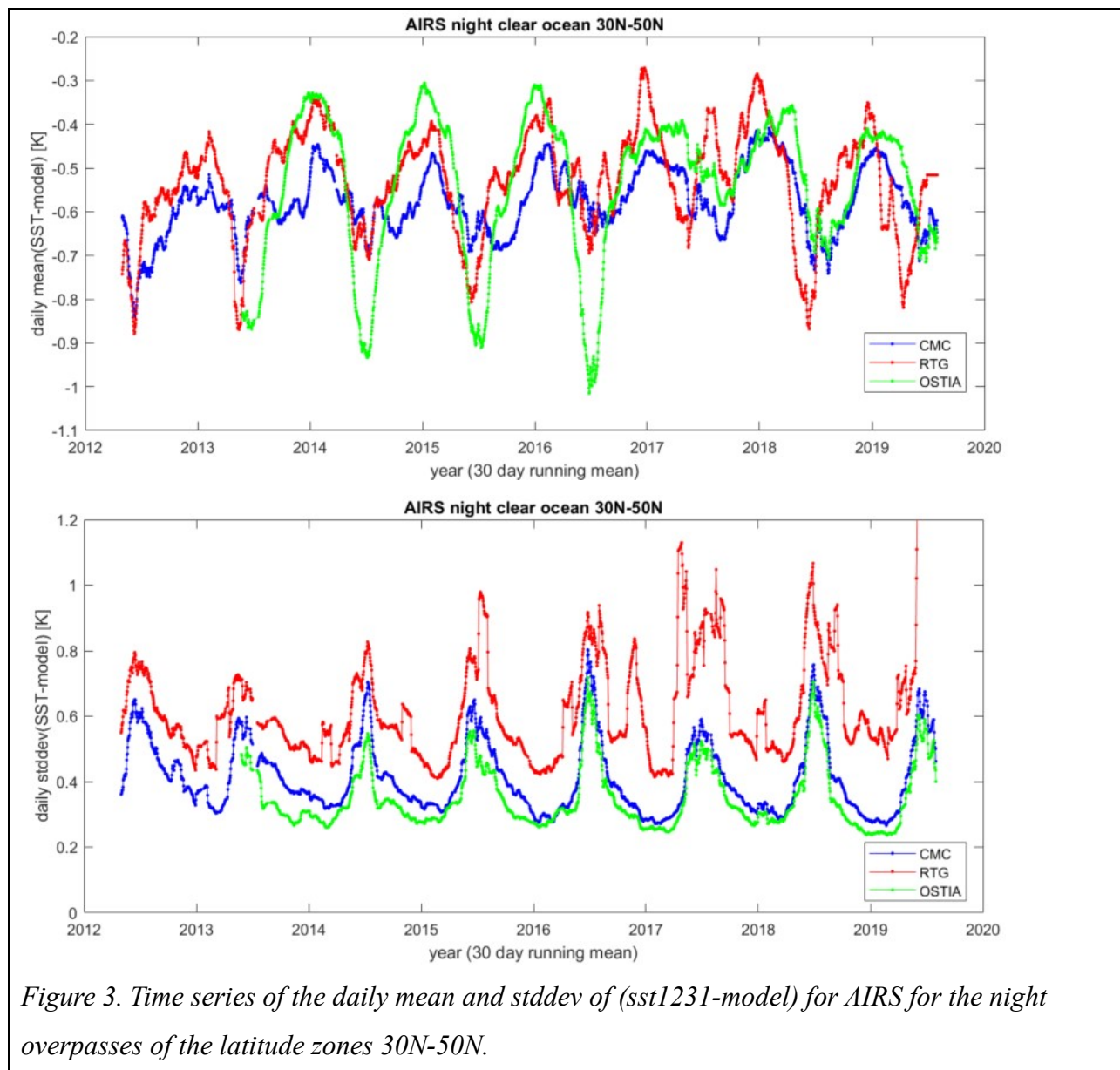


Figure 2 Time series of $\text{stddev}(\text{SST-model})$ for the AIRS and CrIS night overpasses of the 30S to 30N latitude zone of the oceans. Note the difference in the y-scale for AIRS and CrIS.

145

146 The general character, differences and evidence of version changes discussed for the night
147 overpasses of the 30N-30S oceans are seen for the day and night overpasses of other latitude
148 bands. With increasing distance from the equator the seasonal modulation of the diurnal cycle of
149 the difference between the skin and the 2 meter temperatures becomes increasingly more
150 pronounced. Due to the similarity of the results from AIRS and CrIS, we show only the AIRS
151 time series. All results are summarized in Table 1. Figures 3 and 4 shows the time series for the
152 night overpasses of latitudes between 30N-50N and between 30S-50S.

153



154

155 Version changes make the interpretation of the anomaly trends relative to the RTG or the OSTIA
156 even more limited. For this reason Table 1 a. and b. summarizes anomaly trends and bias only for
157 AIRS-CMC, and CrIS-CMC as function of the latitude bands and day and night. AIRS and CrIS
158 both shows a pattern of a positive trend relative to the CMC above the equator, negative below
159 the equator.

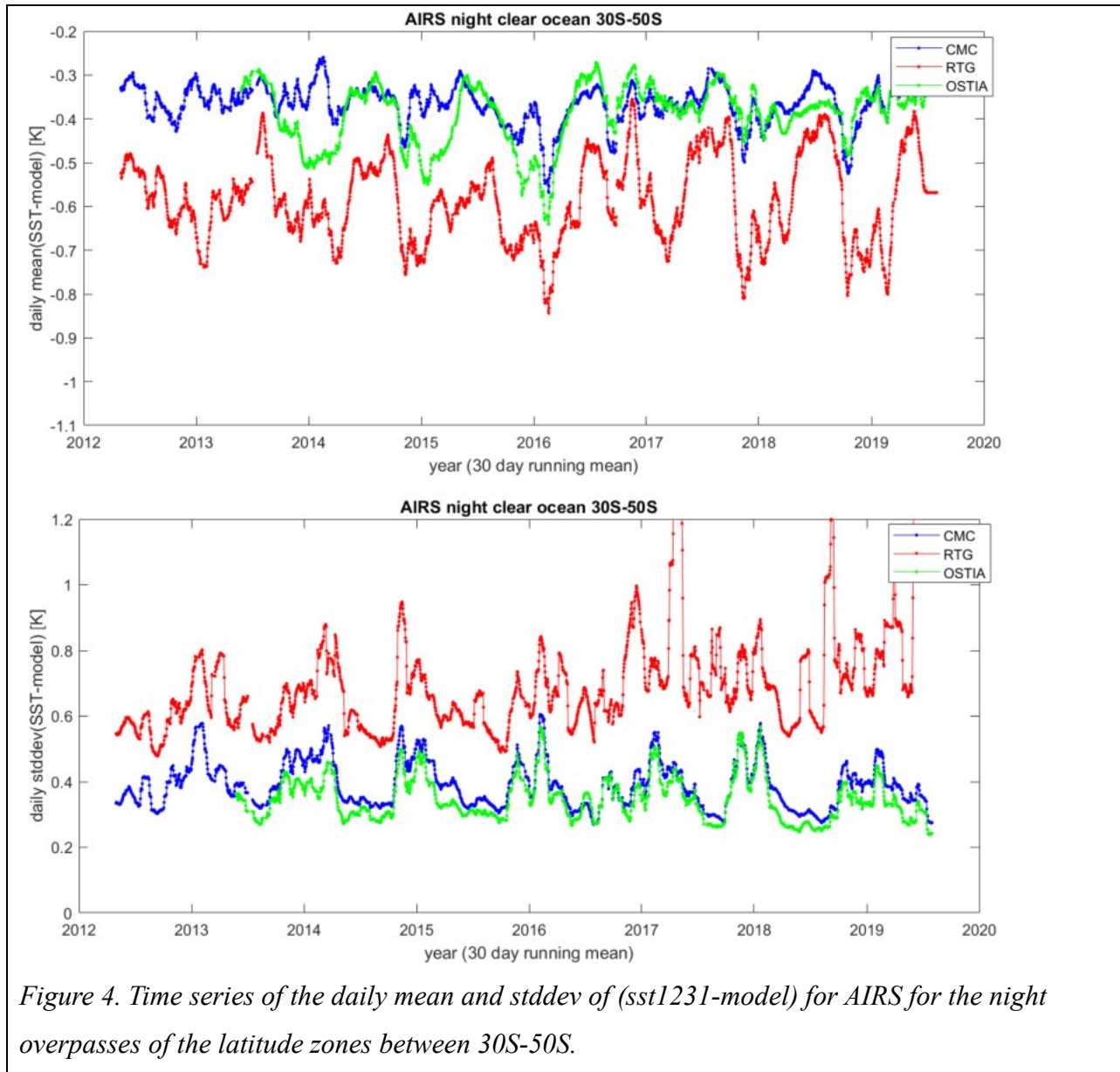


Figure 4. Time series of the daily mean and stddev of (sst1231-model) for AIRS for the night overpasses of the latitude zones between 30S-50S.

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162

AIRS- CMC.airs	Day trend [mK/yr]	Day bias [K]	Day count	Night trend [mK/yr]	Night bias [K]	Night count
30-50N	+ 8.4 ± 1.6	-0.23	2154	+10.5 ± 1.5	-0.56	1848
0-30N	+ 6.9 ± 1.1	-0.21	6935	+ 6.3 ± 0.7	-0.62	4623
30S-0	+ 1.5 ± 0.9	-0.21	8187	+ 0.6 ± 0.8	-0.60	4177
50S-30S	-4.6 ± 1.2	-0.07	2773	-3.2 ± 0.9	-0.34	1892

163 a)

CrIS- CMC.cris	Day trend [mK/yr]	Day bias [K]	Day count	Night trend [mK/yr]	Night bias [K]	Night count
30-50N	+ 7.5 ± 1.7	+ 0.04	1771	+ 7.1 ± 2.0	-0.28	1497
0-30N	+ 4.2 ± 1.5	-0.03	6153	+ 3.6 ± 1.4	-0.42	3920
30S-0	+ 1.0 ± 1.2	-0.08	7274	+ 0.3 ± 1.3	-0.15	2231
50S-30S	-5.1 ± 1.4	+0.22	2200	-2.2 ± 1.4	-0.06	1514

164 b)

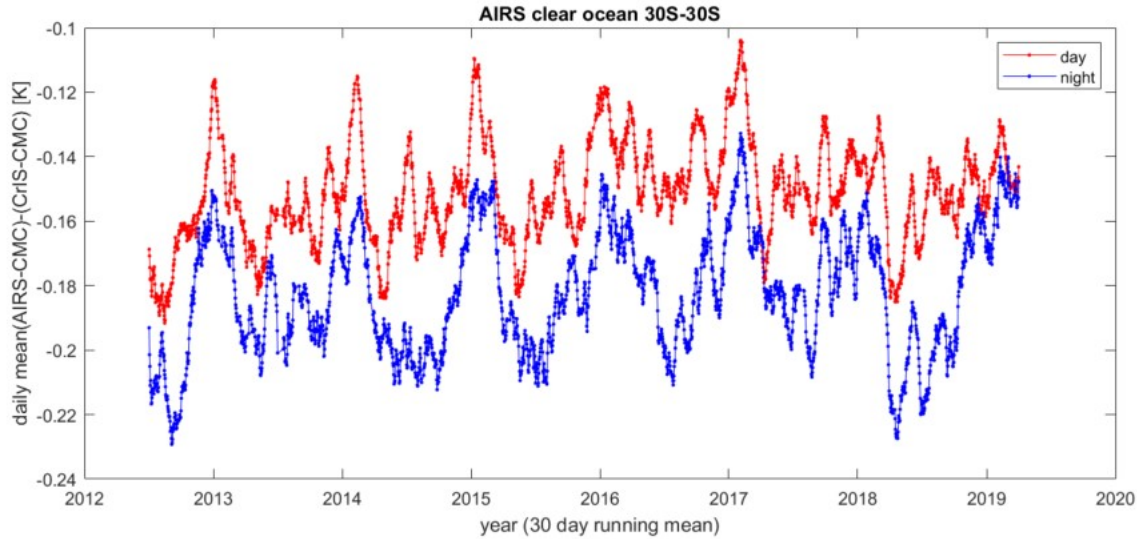
165 Table 1. Anomaly trends and mean in a) (AIRS-CMC.airs) and b) (CrIS-CMC.cris) , and daily
166 mean sample count. The number following ± is the 1 sigma confidence uncertainty of the trend.

167

168 2. Direct AIRS/CrIS comparison.

169

170 Between May 2012 and March 2019 AIRS and CrIS data were both available. This allow a direct
171 comparison of AIRS and CrIS derived SST. We already noted that the typically 40,000 clear
172 footprints identified each day are from different locations. In the analysis of daily mean of the
173 double difference (AIRS-CMC.airs) minus (CrIS-CMC.cris), where CMC.airs and CMC.cris are
174 the CMC SST values closest to the AIRS and CrIS locations, respectively, the location
175 differences and model artifacts cancel. However, the time series of the double difference for the
176 30S-30N latitude zones, day (red), night (blue), Figure 5, still shows a seasonal pattern, more
177 pronounced at night than during the day. This requires further analysis. Table 2 shows the
178 anomaly trend and bias for the AIRS-CrIS double difference.



179

180 *Figure 5. (AIRS-CMC.airs) - (CrIS-CMC.cris) double difference*

181

(AIRS-CMC.airs) - (CrIS-CMC.cris)	Day trend [mK/yr]	Mean Day bias [K]	Mean Day count	Night trend [mK/yr]	Mean Night bias [K]	Mean Night count
30-50N	+ 0.9 ± 1.2	-0.27	1771	+ 3.1 ± 1.4	-0.28	1497
0-30N	+ 3.0 ± 0.9	-0.17	6153	+ 2.3 ± 1.1	-0.20	3920
30S-0	+ 0.7 ± 0.7	-0.14	7274	+ 0.0 ± 1.0	-0.17	2231
50S-30S	+ 0.4 ± 0.9	-0.29	2200	-1.4 ± 1.0	-0.29	1514

182

183 Table 2. (AIRS-CMC.airs) - (CrIS-CMC.cris) double difference mean and trends. The number
184 following ± is the 1 sigma confidence uncertainty of the trend.

185

186 4. Discussion

187

188 The daily stddev(AIRS-CMC) and (AIRS-OSTIA), 0.4 K average between 2013 and 2019,
189 indicate a level of agreement between the models and independent SST observations seen
190 previously only relative to the Argo buoys in the 2000-2011 period (Fiedler et al. 2019). The
191 AIRS stddev relative to OSTIA and CMC shows a small but steady decrease, presumably related
192 to improvements in the models. A degradation of the RTG is seen as the steadily increasing
193 stddev starting in 2016, until the RTG production was ended in 2019.

194

195 The evaluation of trends from the data between 2013 and 2019 is impacted by version changes in
196 the CMC, start-up problem and version changes in the OSTIA, and the steady degradation of the
197 RTG throughout our analysis period. Even the small, previously noted 15 mK version change
198 related shift in the CMC would create a 3 mK/yr trend artifact between 2013 and 2019. Table 1
199 shows that the one sigma bias anomaly trend uncertainties for AIRS and CrIS relative to the
200 CMC range from +8 mK/yr to -5 mK/yr depending on latitude. For a longer time series, i.e.
201 AIRS-CMC between 2002 and 2020, the 15 mK shift due to a version change in 2017 would
202 only create a 1 mK/yr trend artifact, i.e. trends larger than 1 mK/yr would be significant.

203

204 Trends in the AIRS-CMC and CrIS-CMC are dominated by CMC version changes. In trends
205 based on the double difference, version changes in the CMC cancel, and the AIRS-CrIS trends
206 range from +3 and -1 mK/yr, relatively latitude zone and day/night consistent. These trends are
207 very low, but pertain only to the channels used for the SST derivation. There is no assurance that
208 they are valid for all channels and the full dynamic range of the scenes encountered by AIRS and
209 CrIS.

210

211 CrIS is typically about 200 mK warmer than AIRS at all latitudes. This difference could be
212 related to the combination of a bias in the absolute calibration of AIRS and CrIS and a difference
213 in the RTM used for the water vapor correction of the SST. The 200 mK difference between
214 AIRS and CrIS appears to be dominated by an absolute calibration difference, consistent in
215 magnitude with expectations.

216

217 Calibration bias: AIRS is a grating array spectrometer. As such the SST is calculated from
218 the signal measured by a single detector at 1231.3 cm^{-1} (8 microns). Pagano et al. (2018, Figure
219 4) show a 2-sigma absolute radiometric uncertainty for the channels near 8 microns of 80 mK for
220 scene temperatures in the 280-300K range for the L1b V5 calibration. CrIS is an interferometer
221 with a 3x3 focal plane array, where 9 independent detectors measure the signal for a wide
222 wavelength range. In the previous figures we showed the daily mean from CrIS averaged over all
223 nine detectors. If we instead were to calculate the daily mean for each detector, we find a 140
224 mK day/night independent difference between the warmest and coldest detector. The difference
between AIRS and CrIS seen in our data could thus be simply related to the absolute calibration

225 difference.

226 Water vapor correction: We used SARTA with the latest CKHD 3.2 water continuum. For
227 CrIS we used the CRTM 2.1.3 with a 2008 vintage MT_CK 2.4. The effect of the water vapor
228 absorption on for the CrIS bt1232 is typically 3.5K. The water vapor continuum in MTCK 2.4
229 was decreased in more recent editions. CRTM would have to be 6% too strong to create a warm
230 bias in sst1232 of 200 mK ($3.5 \times 0.06 = 0.2K$). A 2% overcorrection is non unlikely, which would
231 explain 70 mK of the difference between AIRS and CrIS. A CRTM overcorrection of the water
232 vapor continuum may explain the consistently slightly larger stddev of CRIS relative to the CMC
233 and OSTIA models.

234

235 **4. Summary**

236

237 The mean and standard deviation of SST models relative to independent observation with the
238 SST derived from AIRS or CrIS under clear conditions are an effective way of evaluating the
239 fidelity and trends of SST models. AIRS and CrIS show very similar patterns relative to the
240 models. CrIS being 200 mK warmer than AIRS appears to be consistent with the combination of
241 expected water vapor continuum correction and absolute calibration differences. The CMC and
242 OSTIA are major improvements over the RTG. The daily standard deviation of the difference
243 between the AIRS, the CMC and the OSTIA of 0.4 K is a level previously only seen in SST
244 models relative to the Argo buoys. Between 2013 and 2019, all three models show artifacts
245 related to version changes, which limit the interpretation of zonal trend from the 2013-2019 time
246 period. Only the CMC is available for the 2013 – 2019 AIRS and CrIS availability. The 5 year
247 anomaly trends in (AIRS-CrIS) double difference avoid the effect of version changes. There we
248 find trends between at +3 and -1 mK/yr. This requires additional analysis.

249

250 **Acknowledgments**

251

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253 under contract with NASA. Sergio DeSouza-Machado, U. Maryland Baltimore County, provided
254 support with the SARTA RTM. We used the prototype of the daily AIRS Calibration Data
255 Subset (ACDS v7), which will be available free of charge from the GSFC DISC. Currently the

256 ACDS v5 is available at
257 https://disc.gsfc.nasa.gov/datasets/AIRXBCAL_005/summary?keywords=AIRXBCAL.
258 An equivalent CrIS Calibration data subset, conceptually equivalent to the ACDS v7, will be
259 available from the GSFC/DISC soon.

260 Between 2001 and 2016, the RTGSST was freely available on a 0.5 degree grid via FTP at
261 ftp://polar.ncep.noaa.gov/pub/history/sst/rtg_low_res/, for example, [rtg_sst_grb_0.5.20020831](ftp://polar.ncep.noaa.gov/pub/history/sst/rtg_low_res/rtg_sst_grb_0.5.20020831).
262 More recently it was available from https://polar.ncep.noaa.gov/sst/rtg_high_res/. As of 2019 its
263 production has been discontinued. The CMC and OSTIA data are freely available from
264 [https://podaac.jpl.nasa.gov/dataset/CMC0.2deg-CMC-L4-GLOB-](https://podaac.jpl.nasa.gov/dataset/CMC0.2deg-CMC-L4-GLOB-v2.0?ids=&values=&search=CMC)
265 [v2.0?ids=&values=&search=CMC](https://podaac.jpl.nasa.gov/dataset/CMC0.2deg-CMC-L4-GLOB-v2.0?ids=&values=&search=CMC) and [https://podaac.jpl.nasa.gov/dataset/OSTIA-UKMO-L4-](https://podaac.jpl.nasa.gov/dataset/OSTIA-UKMO-L4-GLOB-v2.0?ids=&values=&search=%22OSTIA-UKMO-L4-GLOB-v2.0%22)
266 [GLOB-v2.0?ids=&values=&search=%22OSTIA-UKMO-L4-GLOB-v2.0%22](https://podaac.jpl.nasa.gov/dataset/OSTIA-UKMO-L4-GLOB-v2.0?ids=&values=&search=%22OSTIA-UKMO-L4-GLOB-v2.0%22)

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