



1 **Heterogeneity and Chemical Reactivity of the Remote Troposphere defined by**
2 **Aircraft Measurements**

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39 Abstract

40 The NASA Atmospheric Tomography (ATom) mission built a photochemical climatology of air
41 parcels based on in situ measurements with the NASA DC-8 aircraft along objectively planned
42 profiling transects through the middle of the Pacific and Atlantic Oceans. ATom measured
43 numerous gases and aerosols, particularly the gaseous species driving the chemical budgets of O₃
44 and CH₄: i.e., O₃, CH₄, CO, C₂H₆, higher alkanes, alkenes, aromatics, NO_x, HNO₃, HNO₄,
45 peroxyacetylnitrate, other organic nitrates, H₂O, HCHO, H₂O₂, and CH₃OOH. From the 10 s (2
46 km) merged observations, a modeling data stream (MDS) based on observations of the core
47 species, consisting of 146,494 distinct air parcels has been constructed from the 4 ATom
48 deployments, providing a continuous data stream for initializing global chemistry models and
49 calculating the 24-hour chemical tendencies. Tendencies derived from 6 chemistry models using
50 the ATom-1 MDS tend to agree and show a highly heterogeneous troposphere where globally
51 10% of the parcels control as much as 40% of the budget of O₃ and CH₄. Surprisingly, modeled
52 probability distributions (100-km cells) match ATom statistics (2 km parcels), indicating that the
53 majority of the observed heterogeneity can be resolved with current global chemistry models. On
54 the other hand, the models' own chemical climatologies underestimate O₃ production below 4 km
55 in both Pacific and Atlantic basins because they have lower NO_x levels than observed.

56

57 1. Introduction

58

59 The NASA Atmospheric Tomography (ATom) mission completed a four-season deployment.
60 Each deployment flying from the Arctic to Antarctic and back, traveling south through the middle
61 of the Pacific Ocean, across the Southern Ocean, and then north through the Atlantic Ocean, with
62 near-constant profiling of the marine troposphere from 0.2 to 12 km altitude (See Figure S1). The
63 DC8 was equipped with in situ instruments that documented the chemical composition and
64 conditions at time intervals ranging from <1 to about 100 seconds (Wofsy et al., 2018). ATom
65 measured hundreds of gases and aerosols, providing information on the chemical patterns and
66 reactivity in the vast remote ocean basins, where most of the destruction of tropospheric ozone
67 (O₃) and methane (CH₄) occurs. Reactivity is defined here as in Prather et al. (2017) to include
68 the production and loss of O₃ (P-O3 and L-O3, ppb/d) and loss of CH₄ (L-CH4, ppb/d). Here we
69 report on this model-derived product that was proposed for ATom, the daily averaged reaction
70 rates determining the production and loss of O₃ and loss of CH₄ for 10 s averaged air parcels. We
71 calculate these rates with 3D chemical models that include variations in clouds and photolysis,
72 and then assemble the statistical patterns describing the heterogeneity (i.e., high spatial
73 variability) of these rates and the underlying patterns of reactive gases that are responsible for the
74 net reactivity.

75 Tropospheric O₃ and CH₄ contribute to climate warming and global air pollution (Stocker et al.,
76 2013). Their abundances in the troposphere are controlled largely by tropospheric chemical
77 reactions. Thus, chemistry-climate assessments seeking to understand past global change and
78 make future projections for these greenhouse gases have focused on the bulk tropospheric rates of
79 production and loss, and how these reactivities are distributed in large zones throughout the
80 troposphere (Griffiths et al., 2020; Myhre et al., 2013; Naik et al., 2013; Prather et al., 2001;
81 Stevenson, et al., 2006; Stevenson, et al., 2013; Stevenson, et al., 2020; Voulgarakis et al., 2013;
82 Young et al., 2013). The models used in these assessments disagree on these overall CH₄ and O₃
83 reactivities, and resolving the cause of such differences is stymied because of the large number of
84 processes involved and the resulting highly heterogeneous distribution of chemical species that
85 drive the reactions. Simply put, the models use emissions, photochemistry, and meteorological
86 data to generate the distribution of key species such as nitrogen oxides (NO_x = NO + NO₂) and



87 hydrogen peroxide (HOOH) (step 1) and then calculate the CH₄ and O₃ reactivities from these
88 species (step 2). Stratospheric studies such as Douglass et al. (1999) have provided a quantitative
89 basis for testing chemistry and transport, and defining model errors; but few of these studies have
90 tackled the problem of modeling the heterogeneity of tropospheric chemistry. The major model
91 differences lie in the first step, because when we specify the mix of key chemical species, most
92 models agree on the CH₄ and O₃ chemical budgets (Prather et al., 2018). The intent of ATom was
93 to collect an atmospheric sampling of all the key species, the statistics defining their spatial
94 variability, and that of the reactivities of CH₄ and O₃.

95 Many studies have explored the ability of chemistry-transport models (CTMs) to resolve finer
96 scales including pollution layers (Eastham et al., 2017; Rastigejev et al., 2010; Tie et al., 2010;
97 Young et al., 2018; Zhuang et al., 2018), but these have not had the observations of chemical
98 statistics to evaluate model performance. In a great example of using chemical statistics, Yu et al.
99 (2016) used 60 s data (~12 km) from the SEAC⁴RS aircraft mission to compare cumulative
100 probability distribution functions of NO_x, O₃, HCHO, and isoprene over the Southeast US with
101 the GEOS-Chem CTM run at different resolutions. They identified clear biases at the high and
102 low ends of the distribution, providing a new test of models based on the statistics rather than
103 mean values. Heald et al. (2011) gathered high resolution profiling of organic and sulfate
104 aerosols from 17 aircraft missions and calculated statistics (mean, median, quartiles) but only
105 compared with the modeled means. The HIAPER Pole-to-Pole Observations (HIPPO) aircraft
106 mission (Wofsy, 2011) was a precursor to ATom with regular profiling of the mid-Pacific
107 including high frequency 10 s sampling that identified the small scales of variability throughout
108 the troposphere. HIPPO measurements were limited in species, lacking O₃, NO_x and many of the
109 core species needed for reactivity calculations. ATom, with a full suite of reactive species and
110 profiling through the Atlantic basin, provides a wealth of chemical statistics to challenge the
111 global chemistry models.

112 Our task here is the assembly of the modeling data stream (MDS) that provides flight-wise
113 continuous 10 s data (air parcels) for the key reactive species as described in Methods and
114 Supplementary Information. The MDS is based on direct observations and interpolation methods
115 to fill gaps. From the MDS we have six chemical models calculate the 24 h reactivities,
116 producing a reactivity data stream (RDS) as described in Methods and Supplementary
117 Information. In the Results section, we examine the statistics of reactivity over the Atlantic and
118 Pacific Oceans, focusing on air parcels with high reactivity, for example, 10% of the parcels
119 produce 24-35% of total reactivity. We compare these ATom-1 statistics, species and reactivities,
120 with August climatologies from six global chemistry models. Overall, ATom-1 shows a more
121 reactive tropical and mid-latitude summer troposphere than found in most models' climatologies
122 due primarily to higher NO_x levels. Conclusions and discussion of opportunities for the ATom
123 MDS+RDS datasets are presented in the Discussion. ATom measurements should provide
124 fundamental performance metrics for all global chemistry models, metrics that define the
125 chemical statistics of the troposphere and address the cause of differences in the O₃ and CH₄
126 budgets currently seen across the models. Use of the ATom MDS+RDS can help us identify
127 patterns in key species that must be matched to capture the reactive chemistry of the remote
128 troposphere.

129 **2. Models and data**

130

131 **2.1 The modeling data stream**



132 The ATom mission was designed to collect a multi-species, detailed chemical climatology that
133 documents the spatial patterns of chemical heterogeneity throughout the remote troposphere.
134 Supplemental Figure S1 maps the 48 research flights, and the Supplementary Information has
135 tables summarizing each flight. We required a complete set of key species in each air parcel to
136 initialize models to be able to calculate the CH₄ and O₃ reactivities. We choose the key reactive
137 species (H₂O, O₃, CO, CH₄, NO_x, NO_xPSS, HNO₃, HNO₄, PAN, CH₂O, H₂O₂, CH₃OOH, acetone,
138 acetaldehyde, C₂H₆, C₃H₈, i-C₄H₁₀, n-C₄H₁₀, alkanes, C₂H₄, alkenes, C₂H₂, C₅H₈, benzene,
139 toluene, xylene, CH₃ONO₂, C₂H₅ONO₂, RONO₂, CH₃OH) and then add corollary species or other
140 data indicative of industrial or biomass-burning pollution or atmospheric processing (HCN,
141 CH₃CN, SF₆, relative humidity, aerosol surface area (4 modes), and cloud indicator). We choose
142 10 s averages for our air parcels as a compromise and because the 10 s merged data is a standard
143 product (Wofsy et al., 2018). A few instruments measure at 1 s intervals, but the variability at
144 this scale is not that different from 10 s averages (Figure S2). Throughout ATom, gaps occur in
145 individual species on a range of times scales due to calibration cycles, sampling rates, and
146 instrument malfunction. The creation of the MDS used a range of methods to fill these gaps and
147 assigned a flag index to each species, allowing users to avoid parcels with less reliable methods,
148 on a species-by-species basis. The approach to gap filling is described in detail in the
149 Supplementary Information.

150 2.2 The reactivity data stream.

151 The concept of using an MDS to initialize 3D global chemistry models and calculate an RDS was
152 developed in the pre-ATom methodology papers (Prather et al., 2017; Prather et al., 2018). In
153 this paper, we use the 6 models for their August chemical statistics, and use 5 of them plus a box
154 model to calculate the reactivities (i.e., chemical tendencies) from the ATom-1 MDS, see Table 1.
155 Large model differences are clearly seen in the parcel-by-parcel root-mean-square (RMS)
156 differences summed across the RDS_R0 parcels in Table 3. Even when models adopt standard
157 kinetic rates and cross sections (i.e., Burkholder et al., 2015), the number of species and chemical
158 mechanisms included, as well as the treatment of families of similar species or intermediate short-
159 lived reaction products, varies across models. For example, UCI considers about 32 reactive
160 gases, whereas GC and GMI have over 100, and F0AM typically runs with more than 600. The
161 other major difference across models is photolysis, with models having different cloud data and
162 different methods for calculating photolysis rates in cloudy atmospheres (Hall et al., 2018).

163 3. Results

164 In developing the MDS, we mapped higher- and lower-frequency measurements onto 10 s air
165 parcels and then tested our gap-filling procedures, see Methods. Examination of higher-
166 frequency 1 s data available for O₃ and H₂O shows that 10 s averaging preserves most of the
167 heterogeneity (Figure S2). Overall we tested 3 successive MDS revisions designated R0, R1, and
168 R2. MDS_R0 for ATom-1 was distributed to the full ATom science team, and we received the
169 modeled RDS_R0 from 5 of the original 6 global models (Prather et al., 2017) plus 1 box model,
170 see Table 1. Three central models (GC, GMI, UCI) showed excellent agreement. The MDS_R1
171 release corrected the primary mistake in creating MDS_R0 (i.e., use of photostationary NO_x
172 instead of observed NO_x) and was re-run with two models (GMI, UCI), again showing excellent
173 agreement. MDS_R2 optimized and tested the treatments of gap filling and lower limit of
174 detection, along with other quality controls. Because of previous model agreement and the
175 relatively small changes from R1 to R2, we calculated RDS_R2 only with one model (UCI). The
176 R1 and R2 NO_x values are 25% larger on average than the R0 values (unweighted mean of 66 vs.
177 52 ppt), and this affects primarily P-O3 and secondarily L-CH4. For UCI, the RDS_R1 and
178 RDS_R2 differences are very small and so we show here: RDS_R0 for all 6 models, RDS_R1 for



179 GMI, and RDS_R2 for UCI as our best estimate. The MDS_R2 (and higher revisions) for ATom-
180 2/3/4 are now public (<https://espo.nasa.gov/atom/content/ATom>); and while their RDS files are
181 posted in the archive, the reactivities are not analyzed.

182 The reactivities for 3 large domains (Global, Pacific, Atlantic) from the 6 models' RDS_R0 are
183 summarized in Tables 2 and S8. In our analysis, the ATom 10 s parcels are weighted to achieve
184 uniform sampling from 200-1000 hPa and by cosine of latitude. Global reactivities are less than
185 those of the ocean basins because they include high-latitude observations over the Southern
186 Ocean and the Arctic. We include the statistics from UCI using alternate years (1997 and 2015
187 versus the standard 2016) to show the effect of different cloud fields, and also from the GMI
188 RDS_R1 and UCI RDS_R2. The statistical properties of the reactivities include: mean value;
189 percent of total reactivity in the top 10% parcels (Table 2); median; mean value of top 10%
190 parcels; percent of total reactivity in the top 50% and top 3% parcels; mean photolysis rates
191 (Table S8). These high-R parcels (3%, 10% and 50%) are determined separately for each model,
192 each reactivity (R), and each domain.

193 The Atlantic is more reactive on average than the Pacific for all 3 Rs (P-O3, L-O3, L-CH4);
194 however, the Pacific has a much more peaked high-R distribution as seen by the much higher
195 fraction of the total contained in the top 10% (27-36% vs. 19-28%). The average R in the top
196 10% of parcels is typical 4 to 10 times the mean value; except again for Atlantic P-O3 for which
197 it is only about 2.5 times higher. The key photolysis rates are similar across all models except
198 GISS, and because of this and other inexplicable results, we drop these GISS R0 simulations in
199 further comparisons. For the 3 central models, the top 10% parcels by mass in the Global
200 statistics control 37-44% of the total reactivity for all 3 Rs; and the complementary result is that
201 the bottom 50% parcels control only 6-10% of the total reactivity. The dominance of the high
202 reactivity parcels is starkest for L-O3 and L-CH4 where 15% of the parcels (Atlantic or Pacific)
203 control half of the budget, while P-O3 is distributed more evenly across the parcels, especially in
204 the Atlantic.

205 These statistics are readily seen in the convex curves of sorted reactivity values in Figure 1. The
206 mean values (open circles plotted at the 50th percentile) lie well above the median, except for
207 Atlantic P-O3. The rapid, almost hyperbolic increases starting above the 90th percentile indicates
208 the importance of the rare, but most reactive, parcels to the total budgets. The exception to this
209 pattern is Atlantic L-O3, which looks more like two straight lines, breaking at the 50th percentile.
210 In Figure 1 the agreement across all models (except GISS) is clear, indicating that the conclusion
211 in Prather et al., 2018 (i.e., that most global chemistry models agree on the O₃ and CH₄ budgets if
212 given the chemical composition) also holds for the ATom measured chemical composition. For
213 P-O3, the F0AM model (red) has a lower median than the cluster of four 3D models, but the
214 curve steepens faster at the highest Rs. These differences are probably caused by the F0AM
215 protocol for NO_x (maintain fixed NO_x levels over 24 hours) versus the ATom-run protocol for
216 the 3D models (let NO_x photochemically evolve over 24 hours, see Fig. 2 of Prather et al., 2017),
217 but they do not greatly affect the conclusions here. The shift to observed NO_x with MDS_R2 –
218 compare UCI (brown, R0) with UCI2 (black dashed, R2) – had noticeable changes only in the
219 Pacific where P-O3 increased by 20% from the 10th to 80th percentile, but surprisingly, the
220 Atlantic P-O3 is reduced. It is clear that the reactive chemistry of the Atlantic is distinct from
221 that of the Pacific in ATom-1, but we could not simply identify the cause here. We conclude that
222 accurate modeling of chemical composition of 80th and greater percentile is important, but that
223 modest errors in the lowest 50th percentile are not consequential, effectively, some parcels matter
224 more, as in Prather et al. (2017).



225 How well does this ATom analysis work as a model intercomparison project? Overall, we find
226 that most models give similar results when presented with the ATom-1 MDS. The agreement
227 across a range of model formulations using differing levels of chemical complexity show this
228 approach is robust. The different protocol for calculating reactivities between the 3D and box
229 models, as well as the uncertainty in cloud fields, appear to have small impact on the overall
230 results, but are informative regarding the minimum structural uncertainty in calculating the
231 subsequent 24-hour reactivity of a well-measured air parcel.

232 3.1 Spatial heterogeneity of tropospheric reactivity.

233 A critical unknown for tropospheric chemistry modeling is what resolution is needed to correctly
234 integrate the reaction rates controlling the budgets of key gases? A similar question was
235 addressed in Yu et al. (2016) for the isoprene oxidation pathways using a model with variable
236 resolution (500 km, 250 km, and 30 km) compared to aircraft measurements. ATom's 10 s air
237 parcels measure 2 km (horizontal) by 80 m (vertical) during most profiles. These 10 s parcels
238 have similar variability to that seen on the 1 s scale (200 m by 8 m, see Figure S2). An
239 interesting parallel result from the RDS statistics is that there are not enough extremely rare
240 parcels with high enough reactivity to distort the average. While the sorted reactivity curves
241 (Figure 1, Tables 2 & S8) continue to steepen from the 90th to 97th percentile, the slope does not
242 change that much: the top 3% parcels contribute to about 8-15% of the total reactivity over the
243 Pacific and Atlantic Basins. Thus, if our modeling misses the top 1% of reactive air parcels, for
244 example due to the inability to simulate intensely reactive thin pollution layers, then we miss at
245 most 5% of the total reactivity. This finding is new and encouraging, and it needs to be verified
246 with the ATom-2, 3, and 4 data. The other question is whether the 3D models with 100 km by
247 500 m cells can even match the statistical distributions measured at 2 km by 80 m. There is some
248 encouraging information from the sample descent profile shown in Figure S2: the reactivities for
249 this profile change slowly over the 500-m intervals marked on the plot, and thus the models may
250 capture much of the variance.

251 The curtain plots of Figure 2 follow the profiling of ATom-1 in the Central Pacific, Eastern
252 Pacific, and Atlantic tropical domains. The UCI RDS_R2 are averaged and plotted in 1° latitude
253 by 200 m thick cells, and thus some of the small scale heterogeneity is lost here. We separate the
254 Eastern (121°W, research flight (RF) 1) from the Central Pacific (RF 3, 4 and 5) because we are
255 looking for contiguous latitude-by-pressure structures. In the Central Pacific (row 1), highly
256 reactive (hot) P-O3 parcels (> 6 ppb/day) occur in a partially resolved air mass at latitudes 20-
257 22°N and pressure altitudes 2-3 km, and in more scattered parcels (> 3 ppb/day) below 5 km
258 down to 20°S. High L-O3 and L-CH4 coincide with this 20-22°N air mass, and also with some
259 high P-O3 at lower latitudes. This pattern of overlapping extremes in all 3 Rs is surprising
260 because the models' mid-Pacific climatologies show a separation between regions of high L-O3
261 and high P-O3 (Figure 3 of Prather et al., 2017). In contrast to the pattern here, the modeled P-
262 O3 tends to peak in the upper troposphere and the marine boundary layer, reflecting perhaps the
263 models' parameterizations used to distribute lightning NOx. The ATom profiling seems to catch
264 similarly reactive parcels in adjacent profiles separate by a few hundred km, scales easily
265 resolvable with 3D models.

266 In the Eastern Pacific (row 2), the overlap of outbound and return profiles enhances the spatial
267 sampling over the 10 h flight. The region of very large L-O3 (> 5 ppb/day) is extensive,
268 beginning at 5-6 km at 10°N and broadening to 2-8 km at 28°N. The region of L-CH4 is similar,
269 but loss at higher altitudes is attenuated, because of the temperature dependence of L-CH4 and
270 possibly because of differing OH:HO₂ ratios with altitude. Large P-O3 (> 6 ppb/day) occurs in



271 some but not all of these highly reactive regions, suggesting that NO_x is not as evenly distributed
272 as is HO_x. P-O₃ also show regions of high reactivity above 8 km that are not in the high L-O₃
273 and L-CH₄ regions, probably evidence of convective sources of HO_x and NO_x, but too cold and
274 dry for the L-O₃ and L-CH₄ reactions. ATom-1 RF1 (29 Jul 2016) occurred during the North
275 American Monsoon when there was easterly flow off Mexico, thus the high reactivity of this
276 large air mass indicates that continental deep convection is a source of high reactivity for both O₃
277 and CH₄.

278 In the Atlantic (row 3) we see similar air masses through successive profiles again, particularly in
279 the northern tropics. The Atlantic P-O₃ shows high-altitude reactivity similar to the Eastern
280 Pacific. Likewise, the large values of L-O₃ and L-CH₄ match the Eastern Pacific and not Central
281 Pacific. Unlike either Pacific transect, the Atlantic L-O₃ and L-CH₄ show some high reactivity
282 below 1 km altitude. Overall, the ATom-1 profiling clearly identifies extended air masses of high
283 L-O₃ and L-CH₄ extending over 2-5 km in altitude and 10° of latitude. The high P-O₃ regions
284 tend to be much more heterogeneous with greatly reduced spatial extent.

285 Overall, the extensive ATom profiling identifies a highly heterogeneous mix of chemical
286 composition in the tropical Atlantic and Pacific with a large range of reactivities. What is
287 important for those trying to model tropospheric chemistry is that the spatial scales of variability
288 are within the capability of modern global models.

289 3.2 Testing model climatologies.

290 The ATom data set, including its derived RDS, provide a unique opportunity to test models, both
291 chemistry-transport (CTM) and chemistry-climate models (CCMs). As a first look, we make this
292 comparison using ATom-1 data and the modeled chemical statistics for mid-August used in
293 Prather et al. (2017). The complex patterns of the 3Rs seen in Figure 2 cannot be matched
294 directly with CCMs and thus we seek simpler statistics combining average profiles with
295 probability distributions. The mean profiles of the 3Rs from the Figure 2 tropical ocean are
296 compared with the 6 models in Figure 3.

297 For P-O₃ (top row), the discrepancy between models and measurements is stark. In the Central
298 Pacific at 8-12 km, the ATom-1 results tend to agree with models, showing ozone production of
299 about 1 ppb/day. Below 8 km, ATom's P-O₃ increases to a peak of 4 ppb/day at 2 km, while the
300 models' P-O₃ stay constant down to 4 km and then decrease to about 0.5 ppb/day below 2 km.
301 This pattern indicates that in the middle of the Pacific, the NO_x+HO_x combination that produces
302 ozone is suppressed throughout the lower troposphere in the models. In the Eastern Pacific and
303 Atlantic, both models' and ATom reactivities indicate that that P-O₃ is greatly enhanced above 6
304 km as compared to the Central Pacific, but below 6 km the ATom P-O₃ are much larger than that
305 of the models', by a factor of 2. In the upper troposphere, the agreement indicates that both
306 models and ATom find the influence of deep continental convection bringing reactive NO_x+HO_x
307 air masses to the nearby oceanic regions, but not to the mid Pacific. The difference below 5 km
308 in all 3 regions implies a consistent bias across the models in some combination of HO_x sources
309 and/or the vertical redistribution of lightning NO_x. This difference is unlikely to be a sampling
310 bias in ATom, given it occurs in all 3 regions.

311 For L-O₃ (middle row), the agreement in the Central Pacific is very good throughout the 0-12 km
312 range. Moving to the Eastern Pacific and Atlantic, both models and ATom show increased
313 reactivity consistent with continental convective outflow. The excess ATom reactivity in the
314 Eastern Pacific (3-8 km) is clear in Figure 2 and likely due to easterly mid-tropospheric flow from



315 convection over Mexico at that specific time (29 July 2016). Similarly, the ATom excess
316 reactivity at low-level (1-3 km) in the Atlantic is associated with biomass burning in Africa and
317 was measured in other trace species. Thus, in terms of L-O₃, the ATom-model differences may
318 be due to specific meteorological conditions and this could be tested with CTMs using 2016
319 meteorology and wildfires.

320 For L-CH₄ (bottom row), the ATom-model pattern is similar to L-O₃, but the excess ATom
321 reactivity occurs at lower altitudes. Overall, the ATom L-CH₄ is slightly greater than the
322 modeled L-CH₄. L-O₃ is dominated by O(1D) and HO₂ loss, while L-CH₄ is limited to OH loss.
323 Increases in NO_x shift the HO_x balance to OH, and we see that in the 1-2 km region of peak P-
324 O₃ in the Pacific, but less so in the Atlantic. Overall, there is clear evidence that the Atlantic and
325 Pacific have very different chemical mixtures controlling the reactivities and that convection over
326 land (monsoon or biomass burning) creates air masses that are still highly reactive a day or so
327 later.

328 Mean profiles do not reflect the heterogeneity seen in Figure 2, and so we also examine the
329 statistical distribution of reactivities underlying the profiles in Figure 3, as shown in supplemental
330 Figure S7. The frequency of occurrence uses the same weighting as before with 0.5 ppb/d bins
331 (0.25 ppb/d for L-CH₄). The model climatologies (6 colors) show a monotonically decreasing
332 frequency that mostly matches the ATom statistics. The obvious discrepancy is with P-O₃ in
333 both Pacific and Atlantic basins. ATom data has very low occurrence of P-O₃ < 1 ppb/d and a
334 broad, almost uniform frequency (~0.1) extending out to 4 ppb/d. This result is consistent with
335 the mean profile errors (Figure 3) and emphasizes where the error is occurring. The match for L-
336 CH₄ is excellent in both basins, although the models have a greater occurrence in the 0.5-1.5
337 ppb/d middle range, and reduced occurrence in the higher 1.5-2.5 ppb/d range. For L-O₃, the
338 match is very good and similar, with models overestimating the middle range (1-3 ppb/d) and
339 missing the upper range (>5 ppb/d). The frequency of reactivities greater than the plot range is
340 included in the last point, and for L-O₃ the ATom parcels have a large occurrence of L-O₃ > 6
341 ppb/d that is not seen in the models (except GISS). Similar results apply to P-O₃. For the
342 Pacific, both Central and Eastern have been combined (black line), but a separate line (grey)
343 including only the Central Pacific is also shown. It is clear that the extreme reactivities are in the
344 Eastern Pacific, as is evident from Figure 2.

345 The ability to test the model's reactivity statistics with the ATom 10 s data is not obvious. The
346 disconnect between statistics for 2-km parcels versus 100-km grid cells is likely. If there is
347 significant chemical variability at scales below 50 km, then the averaging occurring in model grid
348 cells would greatly alter the frequencies. For example, in Figure S8, we take the Pacific P-O₃
349 frequency (black), generate a random set of points from that distribution, and then start averaging
350 adjacent points in groups (2, 4, 8; denoted as 4 km, 8km, 16 km). The resulting statistics rapidly
351 evolve into a Gaussian-like distribution about the mean value. Thus, the ability to nearly match
352 the ATom-1 statistics with our global chemistry models is significant, and we cannot explain the
353 P-O₃ discrepancy as a model-averaging problem.

354 This deficit in modeled P-O₃ points to a NO_x deficiency in the models, and this is clear in the
355 comparison of the probability distribution histograms for NO_x shown in Figure 4. In the Central
356 Pacific over 0-12 km (1st row), ATom has a reduced frequency of parcels with 2-20 ppt and
357 corresponding increase in parcels with 20-80 ppt. This discrepancy is amplified in the lower
358 troposphere, 0-4 km (2nd row). In the middle of the Pacific our chemistry models are missing a
359 large, local source of NO_x. The obvious source is lightning since oceanic sources of organo-
360 nitrates as measured on ATom could not supply this amount. The distribution of NO_x throughout



361 the troposphere also indicates a lightning-related source that is vertically mixed. In the Eastern
362 Pacific, the ATom 0-4 km troposphere appears again to have large amounts of air with 20-50 ppt,
363 while the full troposphere more closely matches the models except for the large occurrence of air
364 with 100-300 ppt NO_x. These high-NO_x regions are probably direct outflow from very deep
365 convection with lightning in the monsoon regions over Mexico at this time. In the Atlantic, the
366 modeled NO_x shows too frequent occurrence of low NO_x (<10 ppt) and thus underestimates the
367 10-100 ppt levels at all altitudes. ATom has a strong peak occurrence about 80-120 ppt in the
368 upper troposphere and, like the East Pacific, this is probably due to lightning NO_x from deep
369 convection over land (Africa or South America). The models appear to be missing significant
370 NO_x sources throughout the tropics, especially below 4 km.

371 In Figure 4 we also look at the histograms for the key HO_x-related species HOOH (3rd row) and
372 HCHO (4th row). For these species, the ATom-model agreement is generally good. If anything,
373 the models tend to have too much HOOH: ATom shows systematically large occurrences of low
374 HOOH (50-200 ppt, especially Central Pacific) indicating, perhaps, that convective or cloud
375 scavenging of HOOH is more effective than is modeled. HCHO shows reasonable agreement in
376 the Atlantic, but in both Central and Eastern Pacific, the modeled low-end (<40 ppt) is simply not
377 seen in the ATom data. Also, the models are missing a strong HCHO peak at 300 ppt in the
378 Eastern Pacific. Thus, in terms of these HO_x precursors, the model climatologies appear to be at
379 least as reactive as the ATom data.

380 While the ATom-1 data in Figure 4 is limited to a single transect, the model discrepancies apply
381 across the three tropical regions, and the simple chemical statistics for these flights are probably
382 enough to identify measurement-model discrepancies in a climatological sense and focusing on
383 specific abundance ranges. For the individual models, a more thorough comparison with these
384 and other key species (e.g., CO, O₃, H₂O, CH₃OOH, PAN, C₂H₆) is likely to identify areas for
385 model development. An encouraging result here is that where individual models tend to deviate
386 from their peers, they also deviate from the ATom-1 probability distributions. In most cases (but
387 not NO_x), the models match the first-order statistics from ATom.

388 Some of the structure in these chemical distributions is driven by large-scale physics, such as the
389 cold and dry upper troposphere having low HOOH and HCHO levels and reactivities < 1 ppb/d.
390 Nevertheless, however one parses these statistical comparisons in Figures 4 and S7, it appears
391 that current global chemistry models with resolutions of about 100 km by 500 m can resolve
392 much of the wide range of chemical heterogeneity in the atmosphere, which for the oceanic
393 transects is, we believe, adequately resolved by the 10 s ATom measurements.

394

395 **4. Conclusion**

396

397 This paper opens the door for what can be done by the community with the publicly available
398 ATom data. The mix of key species that allows us to calculate the reactivity of the air parcels is
399 near unique. We find that the reactivity of the troposphere with respect to O₃ and CH₄ is
400 dominated by a fraction of the air parcels, but not by so small and infrequent a fraction as to
401 challenge the resolution of current CTMs to simulate these observations. In comparing ATom
402 results with modeled climatologies, we find a clear model error, missing O₃ production over the
403 tropical oceans' lower troposphere, and traced it to the lack of NO_x below 4 km. The occurrence
404 of the same error over the Central and Eastern Pacific as well as the Atlantic Ocean makes this a
405 robust model-measurement discrepancy.



406 Building our chemical statistics from the ATom 10 s air parcels on a scale of 2 km by 80 m, we
407 can identify the fundamental scales of spatial heterogeneity in tropospheric chemistry. Although
408 heterogeneity occurs at the finest scales (such as seen in some 1 s observations) the majority of
409 variability in terms of the O₃ and CH₄ budgets occurs across scales larger than neighboring 2 km
410 parcels. It appears to be largely resolved by the models' 100 km by 500 m grid cells. This
411 surprising result is evident by the similarity of many of the ATom 1D probability densities – both
412 species and reactivities – with those from the models' August climatologies. These comparisons
413 show that the modeled chemical heterogeneity is consistent with the innate heterogeneity of the
414 troposphere as measured by the 10 s parcels in ATom. A similar conclusion for biomass burning
415 smoke particles is found by Schill et al (2020), where the most of the smoke appears in the
416 background rather than in pollution plumes, and therefore much of the variability occurs on
417 synoptic scales resolved by global models (see their Figure 1 compared with Figure 2 here).

418 As a quick look at the opportunities provided by the ATom data, we present an example based on
419 Wolfe et al (2019), who used the F0AM model and semi-analytical arguments to show that
420 troposphere HCHO columns (measurable by satellite and ATom) are related to OH columns
421 (measured by ATom) and thus to CH₄ loss. Figure 5 extends the Wolfe et al study using the
422 individual air parcels and plotting L-CH₄ (ppb/d) versus HCHO (ppt) for the 3 tropical regions
423 where most of the CH₄ loss occurs. The relationship is linear with slopes ranging from 4 to 6 /d,
424 but the largest reactivities (0-4 km, 1-3 ppb/d) are not so well correlated with HCHO.

425 Decadal scale shifts in the budgets of O₃ and CH₄ are likely to be evident through the statistical
426 patterns of the key species, rather than simply via average profiles. The underlying design of
427 ATom was to collect enough data to develop such a multivariate chemical climatology. As a
428 quick look, we show the joint probability distribution on a logarithmic scale as in Prather et al.
429 (2017) for HOOH and NO_x in Supplemental Figure S6. The patterns for the tropical Central
430 Pacific are quite similar for the 4 seasons of ATom deployments and the fitted ellipses are almost
431 identical for ATom 2, 3, and 4. Thus, for these species in the Central Pacific, we believe that
432 ATom provides a benchmark of the 2016-2018 chemical state, one that can be revisited with an
433 aircraft mission in a decade to detect changes in not only chemical composition but also
434 reactivity.

435 ATom identifies which 'highly reactive' spatial or chemical environments could be targeted in
436 future campaigns for process studies or to provide a better link between satellite observations and
437 photochemical reactivity. The many corollary species measured by ATom (not directly involved
438 in CH₄ and O₃ chemistry) provide clues to the origin or chemical processing of these
439 environments.

440 As is usual with a new model intercomparison projects, we find new model features and identify
441 errors. In the UCI model a lumped alkane formulation (averaging alkanes C₃H₈ and higher) did
442 not show up in Prather et al. (2018) where UCI supplied all the species, but when the ATom data
443 were used, the UCI model became an outlier. Once found, this problem was easily fixed. The
444 anomalous GISS results have been examined by a co-author, but no clear causes have been
445 identified as of this publication. Inclusion of the F0AM model here for the reactivity calculations
446 provided an interesting contrast with the A-run simulations of the global CCM/CTMs as
447 described in Prather et al. (2017). We hope to engage a wider modeling community beyond the
448 ATom science team, as in Hall et al. (2018), in the calculation of photochemical processes and
449 budgets based on the ATom measurements.



450 **Data Availability.** The MDS and RDS data for ATom 1, 2, 3, and 4 are presented here as core
451 ATom deliverables, and are now posted on the NASA ESPO ATom website
452 (<https://espo.nasa.gov/atom/content/ATom>) and will be archived at the ORNL DAAC alongside
453 the 10 s ATom merged measurements. Details of the ATom mission and data sets are found on
454 the NASA mission web site (<https://espo.nasa.gov/atom/content/ATom>) and at the final archive at
455 Oak Ridge National Laboratory (ORNL;
456 https://daac.ornl.gov/ATOM/guides/ATom_merge.html). The Matlab codes and data sets used in
457 the analysis here are posted on Dryad (<https://doi.org/10.7280/D1Q699>).

458 **Author Contributions:**

459 H.G., C.M.F, S.C.W., and M.J.P. designed the research and performed the data analysis; S. A. S.,
460 S. D. S., L. E., F. L., J. L., A. M. F., G. C., L. T. M., and G. W., contributed original atmospheric
461 chemistry model results; G. W., M. K., J. C., G. D., J. D., B. C. D., R. C., K. M., J. P., T. B. R.,
462 C. T., T. F. H., D. B., N. J. B., E. C. A., R. S. H., J. E., E. H., and F. M. contributed original
463 atmospheric observations; H.G., C.M.F, and M.J.P. wrote the paper.

464 **Competing Interest Statement:** No.

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466
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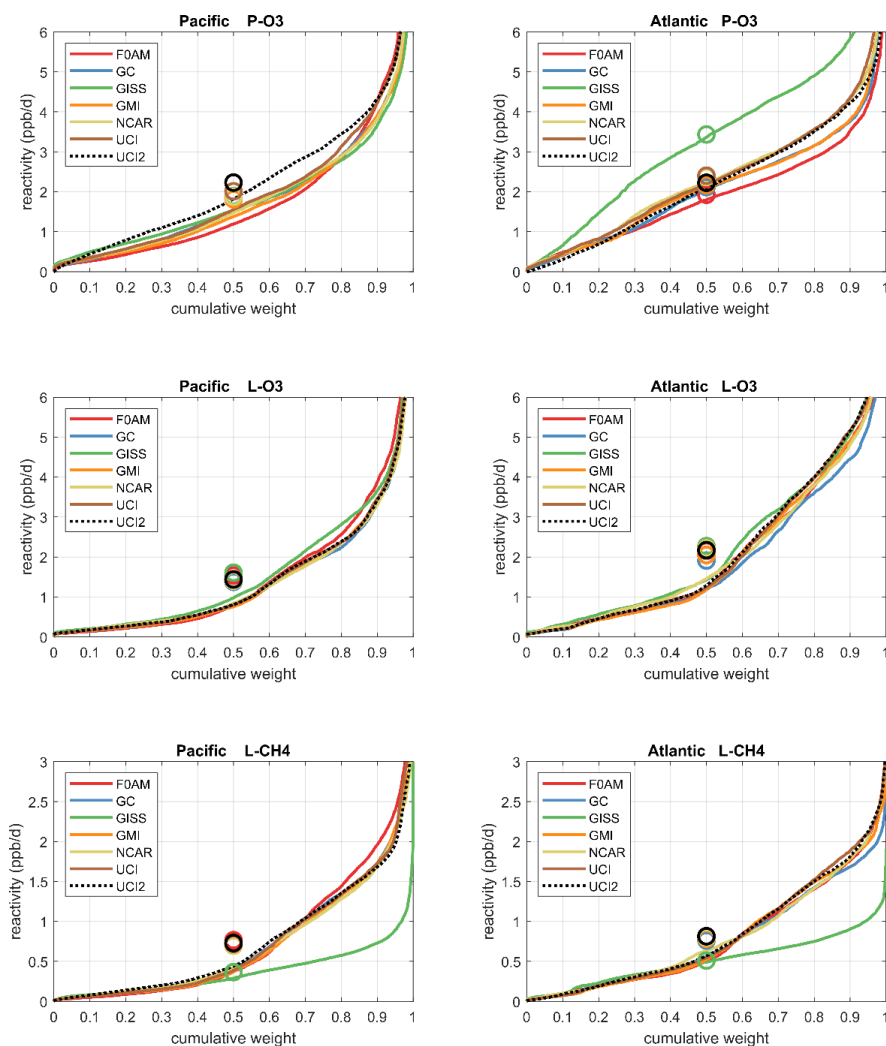
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575 **Figures and Tables**

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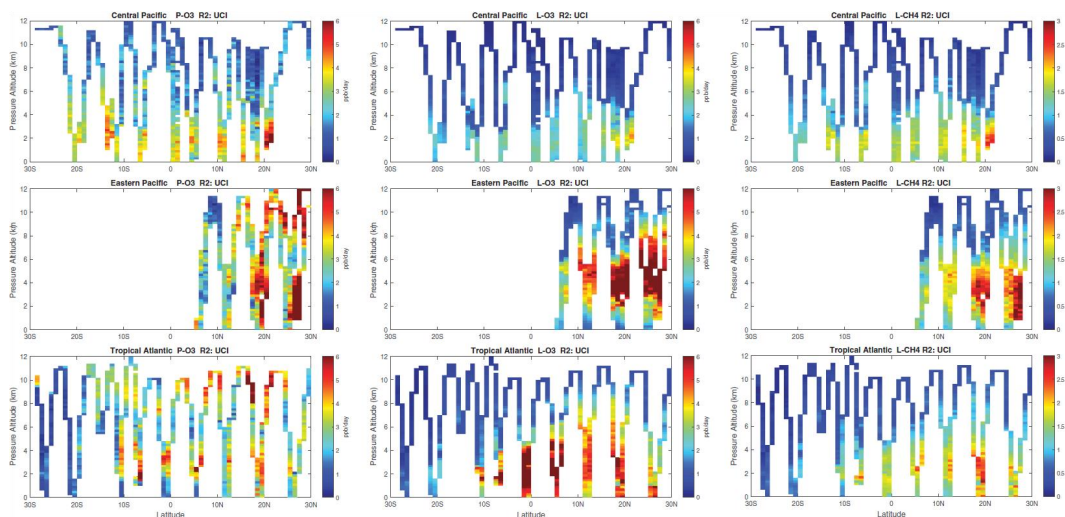
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580 **Figure 1.** Sorted reactivities (P-O3, L-O3, L-CH4, ppb/day) for the Pacific and Atlantic domains
581 of ATom-1. Each parcel is weighted, see text. The six modeled reactivities for MDS_R0 are
582 shown with colored lines and the UCI calculation for MDS_R2 is shown as a black dashed line.
583 The mean value for each model is shown with an open circle plotted at the 50th percentile. (Flip
584 about the axes and it is a cumulative probability density function.)



585

586 **Figure 2.** Curtain plots for P-O3 (0-6 ppb/d), L-O3 (0-6 ppb/d) and L-CH4 (0-3 ppb/d) showing
587 the profiling of ATom-1 flights in the central Pacific (RF 3, 4 and 5), eastern Pacific (RF 1), and
588 Atlantic (RF 7, 8, and 9). Reactivities are calculated for MDS_R2 with the UCI CTM. The 10 s
589 air parcels are averaged into 1° latitude and 200 m bins.

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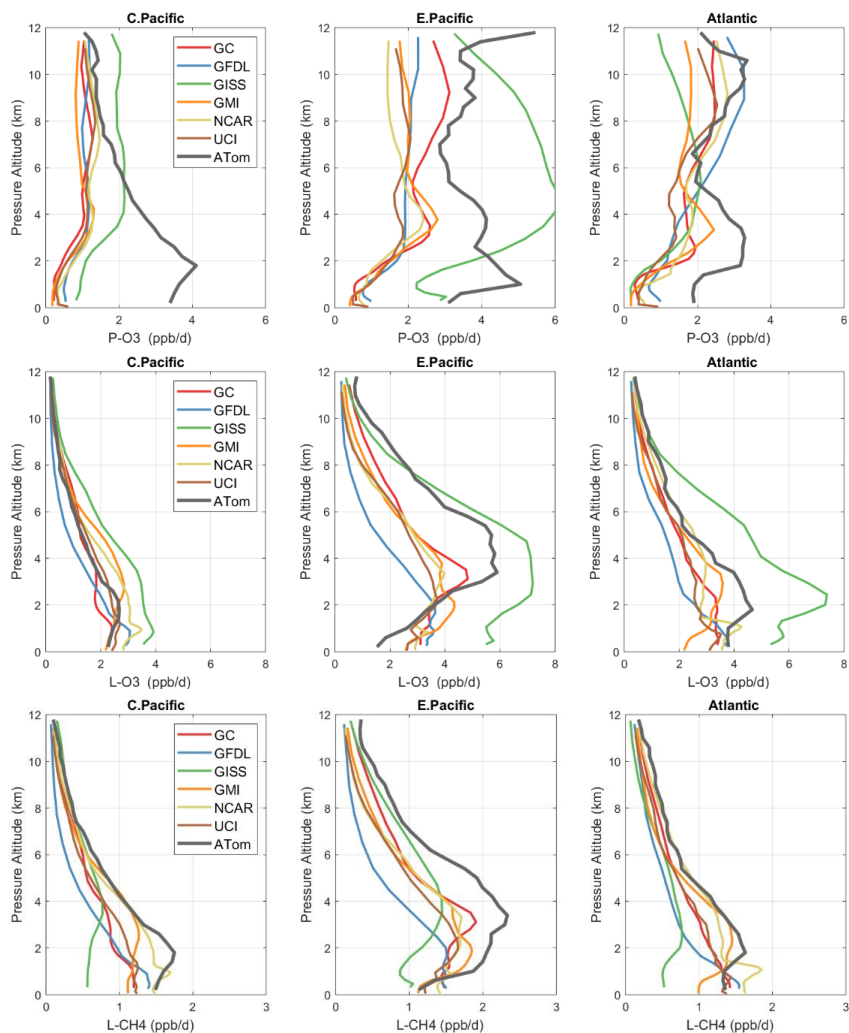
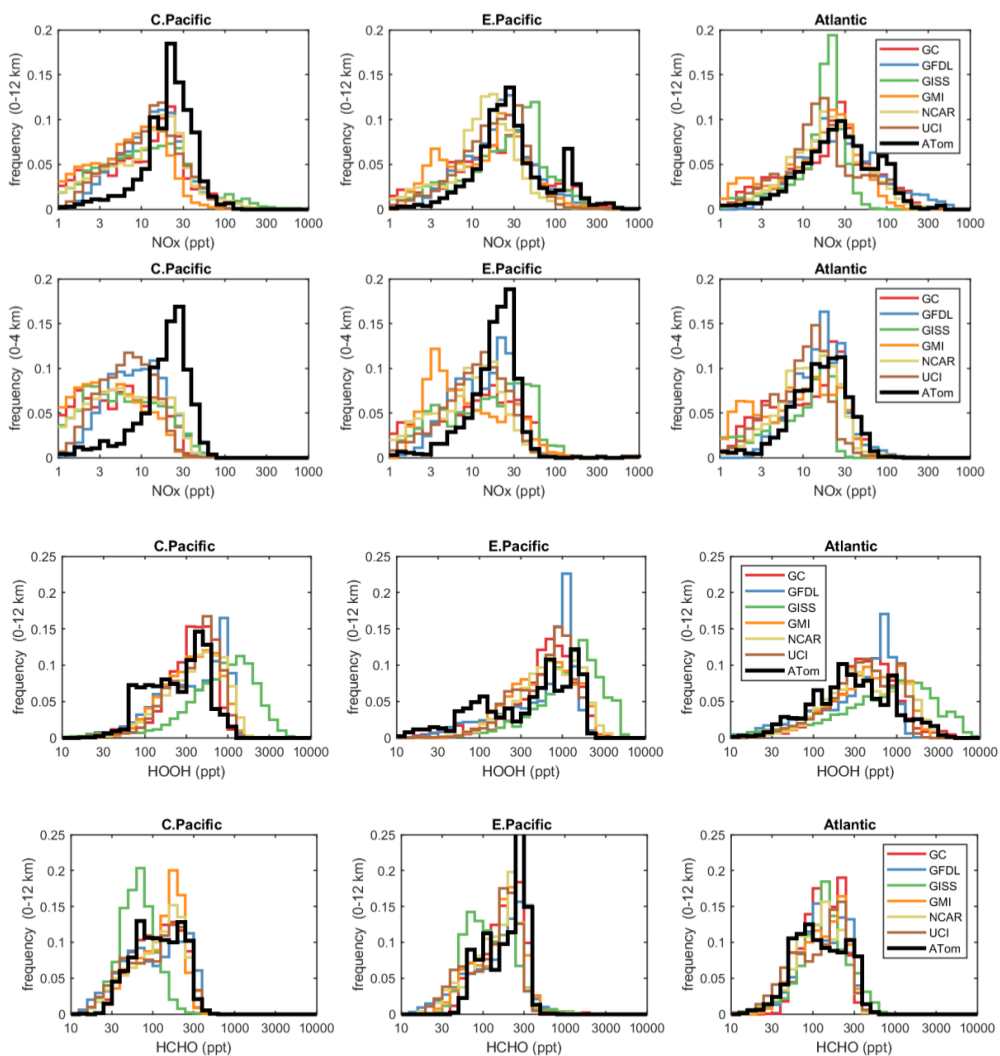


Figure 3. Mean profiles of reactivity (rows: P-O₃, L-O₃, L-CH₄ in ppb/day) in 3 domains (columns: C. Pacific, 30°S-30°N by 180-210°E; E. Pacific, 0-30°N by 230-250°E; Atlantic, 30°S-30°N by 326-343°E). ATom-1 (gray) results are from **Figure 2**, while model results are taken from the August climatologies in Prather et al. (2017).

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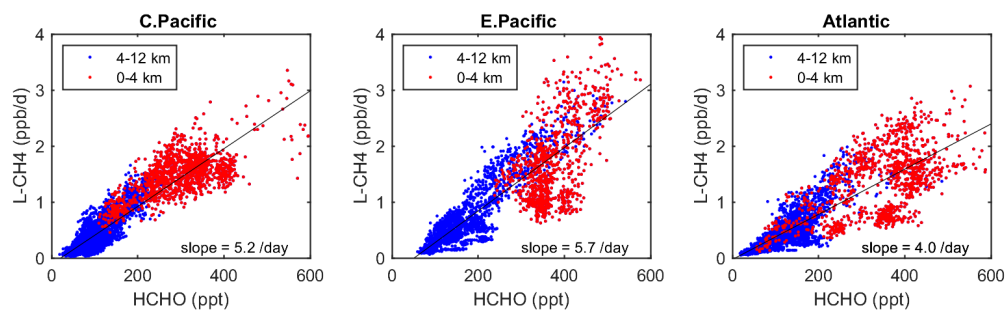
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598 **Figure 4.** Histograms of probability distributions (PDs) of NO_x (0-12 km, row 1), NO_x (0-4 km,
599 row 2), HOOH (0-12 km, row 3), and HCHO (0-12 km, row 4) for the 3 tropical regions (Central
600 Pacific, Eastern Pacific, Atlantic). All species are binned in ppt, uniform on a log scale. The
601 weighting of air parcels is described in the text. The ATom-1 data is plotted on top of the 6
602 global chemistry models' results for a day in mid-August and sampled for the regions used in
603 **Figure 3.**

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608 **Figure 5.** Scatterplot of L-CH₄ (ppb/d) versus HCHO (ppt) for ATom 1 in the 3 tropical regions
609 shown in **Figure 3**. The air parcels are split into lower troposphere (0-4 km pressure altitude, red
610 dots) where most of the reactivity lies and mid+upper troposphere (4-12 km, blue). A simple
611 linear fit to all data is shown (thin black line) and the slope is given in units of 1/day.

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Table 1. Chemistry models

Used in	ID	Model name	Model type	Meteorology	Model Grid
clim, R0	GFDL	GFDL-AM3	CCM	NCEP (nudged)	C180 x L48
clim, R0	GISS	GISS-E2.1	CCM	Daily SSTs, nudged to MERRA	2° x 2.5° x 40L
clim, R0, R1	GMI	GMI-CTM	CTM	MERRA	1° x 1.25° x 72L
clim, R0	GC	GEOS-Chem	CTM	MERRA-2	2° x 2.5° x 72L
clim, R0	NCAR	CAM4-Chem	CCM	MERRA	0.47° x 0.625° x 52L
clim, R0, R1, R2	UCI	UCI-CTM	CTM	ECMWF IFS Cy38r1	T159N80 x L60
R0	F0AM	F0AM	box	MDS+scaled ATom Js	N/A

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Table 2. Reactivity statistics for the 3 large domains (Global, Pacific, Atlantic).

Value	Region	Models with R0							Models with R2		
		F0AM	GC	GISS	GMI	NCAR	UCI	U15	U97	GMI1	UCI2
P-O ₃ , mean, ppb/d	Global	1.83	1.58	1.98	1.53	1.64	1.75	1.75	1.75	1.75	1.83
	Pacific	1.96	1.97	2.00	1.92	1.99	2.13	2.09	2.10	2.19	2.35
	Atlantic	2.11	2.29	3.73	2.38	2.57	2.61	2.60	2.61	2.74	2.63
L-O ₃ , mean, ppb/d	Global	1.55	1.17	1.39	1.22	1.24	1.27	1.27	1.27	1.23	1.29
	Pacific	1.62	1.46	1.69	1.49	1.52	1.53	1.49	1.51	1.48	1.53
	Atlantic	2.32	2.17	2.57	2.36	2.43	2.48	2.47	2.49	2.39	2.51
L-CH ₄ , mean, ppb/d	Global	0.68	0.53	0.32	0.52	0.51	0.55	0.55	0.55	0.55	0.56
	Pacific	0.81	0.76	0.39	0.74	0.75	0.77	0.75	0.76	0.77	0.78
	Atlantic	0.90	0.88	0.57	0.92	0.90	0.95	0.95	0.95	0.96	0.97
P-O ₃ , %of total R in top 10%	Global	38%	38%	36%	38%	37%	40%	40%	40%	38%	37%
	Pacific	35%	28%	27%	29%	28%	30%	29%	29%	28%	26%
	Atlantic	22%	22%	20%	22%	21%	23%	23%	23%	21%	21%
L-O ₃ , %of total R in top 10%	Global	39%	44%	41%	44%	43%	45%	45%	45%	45%	44%
	Pacific	36%	34%	32%	35%	33%	34%	34%	34%	36%	34%
	Atlantic	26%	25%	25%	25%	28%	26%	26%	26%	25%	26%
L-CH ₄ , %of total R in top 10%	Global	37%	39%	33%	40%	39%	41%	41%	41%	41%	39%
	Pacific	35%	30%	26%	32%	30%	31%	31%	31%	31%	29%
	Atlantic	25%	23%	19%	23%	24%	24%	24%	24%	24%	24%

Global includes all AToM-1 parcels (32,383), Pacific considers all measurements (11,486) over the Pacific Ocean from 54°S to 60°N, and Atlantic uses parcels from 54°S to 60°N over the Atlantic basin (7,501). All parcels are weighted at described in the text, including a cosine(latitude) factor. See expanded Table S1.

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Table 3. Cross-model RMS differences (% of mean) for reactivities

	F0AM	GC	GISS	GMI	NCAR	UCI
P-O3						
F0AM		48%	95%	45%	55%	42%
GC	48%		78%	26%	42%	32%
GISS	95%	78%		81%	72%	74%
GMI	45%	26%	81%		40%	35%
NCAR	55%	42%	72%	40%		42%
UCI	42%	32%	74%	35%	42%	(10%)
L-O3						
F0AM		40%	44%	43%	76%	38%
GC	40%		33%	25%	60%	24%
GISS	44%	33%		36%	66%	30%
GMI	43%	25%	36%		62%	28%
NCAR	76%	60%	66%	62%		60%
UCI	38%	24%	30%	28%	60%	(11%)
L-CH4						
F0AM		47%	136%	48%	82%	45%
GC	47%		111%	20%	60%	27%
GISS	136%	111%		114%	110%	121%
GMI	48%	20%	114%		57%	30%
NCAR	82%	60%	110%	57%		68%
UCI	45%	27%	121%	30%	68%	(13%)

Calculated with the 31,376 MDS_R0 unweighted parcels that are a subset of the 32,383 MDS_R2 parcels. F0AM lacks 5,510 parcels because of the lack of observed J-values. UCI shows RMSD between years 2016 (default) and 1997 as the value in parentheses on diagonal. The unweighted mean from 3 core models (GC, GMI, UCI) are: P-O3 = 1.90, L-O3 = 1.24, L-CH4 = 0.51, all ppb/day.

619
 620