

1 **Heterogeneity and chemical reactivity of the remote troposphere defined by aircraft**
2 **measurements - Corrected**

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43 ATom

44 **Abstract.** The NASA Atmospheric Tomography (ATom) mission built a photochemical
45 climatology of air parcels based on in situ measurements with the NASA DC-8 aircraft
46 along objectively planned profiling transects through the middle of the Pacific and
47 Atlantic oceans. In this paper we present and analyze a data set of 10 s (2 km) merged
48 and gap-filled observations of the key reactive species driving the chemical budgets of O₃
49 and CH₄ (O₃, CH₄, CO, H₂O, HCHO, H₂O₂, CH₃OOH, C₂H₆, higher alkanes, alkenes,
50 aromatics, NO_x, HNO₃, HNO₄, peroxyacetyl nitrate, other organic nitrates), consisting of
51 146,494 distinct air parcels from ATom deployments 1 through 4. Six models calculated
52 the O₃ and CH₄ photochemical tendencies from this modeling data stream for ATom 1.
53 We find that 80 % – 90 % of the total reactivity lies in the top 50 % of the parcels; and 25
54 % – 35 %, in the top 10 %, supporting previous model-only studies that tropospheric
55 chemistry is driven by a fraction of all the air. Surprisingly, the probability densities of
56 species and reactivities averaged on a model scale (100 km) differ only slightly from the
57 2 km ATom 10 s data, indicating that much of the heterogeneity in tropospheric
58 chemistry can be captured with current global chemistry models. Comparing the ATom
59 reactivities over the tropical oceans with climatological statistics from six global
60 chemistry models, we find generally good agreement with the reactivity rates for O₃ and
61 CH₄. Models distinctly underestimate O₃ production below 2 km relative to the mid-
62 troposphere, and this can be traced to lower NO_x levels than observed. Attaching
63 photochemical reactivities to measurements of chemical species allows for a richer, yet
64 more constrained-to-what-matters, set of metrics for model evaluation. This paper
65 presents a corrected version of the paper published under the same authors and title (sans
66 'Corrected') as <https://doi.org/10.5194/acp-21-13729-2021>.

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68
69 **Preface.** While continuing our analysis of the ATom data we found several major
70 mistakes or decision errors. The main conclusions were unchanged except those
71 regarding production of O₃, but most of the numbers and many of the figures changed
72 slightly. A corrigendum to the original 2021 paper was prepared, but the changes were
73 extensive enough so that the ACP editors and the authors decided that a completely new
74 paper should be produced and the 2021 paper withdrawn. The errors that were corrected
75 are described in this preface and discussed at most briefly in the paper. First, we found
76 that measurement errors in PAN and HNO₄ were large (~100 ppt), and when this
77 occurred in the lower troposphere, the rapid thermal decomposition released large
78 amounts of NO_x. There is no easy fix for this, and we developed a new protocol (RDS*)
79 for computing reactivities by allowing the species to thermally decompose before use in
80 the model, as described below. This fix greatly reduced O₃ production (P-O3) in the
81 lower troposphere. A second NO_x problem involved the propagation of polluted profiles
82 from the Los Angeles basin to gap filling over the tropical eastern Pacific. This
83 correction resulted in the update of the Modeling Data Stream to version 2b. These NO_x
84 errors cause noticeable changes in reactivities, especially P-O3. Other decision errors led
85 us to decrease the southern latitude extent of the Atlantic and Pacific transects from 54° S
86 to 53° S to avoid spurious parcels being included. Also, cosine of latitude weighting was
87 applied to data for all figures and tables. The UCI model now includes all higher alkanes
88 and alkenes in the ATom data as C₃H₈ and C₂H₄, respectively. These last three decision
89 errors had detectable but small impacts.

90
 91 The most worrisome error was the evolution of the UCI CTM model's ATom version
 92 from its use in the MDS-0 results shown here to the final calculations with MDS-2 as the
 93 UCI2* model in the 2021 paper. The first MDS-0 UCI model was taken directly from
 94 the main CTM code line and developed for Prather et al. (2017; 2018) by Xin Zhu (not on
 95 the 2021 paper). This model was then further adapted and developed for the 2021 paper
 96 and for additional complex sensitivity tests. At this stage (i.e., the UCI2* simulations in
 97 the 2021 paper), the results failed several logic tests and were irreproducible. With the
 98 decision to withdraw the paper, we returned to the MDS-0 UCI model, and Xin Zhu
 99 adapted it to more efficient ATom runs as well as adding several new diagnostics and
 100 checks to ascertain the ATom runs were being calculated correctly. As noted in the paper
 101 below, we carefully checked the O₃ budget in terms of rates and tendencies, and these are
 102 now consistent in model UCIZ. Further, the sensitivity coefficients ($\partial \ln R / \partial \ln X$ and
 103 $\partial^2 \ln R / \partial \ln X \partial \ln Y$) calculated for a subsequent paper are now closer to theoretical
 104 expectations for a quasi-linear system. The UCIZ* model results here are our best,
 105 revised estimate of the ATom reactivities.

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108 **1 Prologue**

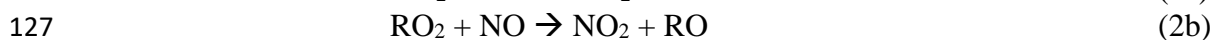
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110 This paper is based on the methods and results of papers that established an approach for
 111 analyzing aircraft measurements, specifically the NASA Atmospheric Tomography
 112 Mission (ATom), with global chemistry models. Here we present a brief overview of
 113 those papers to help the reader understand the basis for this paper. The first ATom
 114 modeling paper ("Global atmospheric chemistry – which air matters", Prather et al., 2017,
 115 hence P2017) gathered six global models, both chemistry-transport models (CTMs) and
 116 chemistry-climate models (CCMs). The models reported a single-day snapshot for mid-
 117 August (the time of the first ATom deployment, ATom-1), and these included all species
 118 relevant for tropospheric chemistry and the 24 h reactivities. We limited our study to
 119 three reactivities (Rs) controlling methane (CH₄) and tropospheric ozone (O₃) using
 120 specific reaction rates to define the loss of CH₄ and the production and loss of O₃ in parts
 121 per billion (ppb) per day. The critical photolysis rates (*J* values) were also reported as 24
 122 h averages.

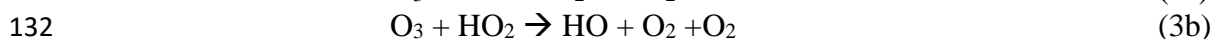
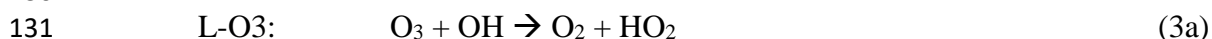
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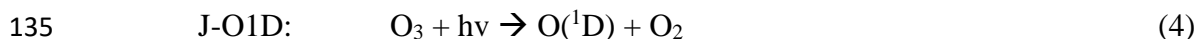
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137 J-NO2: $\text{NO}_2 + h\nu \rightarrow \text{NO} + \text{O}$ (5)

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139

140 Models also reported the change in O_3 over 24 h, and these match the P-O3 minus L-O3
141 values over the Pacific basin (a focus of this study). The models showed a wide range in
142 the three Rs average profiles across latitudes over the Pacific basin, as well as 2D
143 probability densities (PDs) for key species such as NO_x ($\text{NO} + \text{NO}_2$) versus HOOH. A
144 large part of the model differences was attributed to the large differences found in
145 chemical composition rather than the calculation of rates from that composition. We
146 found that single transects from a model through the tropical Pacific at different
147 longitudes produced nearly identical 2D PDs, but these PDs were distinctly different
148 across models. This result supported the premise that the ATom PDs would provide a
149 useful metric for global chemistry models.

150

151 In P2017, we established a method for running the chemistry modules in the CTMs and
152 CCMs with an imposed chemical composition from aircraft data: the ATom run, or “A
153 run”. In the A run, the chemistry of each grid cell does not interact with its neighbors or
154 with externally imposed emission sources. Effectively the CTM/CCM is initialized and
155 run for 24 h without transport, scavenging or emissions. Aerosol chemistry is also turned
156 off in the A runs. This method allows each parcel to evolve in response to the daily cycle
157 of photolysis in each model and be assigned a 24 h integrated reactivity. The
158 instantaneous reaction rates at the time an air parcel is measured (e.g., near sunset at the
159 end of a flight) do not reflect that parcel's overall contribution to the CH_4 or O_3 budget; a
160 full diel cycle is needed. The A run assumption that parcels do not mix with neighboring
161 air masses is an approximation, and thus for each model we compared the A runs using
162 the model's restart data with a parallel standard 24 h simulation (including transport,
163 scavenging, and emissions). Because the standard grid-cell air moves and mixes, we
164 compared averages over a large region (e.g., tropical Pacific). We find some average
165 biases of order $\pm 10\%$ but general agreement. The largest systematic biases in the A runs
166 are caused by buildup of HOOH (no scavenging) and decay of NO_x (no sources). The A
167 runs are relatively easy to code for most CTM/CCMs and allow each model's chemistry
168 module, including photolysis package, to run normally. The A runs do not distinguish
169 between CTMs and CCMs, except that each model will generate/prescribe its own cloud
170 fields and photolysis rates. Our goal is to create a robust understanding of the chemical
171 statistics including the reactivities with which to test and evaluate the free-running
172 CCMs, and thus we do not try to model the specific period of the ATom deployments.
173 Others may use the ATom data with hindcast CTMs to test forecast models, but here we
174 want to build a chemical climatology.

175

176 The first hard test of the A runs came with the second ATom modeling paper ("How well
177 can global chemistry models calculate the reactivity of short-lived greenhouse gases in
178 the remote troposphere, knowing the chemical composition", Prather et al., 2018, hence
179 P2018). The UCI CTM simulated an aircraft-like data set of 14,880 air parcels along the
180 International Date Line from a separate high-resolution (0.5°) model. Each parcel is
181 defined by the following core species: H_2O , O_3 , NO_x , HNO_3 , HNO_4 , PAN (peroxyacetyl

182 nitrate), CH_3NO_3 , HOOH , CH_3OOH , HCHO , CH_3CHO (acetaldehyde), $\text{C}_3\text{H}_6\text{O}$ (acetone),
183 CO , CH_4 , C_2H_6 , alkanes (C_3H_8 and higher), C_2H_4 , aromatics (benzene, toluene, xylene)
184 and C_5H_8 (isoprene), plus temperature. Short-lived radicals (e.g., OH , HO_2 , CH_3OO)
185 were initialized at small concentrations and quickly reached daytime values determined
186 by the core species. The six CTM/CCMs overwrote the chemical composition of a restart
187 file, placing each pseudo-observation in a unique grid cell according to its latitude,
188 longitude, and pressure. If another parcel is already in that cell, then it is shifted east–
189 west or north–south to a neighboring model cell. For coarse-resolution models, multiple
190 restart files and A runs were used to avoid large location shifts. CTM/CCMs usually
191 have a locked in 24 h integration step starting at 0000 UTC that is extremely difficult to
192 modify in order to try to match the local solar time of observation, especially as it
193 changes along aircraft flights. We tested the results with a recoded UCI CTM to start at
194 1200 UTC but retain the same clouds fields over the day and found only percentage-level
195 differences between a midnight or noon start.

196

197 These A runs averaged over cloud conditions by simulating 5 d in August at least 5 d
198 apart. Assessment of the modeled photolysis rates and comparison with the ATom-
199 measured J values is presented in Hall et al. (2018, hence H2018). All models agreed
200 that a small fraction of chemically hot air parcels in the synthetic data set controlled most
201 of the total reactivity. Some models had difficulty in implementing the A runs because
202 they overwrote the specified water vapor with the modeled value, but this problem is
203 fixed here. In both P2017 and P2018, the GISS-E2 model stood out with the most
204 unusual chemistry patterns and sometimes illogical correlations. Efforts by a co-author
205 to clarify the GISS results or identify errors in the implementation have not been
206 successful. GISS results are included here for completeness in the set of three papers but
207 are not reconciled. Overall, three models showed remarkable inter-model agreement in
208 the three Rs with less than half of the RMSD (root-mean-square difference) as compared
209 with the other models. UCI also tested the effect of different model years (1997 and
210 2015 versus reference year 2016), which varies the cloud cover and photolysis rates, and
211 found an inter-year RMSD about half of that of the core model’s RMSD. Thus, there is a
212 fundamental uncertainty in this approach due to the inability to specify the
213 cloud/photolysis history seen by a parcel over 24 h, but it is less than the inter-model
214 differences among the most similar models.

215

216 **2 Introduction**

217

218 The NASA Atmospheric Tomography (ATom) mission completed a four-season
219 deployment, each deployment flying from the Arctic to Antarctic and back, traveling
220 south through the middle of the Pacific Ocean, across the Southern Ocean and then north
221 through the Atlantic Ocean, with near-constant profiling of the marine troposphere from
222 0.2 to 12 km altitude (see Fig. S1). The DC8 was equipped with in situ instruments that
223 documented the chemical composition and conditions at time intervals ranging from <1
224 to about 100 seconds (Wofsy et al., 2018). ATom measured hundreds of gases and
225 aerosols, providing information on the chemical patterns and reactivity in the vast remote
226 ocean basins, where most of the destruction of tropospheric ozone (O_3) and methane
227 (CH_4) occurs. Reactivity is defined here as in P2017 to include the production and loss

228 of O₃ (P-O₃ and L-O₃, ppb/d) and loss of CH₄ (L-CH₄, ppb/d). Here we report on this
229 model-derived product that was proposed for ATom, the daily averaged reaction rates
230 determining the production and loss of O₃ and the loss of CH₄ for 10 s averaged air
231 parcels. We calculate these rates with 3D chemical models that include variations in
232 clouds and photolysis, and then assemble the statistical patterns describing the
233 heterogeneity (i.e., high spatial variability) of these rates and the underlying patterns of
234 reactive gases.

235 Tropospheric O₃ and CH₄ contribute to climate warming and global air pollution (Stocker
236 et al., 2013). Their abundances in the troposphere are controlled largely by tropospheric
237 chemical reactions. Thus, chemistry–climate assessments seeking to understand past
238 global change and make future projections for these greenhouse gases have focused on
239 the average tropospheric rates of production and loss and how these reactivities are
240 distributed in large semi-hemispheric zones throughout the troposphere (Griffiths et al.,
241 2021; Myhre et al., 2014; Naik et al., 2013; Prather et al., 2001; Stevenson, et al., 2006;
242 Stevenson, et al., 2013; Stevenson, et al., 2020; Voulgarakis et al., 2013; Young et al.,
243 2013). The models used in these assessments disagree on these overall CH₄ and O₃
244 reactivities (a.k.a. the budgets), and resolving the cause of such differences is stymied
245 because of the large number of processes involved and the resulting highly heterogeneous
246 distribution of chemical species that drive the reactions. Simply put, the models use
247 emissions, photochemistry, and meteorological data to generate the distribution of key
248 species such as nitrogen oxides (NO_x = NO + NO₂) and hydrogen peroxide (HOOH)
249 (step 1) and then calculate the CH₄ and O₃ reactivities from these species (step 2). There
250 is no single average measurement that can test the verisimilitude of the models.
251 Stratospheric studies such as Douglass et al. (1999) have provided a quantitative basis for
252 testing chemistry and transport, and defining model errors; but few of these studies have
253 tackled the problem of modeling the heterogeneity of tropospheric chemistry. The major
254 model differences lie in the first step, because when we specify the mix of key chemical
255 species, most models agree on the CH₄ and O₃ chemical budgets (*P2018*). The intent of
256 ATom was to collect an atmospheric sampling of all the key species and the statistics
257 defining their spatial variability, and thus that of the reactivities of CH₄ and O₃.

258 Many studies have explored the ability of chemistry–transport models (CTMs) to resolve
259 finer scales such as pollution layers (Eastham and Jacob, 2017; Rastigejev et al., 2010;
260 Tie et al., 2010; Young et al., 2018; Zhuang et al., 2018), but these have not had the
261 chemical observations (statistics) to evaluate model performance. In a great use of
262 chemical statistics, Yu et al. (2016) used 60 s data (~12 km) from the SEAC⁴RS aircraft
263 mission to compare cumulative probability densities (PDs) of NO_x, O₃, HCHO and
264 isoprene over the Southeast US with the GEOS-Chem CTM run at different resolutions.
265 They identified clear biases at the high and low ends of the distribution, providing a new
266 test of models based on the statistics rather than mean values. Heald et al. (2011)
267 gathered high-resolution profiling of organic and sulfate aerosols from 17 aircraft
268 missions and calculated statistics (mean, median, quartiles) but only compared with the
269 modeled means. The HIAPER Pole-to-Pole Observations (HIPPO) aircraft mission
270 (Wofsy, 2011) was a precursor to ATom with regular profiling of the mid-Pacific
271 including high-frequency 10 s sampling that identified the small scales of variability
272 throughout the troposphere. HIPPO measurements were limited in species, lacking O₃,

273 NO_x and many of the core species needed for reactivity calculations. ATom, with a full
274 suite of reactive species and profiling through the Atlantic basin, provides a wealth of
275 chemical statistics that challenge the global chemistry models.

276 One main task here is the assembly of the modeling data stream (MDS), which provides
277 flight-wise continuous 10 s data (air parcels) for the key reactive species. The MDS is
278 based on direct observations and interpolation methods to fill gaps as documented the
279 Supplement. Using the version 0 of the MDS, we have six chemical models calculating
280 the 24 h reactivities, producing a reactivity data stream (RDS version 0) using protocols
281 noted in the Prologue (P2017) and described further in Sect. 3.2. There, we describe the
282 updated modeling protocol RDS* necessary to address measurement noise in PAN and
283 HNO₄, which can be very short-lived. In Sect. 4, we examine the statistics of reactivity
284 over the Atlantic and Pacific oceans, focusing on air parcels with high reactivity; for
285 example, 10% of the parcels produce 25-35% of total reactivity over the oceans. We
286 compare these ATom-1 statistics, species and reactivities with August climatologies from
287 six global chemistry models. In one surprising result, ATom-1 shows a more reactive
288 tropical troposphere than found in most models' climatologies associated with higher
289 NO_x levels than in the models. Section 5 concludes that the ATom PDs based on 10s air
290 parcels do provide a valid chemistry metric for global models with 1° resolution. It also
291 presents some examples where ATom measurements and modeling can test the chemical
292 relationships and may address the cause of differences in the O₃ and CH₄ budgets
293 currently seen across the models. With this paper we release the full ATom MDS-2b
294 from all four deployments along with the updated RDS-2b reactivities from the UCI
295 model.

296 **3 Models and data**

297 **3.1 The modeling data stream (MDS)**

298 The ATom mission was designed to collect a multi-species, detailed chemical
299 climatology that documents the spatial patterns of chemical heterogeneity throughout the
300 remote troposphere. Figure S1 in the Supplement maps the 48 research flights, and the
301 Supplement has tables summarizing each flight. We required a complete set of key
302 species in each air parcel to initialize the models that calculate the CH₄ and O₃
303 reactivities. We choose the key reactive species (H₂O, O₃, CO, CH₄, NO_x, NO_xPSS,
304 HNO₃, HNO₄, PAN, CH₂O, H₂O₂, CH₃OOH, acetone, acetaldehyde, C₂H₆, C₃H₈, *i*-
305 C₄H₁₀, *n*-C₄H₁₀, alkanes, C₂H₄, alkenes, C₂H₂, C₅H₈, benzene, toluene, xylene,
306 CH₃ONO₂, C₂H₅ONO₂, RONO₂, CH₃OH) directly from the ATom measurements and
307 then add corollary species or other observational data indicative of industrial or biomass
308 burning pollution or atmospheric processing (HCN, CH₃CN, SF₆, relative humidity,
309 aerosol surface area (four modes), and cloud indicator). We choose 10 s averages for our
310 air parcels as a compromise and because the 10 s merged data are a standard product
311 (Wofsy et al., 2018). A few instruments measure at 1 s intervals, but the variability at
312 this scale is not that different from 10 s averages (Fig. S2). Most of the key species are
313 reported as 10 s values, with some being averaged or sampled at 30 s or longer such as
314 ~90 s for some flask measurements.

315 Throughout ATom, gaps occur in individual species on a range of timescales due to
316 calibration cycles, sampling rates or instrument malfunction. The generation of the MDS
317 uses a range of methods to fill these gaps and assigns a flag index to each species and
318 data point to allow users to identify direct measurements and methods used for gap-
319 filling. Where two instruments measure the same species, the MDS selects a primary
320 measurement and identifies which instrument was used with a flag. The methodology
321 and species-specific information on how the current MDS version 2 (MDS-2) is
322 constructed, plus statistics on the 48 research flights and the 146,494 10 s air parcels in
323 MDS-2 are given in the Supplement.

324 Over the course of this study, several MDS versions were developed and tested, including
325 model-derived RDSs from these versions, some of which are used in this paper. In early
326 ATom science team meetings, there was concern about the accuracy of NO₂ direct
327 measurements when at very low concentrations. A group prepared an estimate for NO_x
328 using the NO and O₃ measurements to calculate a photostationary value for NO₂ and thus
329 NO_x. This PSS-NO_x became the primary NO_x source in version 0 (i.e., MDS-0). With
330 MDS-0, we chose to gap-fill using correlations with CO to estimate the variability of the
331 missing measurement over the gap. The science team then rejected PSS-NO_x as a proxy,
332 and we reverted to the observed NO + NO₂ resulting in NO_x values that are 25 % larger
333 on average than in MDS-0 (unweighted mean of 66 vs. 52 ppt). This change affected P-
334 O₃ most and L-CH₄ least. We then estimated errors in the gap-filling and found that CO
335 had little skill as a proxy for most other species. With MDS-2, we optimized and tested
336 the treatments of gap-filling and lower limit of detection, along with other quality
337 controls. With continued analysis of the unusually reactive East Pacific region, we
338 determined that the method of long-gap filling for NO_x resulted in propagation of high
339 NO_x levels from the over-land profiles into the over-water profiles in the tropics. We
340 separated these two set of profiles used for long-gap NO_x filling and created an updated
341 version 2b. This experience points to the importance of having reliable, continuous NO_x
342 measurements. MDS-2b is fully documented in the Supplementary Information.
343

344 **3.2 The reactivity data stream (RDS)**

345 The concept of using an MDS to initialize 3D global chemistry models and calculate an
346 RDS was developed in the pre-ATom methodology papers (P2017; P2018). In this
347 paper, we use the original six models for their August chemical statistics, and we use 5 of
348 them plus a box model to calculate the reactivities, see Table 1. The RDS is really a
349 protocol applied to the MDS. It is introduced in the Prologue and the details can be
350 found in P2018. A model grid cell chosen to be close to the measured parcel is initialized
351 with all the core reactive species needed for a regular chemistry simulation. The model is
352 then integrated over 24 h without transport or mixing, without scavenging, and without
353 emissions. Each global model uses its own varying cloud fields for the period to
354 calculate photolysis rates; but the FOAM box model simply takes the instant J-values as
355 measured on the flight and applies a diurnal scaling. We initialize with the core species
356 and let the radicals (OH, HO₂, RO₂) come quickly into photochemical balance. The 24 h
357 integration is not overly sensitive to the start time of the integration, and thus models do

358 not have to synchronize with the local time of observation (see P2018's Fig. S8 and Table
359 S8).

360 The initial ATom-1 reactivities came from MDS-0 and six of the models in Table 1.
361 Although these RDS-0 model results are now out of date because of the move to MDS-
362 2b, they provide critical information on how models agree, or disagree, in calculating the
363 RDS using the ATom protocol. Thus we include them here as a cross-model comparison.
364 Given the excellent agreement at the parcel level using three models (GC, GMI, UCI),
365 and with a desire to avoid wasting the community's time, we continued the analysis of
366 MDS-2b with just our local UCI CTM. This decision may need to be revisited.

367 Statistics for the three reactivities for six models using MDS-0 are given in Table 2 and
368 Table S8 for three domains: global (all points), Pacific (oceanic data from 53° S to 60° N)
369 and Atlantic (same constraints as Pacific). The statistics try to achieve equal latitude-by-
370 pressure sampling by weighting each ATom parcel inversely according to the number of
371 parcels in each 10° latitude by 100 hPa bin, and each point is also cosine(latitude)
372 weighted. We calculate the means and medians plus the percent of total reactivity in the
373 top 10 % of the weighted parcels (Table 2) and also the mean reactivity of the top 10 %,
374 percent of total reactivity in the top 50 %, 10 % and 3 % plus the mean *J* values (Table
375 S8).

376 These six-model version 0 statistics are shown alongside the version 2b results using the
377 current UCIZ model but with a new protocol designated RDS*. While investigating
378 sensitivities in the RDS, we found an inconsistency between the reported concentrations
379 of both pernitric acid (HNO₄) and peroxyacetyl nitrate (PAN) with respect to the
380 chemical kinetics used in the models. High concentrations (100 ppt, attributed to
381 instrument noise) were reported under conditions where the thermal decomposition
382 frequency was > 0.4 per hour in the lower troposphere (> 253 K for HNO₄ and > 291 K
383 for PAN). Thus, these species instantly become NO_x. While these measurements are
384 clearly spurious, there is no easy fix. We developed a new protocol, RDS*, that allows
385 both species to decay for 24 h using their local thermal decomposition rate before being
386 used in the model. This protocol avoids much of the fast thermal release of NO_x in the
387 lower atmosphere during the first 24 h of the RDS calculation, but does not affect the
388 release of NO_x from photolysis or OH reactions in the upper troposphere where thermal
389 decomposition is inconsequential. It is possible that some of the high concentrations of
390 HNO₄ and PAN in the lower troposphere are real and that we are missing this large
391 source of NO_x with the RDS* protocol, but we find no obvious sources of these species
392 in the remote oceanic regions that would produce enough to match the thermal loss. Both
393 this problem and its solution do not affect the initial NO_x values.

394
395 We present the RDS-2b reactivities calculated under the RDS* protocol with the UCI
396 CTM developed by Xin Zhu for P2017 and P2018 (designated UCIZ*) as our best results
397 in the final column of Tables 2 and S8. We added diagnostics that give us confidence in
398 our O₃ reactivities: the approximate P-O₃ and L-O₃ based on the limited reactions (rates
399 2abd and 3abc above) actually predict the calculated 24 h O₃ tendency, see Fig. S6.
400 Considering the ocean basin observations only, P-L ranges from -12 to +15 ppb/d. The
401 mean error in P-L is about -0.01 ppb/d, and the root-mean-squared error is about 0.04

402 ppb/d, convincing us that we have correctly diagnosed the P-O3 and L-O3 terms.
403 Following the practice of the GMI model, we also record the initial and 24-hour
404 abundances of all the ATom species to check that nothing unusual altered the species
405 abundance in each cell over the 24 hours.

406

407 **3.3 Inter-model differences**

408

409 Variations in reactivities due to clouds are an irreducible source of uncertainty:
410 predicting the cloud-driven photolysis rates that a shearing air parcel will experience over
411 24 h is not possible here. The protocol uses 5 separated 24 h days to average over
412 synoptically varying cloud conditions. The standard deviation (σ) of the 5 d, as a
413 percentage of the 5 d mean, is averaged over all parcels and shown in Table S9 for the
414 five global models. Three central models (GC, GMI, UCI) show 9 %–10 % $\sigma(J_s)$ values
415 and similar $\sigma(R_s)$ values as expected if the variation in J values is driving the reactivities.
416 Two models (GISS, NCAR) have 12 %–17 % $\sigma(J_s)$, which might be explained by more
417 opaque clouds, but the amplified $\sigma(R)$ values (14 %–32 %) are inexplicable. This
418 discrepancy needs to be resolved before using these two models for ATom RDS analysis.

419

420 Inter-model differences are shown in the parcel-by-parcel root-mean-square (rms)
421 differences for RDS-0 in Table 3. Even when models adopt standard kinetic rates and
422 cross sections (i.e., Burkholder et al., 2015), the number of species and chemical
423 mechanisms included, as well as the treatment of families of similar species or
424 intermediate short-lived reaction products, varies across models. For example, UCI
425 considers about 32 reactive gases, whereas GC and GMI have over 100, and FOAM has
426 more than 600. The other major difference across models is photolysis, with models
427 having different cloud data and different methods for calculating photolysis rates in
428 cloudy atmospheres (H2018). The three central models (GC, GMI, UCI) in terms of their
429 5 d variability (Table S9) are also most closely alike in these statistics with rms = 20 %–
430 30 % for L-CH4 up to 26 %–35 % for P-O3. These rms values appear to be about as
431 close as any two models can get. The intra-model rms for different years (UCI 2016
432 versus 1997) is 10 %–13 % and shows that we are seeing basic differences in the
433 chemical models across GC, GMI, and UCI. FOAM is the next closest to these central
434 models, but it will inherently have a larger rms because it is a 1 d calculation and not a 5
435 d average. NCAR's rms is consistently higher and likely related to what is seen in the 5 d
436 σ values in Table S9. GISS is clearly different from all the others (L-CH4 rms > 100 %
437 while L-O3 rms < 66 %).

438

439 **4 Results**

440

441 Our analysis of the reactivities uses the six-model RDS-0 results to examine the
442 consistency in calculating the R_s across models. Thereafter, we rely on the similar results
443 from the three central models (GC, GMI, UCI) to justify use of UCIZ* with MDS-2b as
444 our best estimate for ATom reactivities. The uncertainty in this estimate can be
445 approximated by the inter-model spread of the central models as discussed above. When
446 evaluating the model climatologies for chemical species, we use MDS-2b. A summary of
447 the key data files used here, as well as their sources and contents, is given in Table 4.

448

449 **4.1 Probability densities of the reactivities**

450

451 The reactivities for three large domains (Global, Pacific, Atlantic) from the six-model
452 RDS-0 are summarized in Tables 2 and S8. Sorted PDs for the three Rs and Pacific and
453 Atlantic Ocean basins are plotted in Fig. 1 and show the importance of the most reactive
454 “hot” parcels with deeply convex curves and the sharp upturn in R values above 0.9
455 cumulative weight (top 10 %). Both basins show a similar emphasis on the most reactive
456 hot parcels: 80 % – 90 % of total R is in the top 50 % of the parcels, 25 % – 35 % is in
457 the top 10 %, and about 10 % – 14 % is in the top 3 %. The corollary is that the bottom
458 50 % parcels control only 10 % – 20 % of the total reactivity, which is why the median is
459 less than mean (except for P-O3 in the Atlantic).

460

461 The enhancement factor for the top 50 % L-CH4 parcels is 2.0 (84 % of reactivity in 42
462 % of mass) given that our 53 °S – 60 °N transects cover 83 % of the air mass below 200
463 hPa and assuming that L-CH4 is negligible poleward of these transects. This
464 enhancement factor is a large-scale feature because the tropical lower troposphere, being
465 warm and wet with high sun, dominates the budget. It is seen in previous model
466 intercomparisons that calculate budgets in large tropospheric blocks like Voulgarakis et
467 al. (2013) with 63 % of L-CH4 in 31 % of the air mass (500 hPa–surface, 30 °S – 30 °N).
468 The impact of the extremely hot parcels and the heterogeneity seen in the ATom 10 s
469 parcels is evident in the steep slopes above the 90th percentile, yielding enhancement
470 factors of 3 to 4.

471

472 Each R value and each ocean has a unique shape; for example L-O3 in the Atlantic is
473 almost two straight lines breaking at the 50th percentile. In Fig. 1 the agreement across
474 all models (except GISS) is clear, indicating that the conclusion in *P2018* (i.e., that most
475 global chemistry models agree on the O₃ and CH₄ budgets if given the chemical
476 composition) also holds for the ATom-measured chemical composition. Comparing the
477 brown (UCI, RDS-0) and black-dashed (UCIZ, RDS*-2) lines, we find that the shift from
478 MDS-0 to MDS-2b plus the new RDS* (HNO₄+PAN) protocol produces large reductions
479 in P-O3 for all cumulative weights and small reductions in L-CH4 for the upper 5th
480 percentile. We conclude that accurate modeling of chemical composition of the 80th and
481 greater percentiles is important but that modest errors in the lowest 50th percentile are
482 inconsequential; effectively, some parcels matter more than others (P2017).

483

484 How well does this ATom analysis work as a model intercomparison project? Overall,
485 we find that most models give similar results when presented with the ATom-1 MDS.
486 The broad agreement of the cumulative reactive PDs across a range of model
487 formulations using differing levels of chemical complexity shows this approach is robust.
488 The different protocols for calculating reactivities as well as the uncertainty in cloud
489 fields appear to have a small impact on the shape of the cumulative PDs but are
490 informative regarding the minimum structural uncertainty in estimating the 24 h
491 reactivity of a well-measured air parcel.

492 **4.2 Spatial heterogeneity of tropospheric chemistry**

493 A critical unknown for tropospheric chemistry modeling is what resolution is needed to
494 correctly calculate the budgets of key gases. A similar question was addressed in Yu et
495 al. (2016) for the isoprene oxidation pathways using a model with variable resolution
496 (500 km, 250 km and 30 km) compared to aircraft measurements; see also ship plume
497 chemistry in Charlton-Perez et al. (2009). ATom's 10 s air parcels measure 2 km
498 (horizontal) by 80 m (vertical) during most profiles. There are obviously some chemical
499 structures below the 10 s air parcels. Only some ATom measurements are archived at 1
500 Hz, and we examine a test case using 1 s data for O₃ and H₂O for a mid-ocean descent
501 between Anchorage and Kona in Fig. S2a in the Supplement. Some of the 1 s (200 m by
502 8 m) variability is clearly lost with 10 s averaging, but 10 s averaging preserves most of
503 the variability. Lines in Fig. S2 demark 400 m in altitude, and most of the variability
504 occurs on this larger, model-resolved scale. Fig. S2b shows the 10 s reactivities during
505 that descent and also indicates that much of the variability occurs at 400 m vertical scales.
506 A more quantitative example using all the tropical ATom reactivities is shown in
507 comparisons with probability densities below (Fig. 5).

508 How important is it for the models to represent the extremes of reactivity? While the
509 sorted reactivity curves (Fig. 1, Tables 2 & S8) continue to steepen from the 90th to 97th
510 percentile, the slope does not change that much. Thus we can estimate the 99th+
511 percentile contributes <5% of the total reactivity. Thus, if our model misses the top 1 %
512 of reactive air parcels (e.g., due to the inability to simulate intensely reactive thin
513 pollution layers) then we miss at most 5 % of the total reactivity. This finding is new and
514 encouraging, and it needs to be verified with the ATom-2, 3, and 4 data.

515 The spatial structures and variability of reactivity as sampled by the ATom tropical
516 transects (central Pacific, eastern Pacific and Atlantic) are presented as nine panels in Fig.
517 2. Here, the UCIZ RDS*-2 reactivities are averaged and plotted in 1° latitude by 200 m
518 thick cells, comparable to some global models (e.g., GMI, NCAR, UCI). We separate the
519 eastern Pacific (121° W, research flight (RF) 1) from the Central Pacific (RFs 3, 4 and 5)
520 because we are looking for contiguous latitude-by-pressure structures.

521 In the central Pacific (Fig. 2adg), highly reactive (hot) P-O₃ parcels (> 6 ppb/d) occur in
522 larger, connected air masses at latitudes 20°–22°N and pressure altitudes 2–3 km and in
523 more scattered parcels (> 3 ppb/d) below 5 km down to 20°S. High L-O₃ and L-CH₄
524 coincide with this 20°–22°N air mass and also with some high P-O₃ at lower latitudes.
525 This pattern of overlapping extremes in all three Rs is surprising because the models'
526 mid-Pacific climatologies show a separation between regions of high L-O₃ (lower-middle
527 troposphere) and high P-O₃ (upper troposphere, as seen in P2017's Fig. 3). The obvious
528 explanation is that the models leave most of the lightning-produced NO_x in the upper
529 troposphere. The ATom profiling seems to catch reactive regions in adjacent profiles
530 separate by a few hundred kilometers, scales easily resolvable with 3D models.

531 In the eastern Pacific (Fig. 2beh), the overlap of outbound and return profiles enhances
532 the spatial sampling over the 10 h flight. The region of very large L-O₃ (> 5 ppb/d) is
533 extensive, beginning at 5–6 km at 10°N and broadening to 2–8 km at 28°N. The region
534 of L-CH₄ is similar, but loss at the upper altitudes of this air mass is attenuated because
535 of the temperature dependence of L-CH₄ and possibly because of differing OH:HO₂
536 ratios with altitude. Large P-O₃ (> 3 ppb/day) occurs only in the center of this highly

537 reactive L-O₃/L-CH₄ region, suggesting that NO_x is not as evenly distributed as is HO_x.
538 Highly reactive (hot) P-O₃ parcels (> 4 ppb/day) occur only in the upper troposphere (8–
539 12 km) and only in the sub-tropics. ATom-1 RF1 (29 Jul 2016) occurred during the
540 North American Monsoon when there was easterly flow off Mexico, thus the high
541 reactivity of this large air mass indicates that continental deep convection with lightning
542 NO_x is a source of high reactivity for both O₃ and CH₄.

543 In the Atlantic (Fig. 2cfi) we also see similar air masses through successive profiles,
544 particularly in the northern tropics. The Atlantic P-O₃ shows high-altitude reactivity
545 similar to the eastern Pacific. Likewise, the large values of L-O₃ and L-CH₄ match the
546 eastern Pacific and not central Pacific. Unlike either Pacific transect, the Atlantic L-O₃
547 and L-CH₄ show some high reactivity below 1 km altitude. Overall, the ATom-1
548 profiling clearly identifies extended air masses of high L-O₃ and L-CH₄ extending over
549 2–5 km in altitude and 10° of latitude. The high P-O₃ regions tend to be much more
550 heterogeneous with greatly reduced spatial extent, likely of recent convective origin as
551 for eastern Pacific.

552 Overall, the extensive ATom profiling identifies a heterogeneous mix of chemical
553 composition in the tropical Atlantic and Pacific, with a large range of reactivities. What
554 is important for those trying to model tropospheric chemistry is that the spatial scales of
555 variability seen in Fig. 2 should be within the capability of modern global models.

556 **4.3 Testing model climatologies**

557 The ATom data set provides a unique opportunity to test CTMs and CCMs in a
558 climatological sense. In this section, we compare ATom-1 data and the six models'
559 chemical statistics for mid-August used in P2017. The ATom profiles cannot be easily
560 compared point by point with CCMs, and we use statistical measures of the three
561 reactivities in the three tropical basins: mean profiles in Fig. 3 and PDs in Fig. 5.

562 **4.3.1 Profiles**

563 For P-O₃ profiles (top row, Fig. 3), the agreement between models and measurements is
564 passable except for the 0–2 km region in both Central and Eastern Pacific, where the
565 models fail to predict the observed 2 ppb/d O₃ production. In the Central Pacific at 3–12
566 km, ATom-1 results agree with models, showing ozone production of about 1 ppb/day.
567 In the Eastern Pacific and Atlantic at 3–12 km, ATom-1 results also agree with models,
568 but at a higher ozone production of about 2 ppb/day. This pattern indicates that in the
569 Central Pacific, the NO_x+HO_x combination that produces ozone is suppressed below 2
570 km in all the models. In the upper troposphere, 10–12 km, of the Eastern Pacific and
571 Atlantic, ATom P-O₃ values show a jump to 3 ppb/d, which is only partly reproduced in
572 the models. We take this pattern as evidence for lightning NO_x production and export
573 over the adjacent continents.

574 For L-O₃ (middle row, Fig. 3) in the central Pacific, ATom-1 results match the
575 throughout the 0–12 km range (except GISS). Moving to the eastern Pacific and Atlantic,
576 most models show a mid-level peak above 2 km, while ATom-1 shows even larger peak
577 L-O₃, especially in the Eastern Pacific at 3–6 km where L-O₃ > 4 ppb/d. This mid-

578 tropospheric peak is evident in the curtain plots of Fig. 2 and likely due to easterly mid-
579 tropospheric flow from convection over Mexico at that specific time (29 July 2016).
580 Similarly, the ATom reactivity at 1–3 km in the Atlantic is associated with biomass
581 burning in Africa and was measured in other trace species. Thus, in terms of L-O3, the
582 ATom–model differences may be due to specific meteorological conditions, and this
583 could be tested with CTMs using 2016 meteorology and wildfires.

584

585 For L-CH4 (bottom row, Fig. 3), the ATom-model patterns are similar to L-O3, including
586 the large ATom-only losses (> 1.5 ppb/d over 3–6 km) in the eastern Pacific, but with
587 higher reactivities occurring at slightly lower altitudes because of the large negative
588 temperature dependence of reaction (1). L-O3 is dominated by O(1D) and HO₂ loss,
589 while L-CH4 is limited to OH loss. Overall, there is clear evidence that the Atlantic and
590 Pacific have very different chemical mixtures controlling the reactivities and that
591 convection over land (monsoon or biomass burning) creates air masses that are still
592 highly reactive a day or so later.

593

594 **4.3.2 Key species**

595 The deficit in modeled P-O3 in the central and eastern Pacific at 0–2 km altitude points to
596 a NOx deficiency in the models, and this becomes obvious in the comparison of the PD
597 histograms for NOx shown in Fig. 4. Over 0–12 km (first row), ATom has a reduced
598 frequency of parcels with 1–10 ppt and a corresponding increase in parcels with 20–60
599 ppt; this discrepancy is amplified in the lower troposphere, 0–4 km (second row). The
600 obvious source of this oceanic NOx is lightning since oceanic sources of organonitrates
601 or other nitrate species measured on ATom could not supply this amount. The ATom
602 statistics indicate such a lightning source must be mixed down into the boundary layer.
603 In the eastern Pacific and Atlantic, the full troposphere PD more closely matches the
604 models, including bump in 100–300 ppt NOx which is probably direct outflow from very
605 deep convection with lightning over the neighboring continents. Overall, the models
606 appear to be missing significant NOx sources in all three regions below 4 km.

607

608 In Fig. 4, we also look at the histograms for the key HOx-related species HOOH (third
609 row) and HCHO (fourth row). For these species, the ATom–model agreement is
610 generally good. If anything, the models tend to have too much HOOH. ATom shows
611 systematically large occurrences of low HOOH (50–200 ppt, especially central Pacific)
612 indicating, perhaps, that convective or cloud scavenging of HOOH is more effective than
613 is modeled. HCHO shows reasonable agreement in the Atlantic, but in both central and
614 eastern Pacific, the modeled low end (< 40 ppt) is simply not seen in the ATom data.
615 Also, the models are missing a strong HCHO peak at 300 ppt in the eastern Pacific,
616 probably convection-related specific to that time period. Thus, in terms of these HOx
617 precursors, the model climatologies appear to be at least as reactive as the ATom data.

618 While the ATom-1 data in Fig. 4 are limited to single transects, the model NOx
619 discrepancies apply across the three tropical regions, and the simple chemical statistics
620 for these flights alone are probably enough to identify measurement-model discrepancies.
621 For the HOx-related species, the models match the first-order statistics from ATom. In
622 terms of using ATom statistics as a model metric, it is encouraging that where some

623 individual models tend to deviate from their peers, they also deviate from the ATom-1
624 PDs.

625 **4.3.3 Probability densities**

626 Mean profiles do not reflect the heterogeneity seen in Fig. 2, and so we also examine the
627 PDs of the tropical reactivities (Fig. 5). The model PDs (colored lines connecting open
628 circles at the center of each bin) are calculated from the 1 d statistics for mid-August
629 (P2017) using the model blocks shown in Fig. S1. The model grid cells are weighted by
630 air mass and cosine(latitude) and limited to pressures greater than 200 hPa. The ATom
631 PDs (black lines connecting black open circles) are calculated from the 10 s data
632 weighted by (but not averaged over) the number of points in each 10° latitude by 200 hPa
633 pressure bin, and then also by cosine (latitude) to compare with the models. In addition,
634 a PD was calculated from the 1° by 200 m average grid-cell values in Fig. 2 (black Xs),
635 and this is also cosine(latitude)-weighted. To check if the high reactivities in the eastern
636 Pacific affected the whole Pacific PD, a separate PD using only central Pacific 10 s data
637 was calculated (gray lines connecting gray open circles). The mean reactivities (ppb/d)
638 from the models and ATom are given in the legend; note that the model values are based
639 on the August climatologies (P2017) and not the MDS-0 values in the table. The 'ATom'
640 legend values are the same as in Table 2. The PD binning is shown by the open circles,
641 and occurrences of off-scale reactivities are included in the last point.

642
643 For the Pacific (eastern + central, left columns, Fig. 5), the modeled PD climatologies are
644 similar for each of the reactivities (except GISS), and there is fairly good agreement with
645 the ATom-1 PDs. For the Atlantic (right columns, Fig. 5), the models show a larger
646 spread presumably due to the differing influence of pollution from neighboring
647 continents. The ATom-1 Atlantic PDs also show slightly larger disagreement with the
648 models (e.g., the maximum in P-O3 at 1–2 ppb/d and minimum in L-O3 at 2–3 ppb/d)
649 and the notably higher frequency of hot spots with L-O3 > 5 ppb/d. The influence of the
650 extreme eastern Pacific reactivities are seen in the statistics generated from the central
651 Pacific values only (CPac, gray circles), e.g., the mean value for L-O3 drops from 1.42 to
652 1.17 ppb/d.

653
654 The ability to test a model's reactivity statistics with the ATom 10 s data is not obvious,
655 but the PDs based on 1° latitude by 200 m altitude cells (the black Xs) are remarkably
656 close to the PDs based on 2 km (horizontal) by 80 m (vertical) 10 s parcels. With the
657 coarser resolution, we see a slight shift of points from the ends of the PD to the middle as
658 expected, but we find once again, that the loss in high-frequency, below-model grid-cell
659 resolution is not great. Both ATom-derived PDs more closely resemble each other than
660 any model PD. Thus, current global chemistry models with resolutions of about 100 km
661 by 400 m should be able to capture much of the wide range of chemical heterogeneity in
662 the atmosphere, which for the oceanic transects is, we believe, adequately resolved by the
663 10 s ATom measurements. Perhaps more surprising, given the different mean profiles in
664 Fig. 3, is that the five model PDs in Fig. 5 look very much alike.

665

666 **5 Discussion and path forward**

667

668 **5.1 Major findings**

669

670 This paper opens a door for what the community can do with the ATom measurements
671 and the derived products. ATom's mix of key species allows us to calculate the reactivity
672 of the air parcels and hopefully may become standard for tropospheric chemistry
673 campaigns. We find that the reactivity of the troposphere with respect to O₃ and CH₄ is
674 dominated by a fraction of the air parcels but not by so small and infrequent a fraction as
675 to challenge the ability of current CTMs to simulate these observations and thus be used
676 to study the oxidation budgets. In comparing ATom results with modeled climatologies,
677 we find a systematic ATom-model difference: models show a large relative drop in O₃
678 production below 2 km over the tropical oceans but ATom shows an increase (C.Pac.), no
679 change (E.Pac.) or a much lesser drop (Atl.). We traced this result to the lack of NO_x at
680 20–60 ppt levels in the models below 4 km and believe it provides a clear challenge in
681 modeling ozone.

682 Building our chemical statistics (PDs) from the ATom 10 s air parcels on a scale of 2 km
683 by 80 m, we can identify the fundamental scales of spatial heterogeneity in tropospheric
684 chemistry. Although heterogeneity occurs at the finest scales (such as seen in some 1 s
685 observations) the majority of variability in terms of the O₃ and CH₄ budgets occurs across
686 scales larger than neighboring 2 km parcels. The PDs measured in ATom can be largely
687 captured by a global models' 100 km by 200 m grid cells in the lower troposphere. This
688 surprising result is evident by comparing the ATom 1D PDs – both species and
689 reactivities – with those from the models' climatologies (Fig. 5). These comparisons
690 show that the modeled PDs are consistent with the innate chemical heterogeneity of the
691 troposphere as measured by the 10 s parcels in ATom. A related conclusion for biomass
692 burning smoke particles is found by Schill et al. (2020), where most of the smoke appears
693 in the background rather than in pollution plumes, and therefore much of the variability
694 occurs on synoptic scales resolved by global models (see their Fig. 1 compared with Fig.
695 2 here).

696 **5.2 Opportunities and lessons learned**

697 As a quick look at the opportunities provided by the ATom data, we present an example
698 based on the Wolfe et al (2019) study, which used the F0AM model and semi-analytical
699 arguments to show that troposphere HCHO columns (measurable by satellite and ATom)
700 are related to OH columns (measured by ATom) and thus to CH₄ loss. Fig. 6 extends the
701 Wolfe et al study using the individual air parcels and plotting L-CH₄ (ppb/d) versus
702 HCHO (ppt) for the three tropical regions where most of the CH₄ loss occurs. The
703 relationship is linear but with a lot of scatter and has slopes ranging from 3.5 to 4.4 per
704 day over the three tropical regions; but for the largest reactivities (0-4 km, 1–3 ppb/d), L-
705 CH₄ is not so well correlated with HCHO.

706

707 As is usual with new model intercomparison projects, we have an opportunity to identify
708 model 'features' and identify errors. In the UCI model, an error in the lumped alkane
709 formulation (averaging alkanes C₃H₈ and higher) did not show up in P2018, where UCI
710 supplied all the species, but when the ATom data were used, the UCI model became an

711 outlier. Once found, this problem was readily fixed (hence the current UCIZ model
712 version). Inclusion of the F0AM model with its extensive hydrocarbon oxidation
713 mechanism provided an interesting contrast with the simpler chemistry in the global
714 CCM/CTMs. For a better comparison of the chemical mechanisms, we should have
715 F0AM use 5 d of photolysis fields from one of the CTMs. The anomalous GISS results
716 have been examined by a co-author, but no clear causes have been identified as of this
717 publication. The problem goes beyond just the implementation of the RDS protocol, as it
718 shows up in the model climatology (Fig. 4 & 5, also in P2017).

719 Decadal-scale shifts in the budgets of O₃ and CH₄ are likely to be evident through the
720 statistical patterns of the key species, rather than simply via average profiles. The
721 underlying design of ATom was to collect enough data to develop such a multivariate
722 chemical climatology. As a quick look across the four deployments, we show the joint
723 2D PDs on a logarithmic scale as in P2017 for HOOH versus NO_x in Fig. 7. The patterns
724 for the tropical central Pacific are quite similar for the four seasons of ATom
725 deployments, and the fitted ellipses are almost identical for ATom 2, 3 and 4. Thus, for
726 these species in the central Pacific, we believe that ATom provides a benchmark of the
727 2016-2018 chemical state, one that can be revisited with an aircraft mission in a decade to
728 detect changes in not only chemical composition but also reactivity.

729 ATom identifies which ‘highly reactive’ spatial or chemical environments could be
730 targeted in future campaigns for process studies or to provide a better link between
731 satellite observations and photochemical reactivity (e.g., E. Pacific mid-troposphere in
732 August, Fig. 2). The many corollary species measured by ATom (not directly involved in
733 CH₄ and O₃ chemistry) can provide clues to the origin or chemical processing of these
734 environments. We hope to engage a wider modeling community beyond the ATom
735 science team, as in H2018, in the calculation of photochemical processes, budgets, and
736 feedbacks based on all four ATom deployments.

737

738 *Data Availability.* The MDS-2b and RDS*-2b data for ATom 1, 2, 3 and 4 are presented
739 here as core ATom deliverables, and are posted temporarily on the NASA ESPO ATom
740 website (<https://espo.nasa.gov/atom/content/ATom>) and permanently on DRYAD|UCI
741 (<https://doi.org/10.7280/D1B12H>). This publication marks the public release of the
742 reactivity calculations for ATom 2, 3 and 4, but we have not yet analyzed these data, and
743 thus users should be aware and report any anomalous features to the lead authors via
744 haog2@uci.edu and mprather@uci.edu. Details of the ATom mission and data sets are
745 found on the NASA mission website (<https://espo.nasa.gov/atom/content/ATom>) and at
746 the final archive at Oak Ridge National Laboratory (ORNL;
747 https://daac.ornl.gov/ATOM/guides/ATom_merge.html). The MATLAB scripts and data
748 sets used in the analysis here are posted on Dryad (<https://doi.org/10.7280/D1Q699>).

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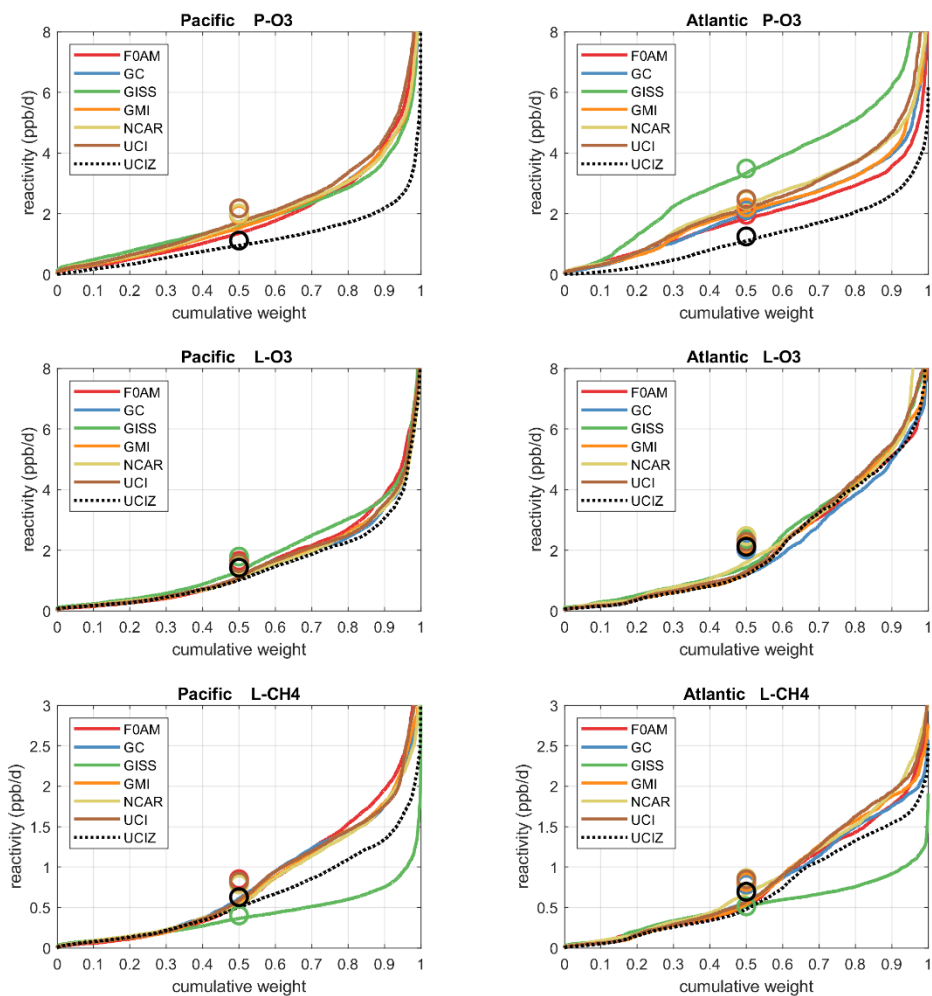
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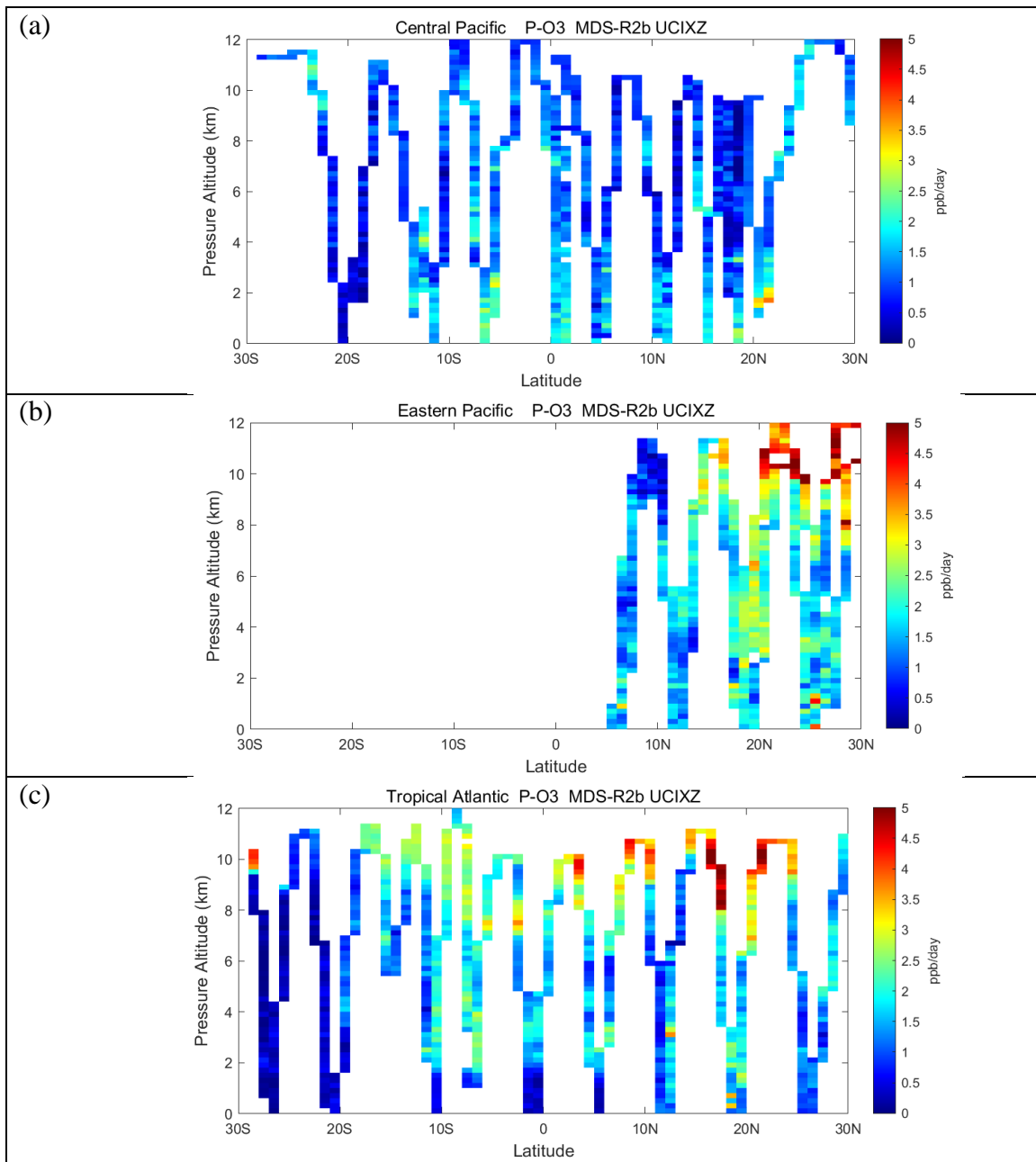
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Figures and Tables

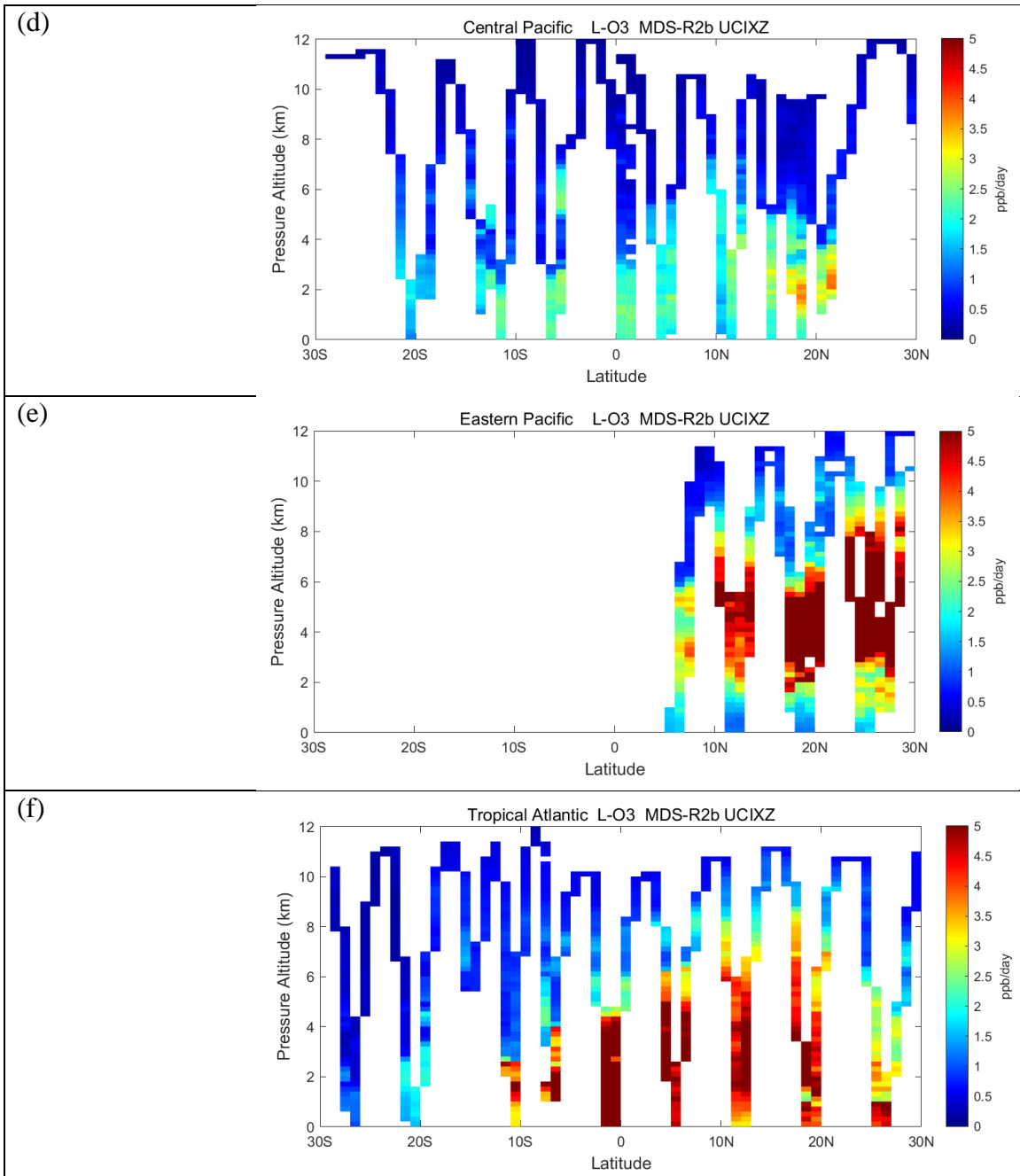


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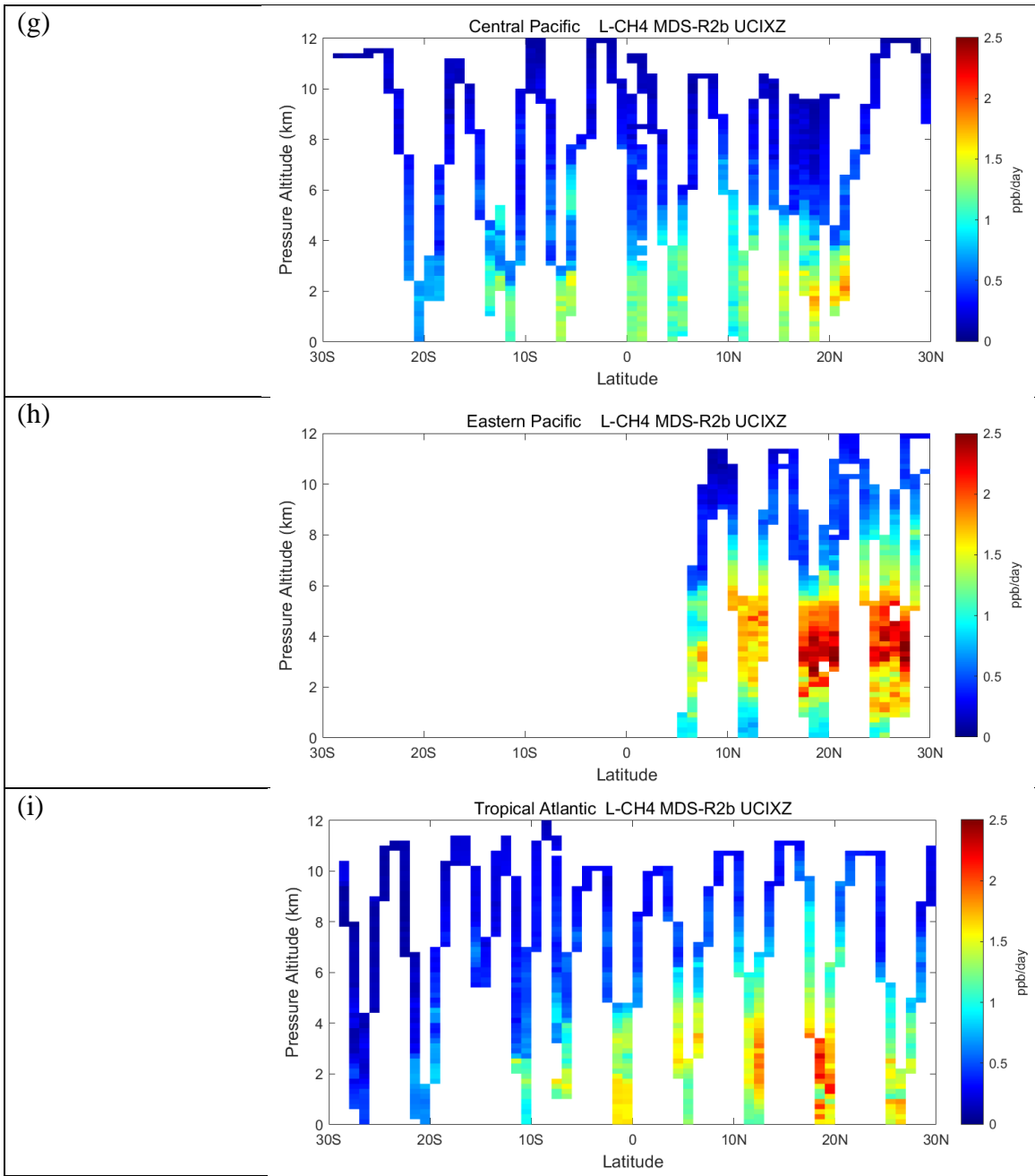
Fig. 1. Sorted reactivities (P-O3, L-O3, L-CH4, ppb/day; three successive rows) for the Pacific and Atlantic domains (53° S–60° N, two columns) of ATom-1. Each parcel is weighted, including cosine(latitude), see text. Results from six models using MDS-0 and the standard RDS protocol are shown with colored lines; the updated UCIZ CTM using MDS-2b with the RDS* protocol (HNO₄ and PAN damping) is shown as a black dashed line. The mean value for each model is shown with an open circle plotted at the 50th percentile. (Flipped about the axes, this is a cumulative probability density function.)



941 **Fig. 2abc.** Curtain plots for P-O3 (0–5 ppb/d; Fig 2abc), L-O3 (0–5 ppb/d; Fig 2def) and L-CH4
 942 (0–2.5 ppb/d; Fig 2ghi) showing the profiling of ATom-1 flights in the central Pacific (RF 3, 4
 943 and 5; Fig 2adg), eastern Pacific (RF 1; Fig2 beh), and Atlantic (RF 7, 8, and 9; Fig2cfi).
 944 Reactivities are calculated with the current UCIZ CTM model using MDS-2b and the RDS*
 945 protocol, see text. The 10 s air parcels are averaged into 1° latitude and 200 m altitude bins.
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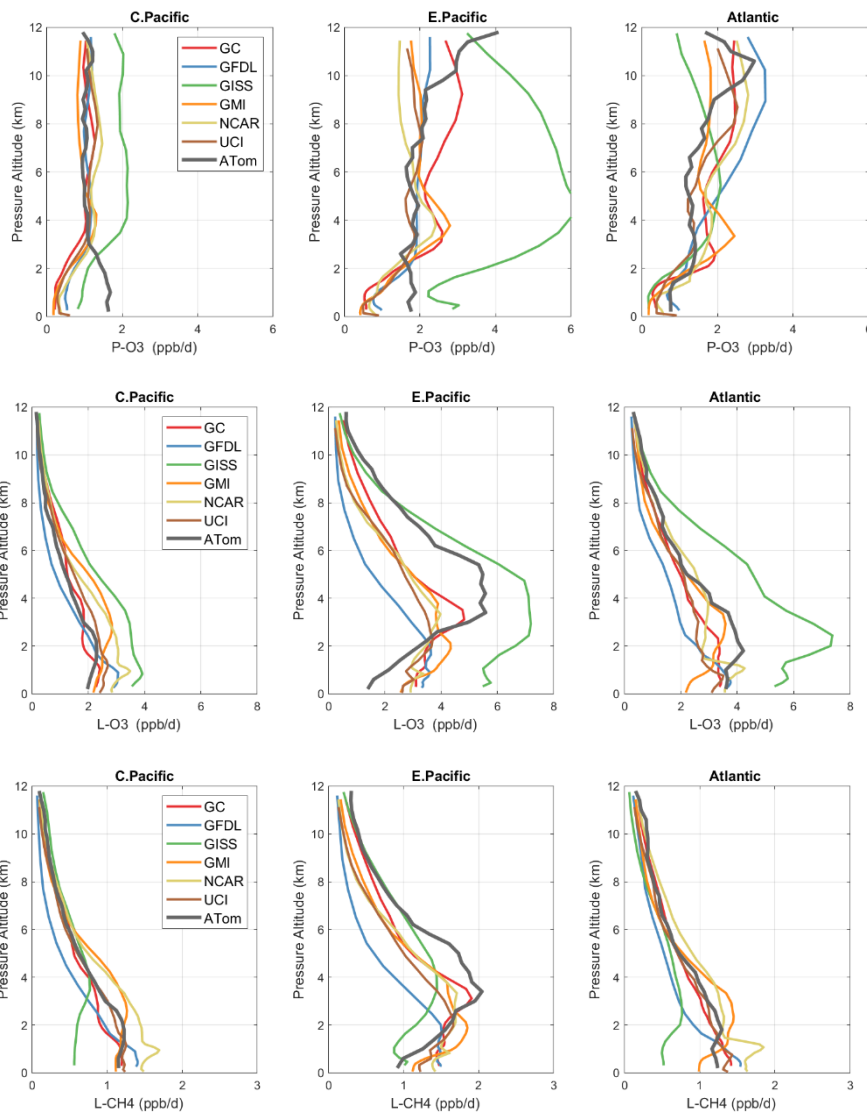


947 **Fig. 2def.**
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950 **Fig. 2ghi.**
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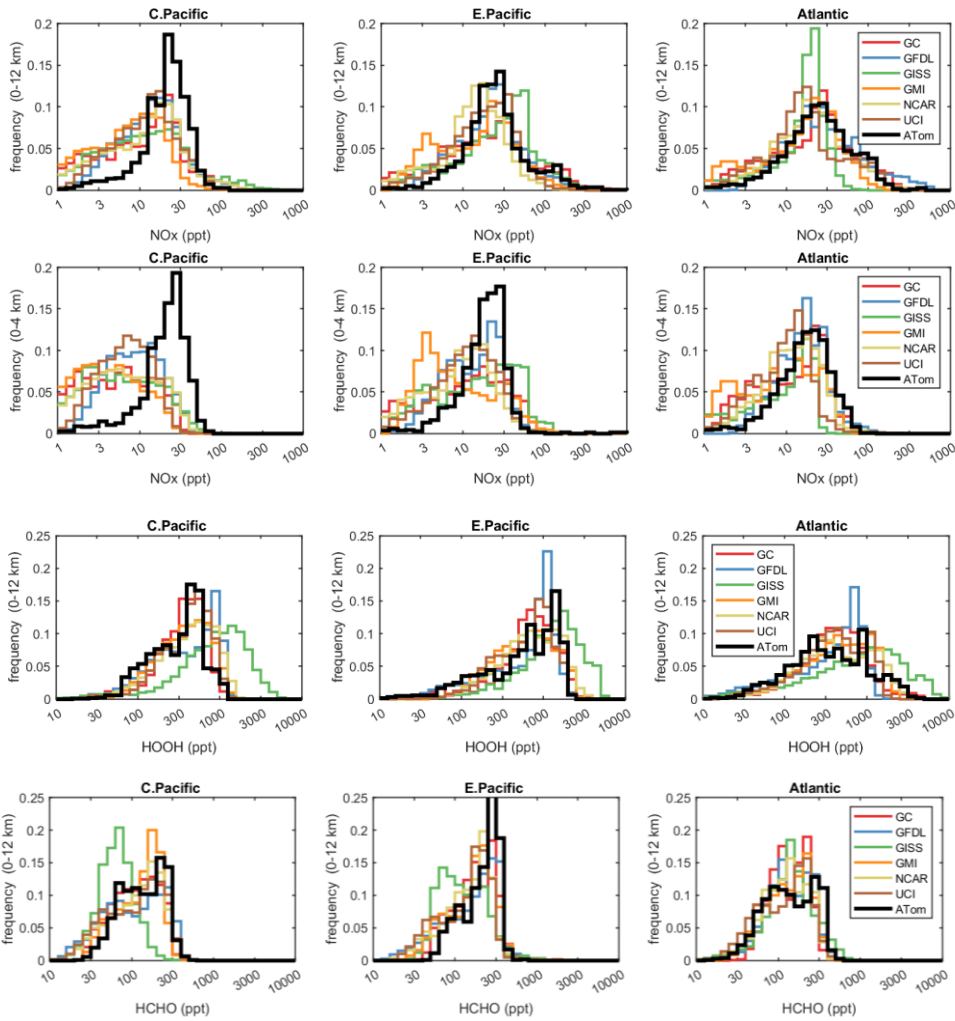


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956 **Fig. 3.** Mean altitude profiles of reactivity (rows: P-O3, L-O3, L-CH4 in ppb/day) in 3 domains
957 (columns: C. Pacific, 30° S–30° N by 180°–210° E; E. Pacific, 0°–30° N by 230°–250° E;
958 Atlantic, 30° S–30° N by 326°–343° E; ranges are the model blocks). Air parcels are
959 cosine(latitude) weighted. ATom-1 (gray) results are from Fig. 2, while model results are taken
960 from the August climatologies in Prather et al. (2017).

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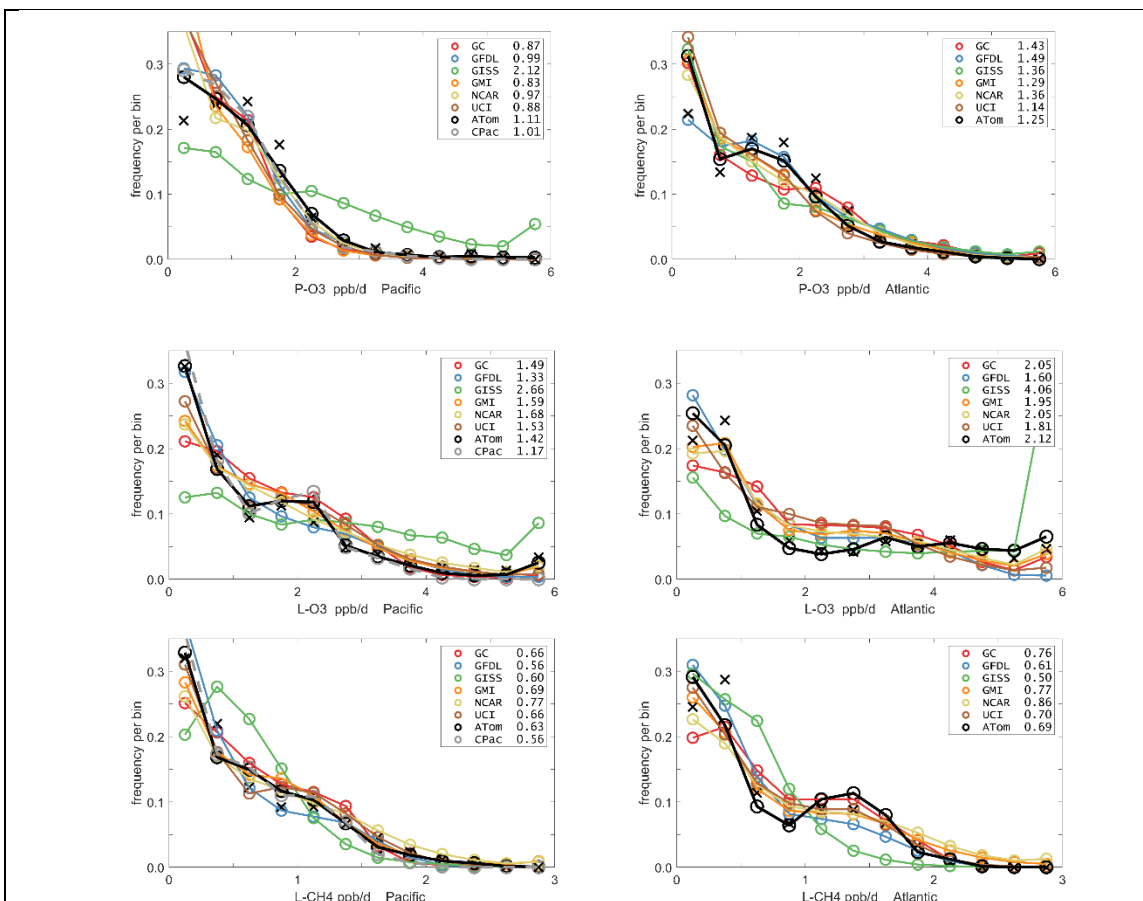
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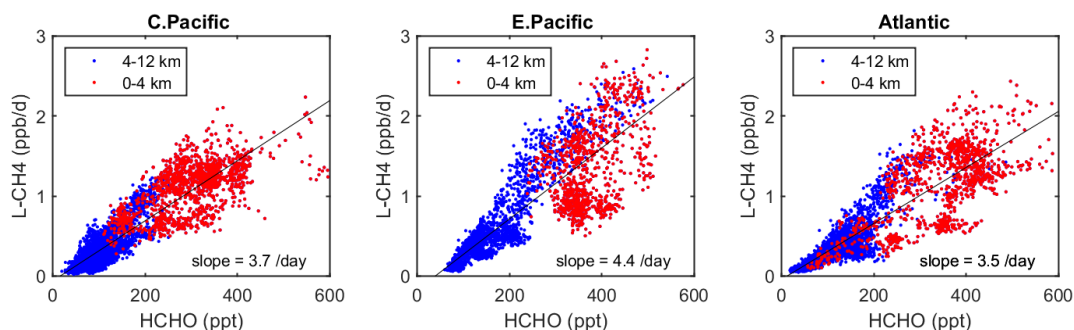
965 **Fig. 4.** Histograms of probability densities (PDs) of NO_x (0–12 km, row 1), NO_x (0–4 km, row 2),
 966 HOOH (0–12 km, row 3), and HCHO (0–12 km, row 4) for the three tropical regions (central
 967 Pacific, eastern Pacific, Atlantic). The ATom-1 data is plotted on top of the six global chemistry
 968 models' results for a day in mid-August and sampled as described in Fig. 3.

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971
 972 **Fig. 5.** Probability densities (PD, frequency of occurrence) for the ATom-1 three reactivities
 973 (rows: P-O3, L-O3, L-CH4 in ppb/day) and for the Pacific and Atlantic from 53° S to 60° N
 974 (columns left and right). Each air parcel is weighted as described in the text for equal frequency
 975 in large latitude-pressure bins, and also by cosine(latitude). The ATom statistics are from the
 976 UCIZ model, using MDS-2b and revised RDS* protocol (HNO₄ and PAN damping). The Pacific
 977 results (solid black) also show the central Pacific alone (dashed gray). The six models' values for
 978 a day in mid-August are averaged over longitude for the domains shown in Fig. S1 in the
 979 Supplement, and then cosine(latitude) weighted. Mean values (ppb/day) are shown in the legend.
 980 The PD derived from the ATom 10 s parcels binned into 1° latitude by 200 m altitude (as shown
 981 for the tropics in Fig. 2) is typical of a high-resolution global model, and denoted by black Xs.
 982



983
 984 **Fig. 6.** Scatterplot of L-CH₄ (ppb/d) versus HCHO (ppt) for ATom 1 in the 3 tropical regions
 985 shown in **Fig. 3**. The air parcels are split into lower troposphere (0–4 km pressure altitude, red
 986 dots) where most of the reactivity lies and mid+upper troposphere (4–12 km, blue). A simple
 987 linear fit to all data is shown (thin black line) and the slope is given in units of 1/day.
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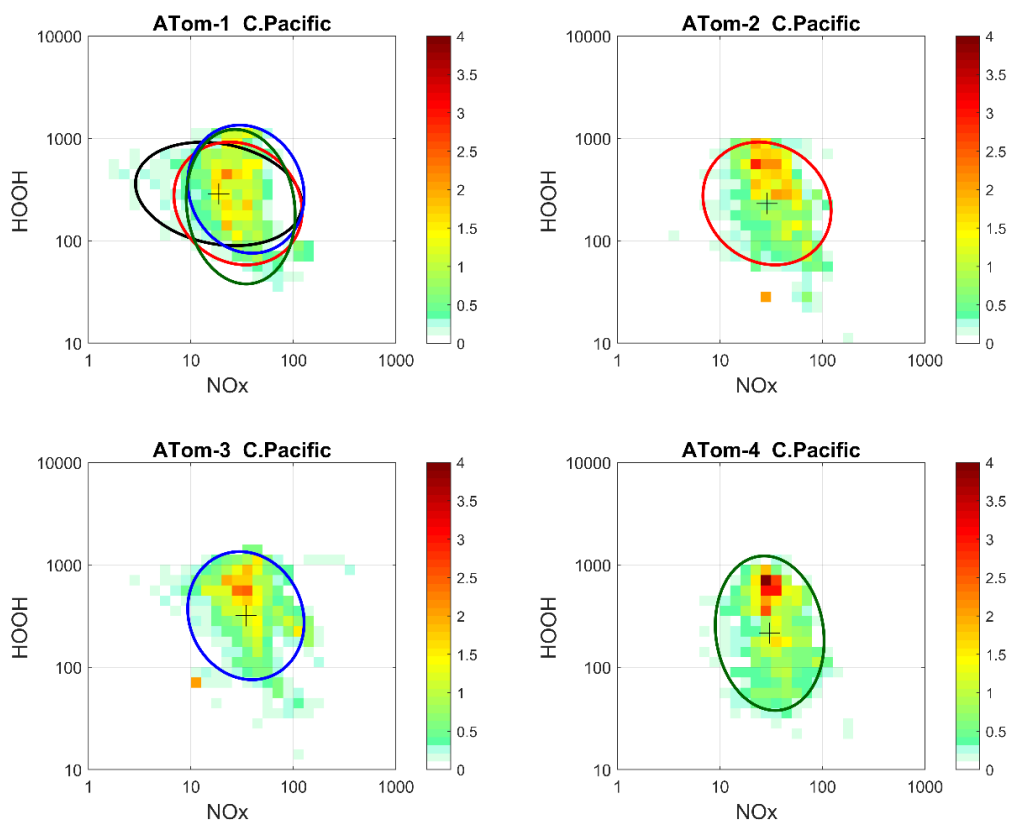


Fig. 7. 2D frequency of occurrence (PDs in log ppt mole fraction) of HOOH vs. NO_x for the tropical Central Pacific for all 4 ATom deployments. The cross marks the mean (in log space), and the ellipse is fitted to the rotated PD having the smallest semi-minor axis. The semi-minor and semi-major axes are 2 standard deviations of PD in that direction. The ellipses from ATom-2 (red), ATom-3 (blue), and ATom-4 (dark green) are also plotted in the ATom-1 quadrant.

990

Table 1. Chemistry models					
Used for	ID	Model name	Model type	Meteorology	Model Grid
clim	GFDL	GFDL-AM3	CCM	NCEP (nudged)	C180 x L48
clim, MDS-0	GISS	GISS-E2.1	CCM	Daily SSTs, nudged to MERRA	2° x 2.5° x 40L
clim, MDS-0	GMI	GMI-CTM	CTM	MERRA	1° x 1.25° x 72L
clim, MDS-0	GC	GEOS-Chem	CTM	MERRA-2	2° x 2.5° x 72L
clim, MDS-0	NCAR	CAM4-Chem	CCM	Nudged to MERRA	0.47° x 0.625° x 52L
clim, MDS-0 & 2b	UCI	UCI-CTM	CTM	ECMWF IFS Cy38r1	T159N80 x L60
MDS-0	FOAM	FOAM	box	MDS + scaled ATom Js	N/A

991 The descriptions of models used in the paper. The first column denotes if the model's August
992 climatology is used ('clim') and also the MDS versions used. FOAM used chemical mechanism
993 MCMv331 plus J-HNO₄ plus O¹D)+CH₄. For the global models see P2017, P2017, and H2018.
994

Table 2. Reactivity statistics for the three large domains (global, Pacific, Atlantic).

Value	Region	Models using MDS-0								MDS-2b
		F0AM	GC	GISS	GMI	NCAR	UCI	U15	U97	UCIZ*
P-O3, mean, ppb/d	Global	2.12	2.12	2.57	2.08	2.22	2.38	2.37	2.37	1.23
	Pacific	1.96	2.00	1.99	e1.9 6	2.01	2.17	2.13	2.15	1.11
	Atlantic	1.96	2.12	3.49	2.20	2.44	2.48	2.48	2.49	1.25
L-O3, mean, ppb/d	Global	1.81	1.63	1.93	1.70	1.76	1.76	1.74	1.75	1.61
	Pacific	1.65	1.51	1.79	1.55	1.52	1.58	1.53	1.56	1.42
	Atlantic	2.15	2.02	2.37	2.17	2.47	2.28	2.28	2.30	2.12
L-CH4, mean, ppb/d	Global	0.81	0.76	0.43	0.75	0.73	0.79	0.78	0.78	0.61
	Pacific	0.85	0.82	0.40	0.80	0.79	0.82	0.80	0.81	0.63
	Atlantic	0.80	0.78	0.51	0.81	0.86	0.85	0.85	0.85	0.69
P-O3, %sum R in top 10%	Global	35%	32%	31%	32%	30%	34%	34%	34%	33%
	Pacific	34%	28%	28%	29%	29%	30%	30%	30%	27%
	Atlantic	24%	25%	24%	26%	24%	27%	27%	28%	27%
L-O3, %sum R in top 10%	Global	35%	35%	33%	35%	36%	36%	36%	36%	36%
	Pacific	33%	32%	29%	32%	31%	32%	32%	32%	32%
	Atlantic	28%	30%	29%	30%	34%	30%	30%	30%	29%
L-CH4, %sum R in top 10%	Global	33%	30%	27%	31%	31%	32%	32%	32%	30%
	Pacific	32%	28%	26%	29%	29%	29%	29%	29%	27%
	Atlantic	27%	25%	21%	26%	27%	27%	27%	27%	25%

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Global includes all ATom-1 parcels, Pacific considers all measurements over the Pacific Ocean from 53°S to 60°N, and Atlantic uses parcels from 53° S to 60° N over the Atlantic Ocean. All parcels are weighted inversely by the number of parcels in each 10° latitude by 100 hPa bin, and by cosine(latitude). Results from MDS-0 are shown because we have results from six models. Results from the updated MDS-2b are shown (UCIZ*) using the using the current UCI CTM model UCIZ and the RDS* protocol that preprocesses the MDS-2b initializations with a 24 h decay of HNO4 and PAN according to their local thermal decomposition frequencies, see text. See additional statistics in Table S8.

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Table 3. Cross-model RMS differences (RMSDs as % of mean) for the three reactivities using MDS-0.

P-O3	F0AM	GC	GISS	GMI	NCAR	UCI
F0AM		48%	95%	45%	55%	42%
GC	48%		78%	26%	42%	32%
GISS	95%	78%		81%	72%	75%
GMI	45%	26%	81%		40%	35%
NCAR	55%	42%	72%	40%		42%
UCI	42%	32%	75%	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%	(14%)

Matrices are symmetric. Calculated with the 31,376 MDS-0 unweighted ATom-1 parcels using the standard RDS protocol. F0AM lacks 5,510 of these parcels because there are no reported J-values. UCI shows RMSD between years 2016 (default) and 1997 as the value in parentheses on diagonal. The unweighted mean R from 3 core models (GC, GMI, UCI) are: P-O3 = 1.97, L-O3 = 1.50, L-CH4 = 0.66, all ppb/d. The three core-model RMSDs with respect to one another are less than 36% and boldened.

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Table 4. ATom data files used here		
Primary Aircraft Data	Formatting and content	Comments
(a) Mor.all.at1234.2020-05-27.tbl (b) Mor.WAS.all.at1234.2020-05-27.tbl (c) Mor.TOGA.all.at1234.2020-05-27.tbl All from Wofsy et al., 2018.	(a) 149133 records x 675 csv columns, 10 s merges of flight data plus chemistry & environmental measurements (b) 6991 records x 729 csv columns, 30-120 s intervals to fill flasks (c) 12168 records x 727 csv columns, 35 s intervals of instrument	Core source of ATom measurements. irregular and difficult formatting; extremely long ascii records; large negative integers or 'NA' for some non-data.
Modeling Data Stream (MDS-2b)		
(a) ATom_MDS2b.nc	(a) netcdf file containing regularly spaced 10 s observations for ATom-1 (32383 records), ATom-2 (33424 records), ATom-3 (40176 records), ATom-4 (40511 records), 146,494 in total; includes physical flight data (11), chemical data (39), miscellaneous data including corrected HNO ₄ and PAN (6), flag data (50).	Regular formatting; all data gap filled with flags to identify the method and extent of filling; NaN's only for flight 46; for use in modeling of the chemistry and related statistics from the ATom 10 s data.
Reactivity Data Stream (RDS-2b)		
(a) ATom_RDS2b.nc	(a) netcdf file containing regularly spaced reactivities for 10 s parcels from ATom-1234 (146,494 in total); includes latitude, longitude and pressure of model grid cell used in the calculation; includes P-O ₃ , L-O ₃ , L-CH ₄ , L-CO, J-O ₁ D, plus dO ₃ /dt = net O ₃ change over 24 h. Reactivities are given for 5 days separated by 5 days in the middle of each deployment, plus the 5-day mean.	Results from newest UCI CTM version (UCIZ) run with RDS* protocol (PAN and HNO ₄ decay) and using MDS-2b. NaN's only for flight 46.