

Global Observations of Acetyl Peroxynitrate (PAN) in the Remote Troposphere

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Abstract

We present global airborne observations of acetyl peroxy (CH₃C(O)OO) radicals reacting with NO₂ to form acetyl peroxy (PA) radicals, PAN) in the remote troposphere from the Atmospheric Tomography (ATom) campaign. These observations show that biomass burning is the dominant source of PAN in the Southern Hemisphere. In the Northern Hemisphere, anthropogenic emissions from Asia and Europe, also contribute significantly to PAN over the Pacific and Atlantic oceans. Model simulations underestimate PAN in the lower troposphere, in part, due to the underestimation of local production driven by acetaldehyde oxidation and β_{NO_2} (the ratio of acetyl peroxy (PA) radicals reacting with NO₂ relative to other pathways). The significant impacts of biomass burning evident in the ATom PAN observations suggest that improving model treatment of plume transport and the conversion of NO_x to PAN in biomass burning plumes is a viable focus for better simulating PAN. Global observations of PAN provide a benchmark for the evaluation of satellite observations and model simulations of PAN.

Plain Language Summary

We report the global airborne observations of acetyl peroxy (PAN) from the Atmospheric Tomography (ATom) campaign. Biomass burning is the dominant source of PAN over the remote oceans in the Southern Hemisphere. Anthropogenic emissions from Asia and Europe significantly contribute to PAN levels over the Pacific and Atlantic oceans, alongside biomass burning. Model simulations underestimate PAN in the lower oceanic troposphere, partly due to an underestimation in local production driven by acetaldehyde. Our observations show strong evidence of the persistent of biomass burning influence on PAN throughout the remote troposphere, particularly over the Southern Hemispheric oceans. This work provides a valuable data set to improve understanding the global distribution of PAN.

Key Points

- Global airborne observations of PAN over the remote troposphere during the ATom campaign are presented.
- Biomass burning emissions are the dominant source of PAN over the Southern Hemispheric oceans, while in the Northern Hemisphere, anthropogenic emissions from Asia and Europe significantly contribute to the remote tropospheric PAN levels alongside biomass burning.
- Model simulations underestimate the local production of PAN in the lower oceanic troposphere, driven by acetaldehyde, as well as the impact of biomass burning on PAN levels throughout the remote troposphere.

1 Introduction

Acetyl peroxyxynitrate (PAN) is an important reservoir for nitrogen oxides (NO_x). In many regions of the troposphere, PAN is the primary form of reactive nitrogen (NO_y) (Roberts et al., 2007). PAN can undergo long-range transport as it is insoluble and quite stable at cold temperatures (lifetime of PAN, τ_{PAN} , up to a month in the upper troposphere) (Singh & Hanst, 1981; Talukdar et al., 1995; Frenzel et al., 2000). Transport and subsequent thermolysis of PAN to release NO_x may be essential for photochemical ozone (O_3) production in the remote troposphere (Kasibhatla et al., 1993; Moxim et al., 1996; Fischer et al., 2014).

Despite significant advances in understanding the distribution of PAN on a global scale, large uncertainties persist in the remote troposphere. Previous studies attributed pollution episodes in remote marine environments, where PAN is a major component of NO_y , to continental transport from urban areas, biomass burning and lightning- NO_x emissions (e.g., Wahner et al., 1994; Singh et al., 1996, 2006 and 2009; Blake et al., 1999; Roberts et al., 2004; Fischer et al., 2014). However, the global impact of these sources, particularly in remote regions like the Southern Hemisphere oceans, remains poorly characterized. Observations in these regions are scarce but are critical for diagnosing the key factors controlling the global distribution of PAN including changing impacts of continental emissions, transport, and the atmospheric chemistry of pollutants transported to the remote troposphere (Roberts et al., 2007). Recent studies from the NASA Atmospheric Tomography (ATom) campaign, considered here, emphasize gaps in understanding remote tropospheric composition. These include missing sources of primary PAN precursors such as acetaldehyde and NO_x over remote oceans (Wang et al., 2019; Shah et al., 2023), and the significant contribution of biomass burning to the global distribution of O_3 , peroxides and particles (Bourgeois et al., 2021; Allen et al., 2022; Schill et al., 2020). Observations of PAN in the remote troposphere, in combination with satellite observations and model simulations, are also critical for addressing gaps in our understanding.

In this work, we present global-scale airborne PAN observations with high temporal resolution and extensive spatial coverage from the ATom campaign (Thompson et al., 2021). These observations provide "background" levels of PAN and its precursors in regions previously lacking extensive measurements, achieved through near-continuous vertical sampling and flight strategies designed for an unbiased assessment of the remote troposphere. We report mixing ratios of PAN over the remote Pacific and Atlantic Oceans in all seasons. We examine the distribution of PAN in the remote troposphere by a combined analysis of observations and model simulations including emission tracer measurements, back-trajectory analyses, and tagged emission tracer simulations. A diagnostic evaluation of both PAN and its precursor simulations from a Chemical Transport Model

75 (CTM) was performed using observations and kinetic calculations highly constrained by measured parameters.
76 Lastly, we utilize measurements of biomass burning tracers and simulated biomass burning contributions to carbon
77 monoxide (CO) to investigate the impact of biomass burning on the relationship between observed and simulated
78 PAN.

79 **2 Methods**

80 A comprehensive suite of chemical, physical and meteorological measurements was conducted aboard the
81 NASA DC-8 research aircraft during the four ATom deployments from 28 July 2016 (ATom-1) to 21 May 2018
82 (ATom-4). Each deployment, lasting for approximately one month in each season, followed near-identical flight
83 plans as described by Figure S1a and Thompson et al. (2021). Instrumentation details are provided in Table S1. Data
84 were merged to a 10-second time base or the Trace Organic Gas Analyzer (TOGA) sampling interval (35 seconds
85 every two minutes) and filtered to exclude data obtained over continental regions. The geographical boundaries used
86 to filter the data are illustrated along with the flight tracks in Figure S1b. We removed data with stratospheric
87 influence defined as when O_3 and O_3/CO were greater than 100 ppbv and 1 ppbv ppbv⁻¹, respectively, at altitudes
88 greater than 6 km. A marked decrease in CFC-113 provided an additional proxy for stratospheric influence,
89 employing the 25th-percentile value (70–72 pptv) across each deployment as a cut-off. This effectively excluded
90 data obtained at the edges of stratospheric-influenced samples that often have O_3 levels lower than 100 ppbv (Figure
91 S2). In addition, data in which the photolysis rate of J_{NO_2} was below its 10th percentile (solar zenith angle $\sim 90^\circ$)
92 were excluded in kinetic calculations.

93 Alongside extensive observations, the ATom dataset includes model products and trajectory calculations.
94 To investigate the potential impact of transpacific transport on the observed trace gas composition, 10-day back
95 trajectory ensembles, which were initialized at each minute along the ATom flight tracks, were utilized (Ray, 2021).
96 Data points whose back trajectories intersected the boundaries of East Asia, as shown in Figure S1b, were identified
97 as having "East Asian influence (EAI)." As transpacific transport is a complex process unlikely to be fully
98 represented using a single method, we assessed systematic differences in measured chemical tracers such as carbon
99 monoxide (CO), hydrogen cyanide (HCN), propionyl peroxyoxynitrate (PPN) and PAN, for data with and without EAI.
100 This analysis is facilitated by incorporating the Goddard Earth Observing System Version 5 (GEOS-5) simulations
101 of tagged-CO concentrations, which well predicted the distribution, source, and episodic enhancements of CO
102 during ATom-1 (Strode et al., 2018; Newman & Pawson, 2021). Our method to identify EAI is described in section
103 4 of the supporting information (SI) using data from the 6 October 2017 research flight, which sampled both East
104 Asian transport and Southern Hemisphere biomass burning events.

105 A key focus of the ATom measurements is to provide observational constraints to evaluate model-
106 simulated remote tropospheric composition, previously hindered by a lack of relevant observations in such regions.
107 For this purpose, we utilized outputs from the Global Modeling Initiative (GMI)-CTM (Strode et al., 2021). The
108 performance of different CTMs with various meteorological and chemical schemes has been discussed elsewhere
109 (e.g., Travis et al., 2020; Shah et al., 2022). Briefly, the GMI-CTM provides trace gas mixing ratios with a spatial
110 resolution of 1° latitude \times 1.25° longitude \times 72 vertical levels and a temporal resolution of 15 minutes, interpolated
111 to the ATom flight tracks (Strode et al., 2021). The same criteria used to filter observational data were applied to the

112 model outputs. For a diagnostic analysis of model simulated PAN, we compared the model outputs with the
113 observed hydrocarbon and radical precursors of PAN, as well as the estimated acetyl peroxy (PA) radical
114 concentrations. The impact of observation-model discrepancies in these variables on PAN simulations are
115 investigated by comparing the simulated PAN to the observed data and steady state (SS) calculation for PAN when
116 temperatures were above 290 K, the median value observed in the lower troposphere (<2 km) during ATom. This
117 allows identification of data in which the SS estimation for PAN is most valid. The details in the calculations and
118 estimated uncertainties are described in the SI.

119 3 Results

120 3.1 PAN mixing ratios

121 Figure 1a illustrates the vertical profiles of median PAN mixing ratios in 1-km altitude bins over the tropics
122 (0° – 20°), mid-latitudes (20° – 60°), and polar regions (60° – 90°) in the Northern Hemisphere (NH) and Southern
123 Hemisphere (SH) in each season. In the NH lower troposphere (<2 km), PAN mixing ratios were generally higher in
124 the extratropics than in the tropics. A notable exception was observed in July–August, where the highest PAN levels
125 (median of 20 pptv) were measured in the tropical lower troposphere. This coincided with the highest PAN
126 production rate (P_{PAN} , median of 13 pptv hr^{-1}) and the shortest $\tau_{\text{PAN}} \sim 0.8$ hours, suggesting efficient local production
127 of PAN over the tropical oceans in this case. The latitudinal differences in vertical PAN mixing ratios are
128 particularly noticeable in April–May, when PAN in the extratropics was higher than that in the tropics at all
129 altitudes. Springtime enhancements of PAN in the NH remote extratropics have been attributed to the transport of
130 continental outflow when photochemical production of PAN is efficient and its thermal decomposition is relatively
131 slow (Penkett & Brice, 1986; Bottenheim et al., 1994).

132 On average, PAN in the NH increased with altitude from the low to mid-troposphere but decreased above 8
133 km, in part due to loss by photolysis in the upper troposphere. In the SH, PAN also leveled off in the upper
134 troposphere. Increasing PAN levels with altitude were associated with the increasing PAN lifetime and concurrent
135 enhancements of combustion tracers such as CO and HCN. The contribution of local P_{PAN} , which was dominated by
136 acetaldehyde oxidation, to this vertical distribution was minimal (Figure S5). Net P_{PAN} (P_{PAN} -loss rate of PAN, L_{PAN})
137 were below 1 ppt hr^{-1} in the free troposphere over both the Pacific and Atlantic Oceans. In July–August, high PAN
138 levels persisted above 8 km in all latitude bins. The elevated PAN was coincident with HCN up to 680 pptv and a
139 high probability of convective influence (a factor of 3.5 greater than the annual median in the upper troposphere),
140 which was estimated based on the back trajectories and satellite detection of cloud locations as used by Allen et al.
141 (2022) to examine global peroxide distributions during ATom. This suggests that transport of biomass burning
142 plumes coupled with convection is critical to explain the NH PAN distribution in summer.

143 Unlike the complicated latitudinal distribution of PAN in the NH, PAN in the SH increased almost linearly
144 at a rate of roughly 0.7 pptv per degree latitude from the South Pole to the tropics on average. In this work, the
145 latitudinal gradient of PAN in the SH is attributed to increasing biomass burning impacts from Africa, South
146 America, and Australia over the SH tropics. Median PAN mixing ratios in the mid- and upper troposphere were
147 significantly higher in austral winter (July–August) and spring (September–October), the dry seasons with frequent
148 burning events (Andreae et al., 2012; Paton-Walsh et al., 2022). The highest median value of PAN was observed at

149 6–8 km during austral spring. This springtime PAN maximum in the SH agrees with previous studies including
150 recent satellite observations (Fischer et al., 2014; Pope et al., 2016). On the basis of strong correlations of PAN with
151 biomass burning tracers (section 3.2), we attribute the austral spring PAN maximum as well as most of the SH PAN
152 variations to biomass burning.

153 Figures 1b and 1c shows median PAN mixing ratios and PAN to NO_y ratios in 4-degree latitude and 1-km
154 altitude bins, respectively. Elevated levels of PAN and its contribution to NO_y were observed throughout the NH
155 extratropical troposphere. PAN remained a major component of NO_y in both the NH and SH extratropical free
156 troposphere but was depleted in the tropical lower troposphere. This is consistent with previous remote surface
157 measurements of PAN (e.g., Rudolph et al., 1987; Müller & Rudolph, 1992; Roberts et al., 2007 and references
158 therein). Our work attributes low PAN levels in the tropics to a net negative PAN production rate (median of P-L = -
159 0.6 pptv hr^{-1}) driven by thermal decomposition. Over the tropical Pacific Ocean, the patterns in latitude-altitude
160 cross sections of PAN were consistent with those in L_{PAN} (Figures 1b and S5d). Despite the similarity of L_{PAN} over
161 the Atlantic Ocean to the global trend, PAN was more than 30% of NO_y throughout the tropical Atlantic
162 troposphere. This is consistent with strong enhancements of HCN ($>200 \text{ pptv}$) and CO ($>100 \text{ ppbv}$) throughout the
163 subtropical Atlantic troposphere (Figure S6a-b), indicating the influence of biomass burning on PAN and its
164 contribution to NO_y in the regions. Consequently, PAN and PAN/ NO_y over the Atlantic Ocean were more consistent
165 across latitudes and altitudes in contrast to the significant variations observed between the NH and SH Pacific
166 Oceans.

167 3.2 Sources of PAN

168 We investigate the influence of continental sources on PAN in the remote mid- and upper troposphere
169 (altitudes $> 4 \text{ km}$), where PAN has a relatively long lifetime (up to three weeks). We utilize chemical tracers of
170 specific pollution sources and the categorizations of data based on back-trajectory analysis. Namely, we used high
171 temporal resolution measurements of HCN and CO from biomass burning and general combustion emissions,
172 respectively. (Holzinger et al., 1999). PPN was used as a photochemical tracer of anthropogenic emissions (Roberts
173 et al., 2007; Lee et al., 2021). Note that measurements of PPN were not available for July–August (ATom-1).

174 Figure 2 illustrates the seasonal trends of PAN along with those of chemical tracers using box plots. The
175 marked increases in PAN during the SH fire seasons (September–October) were consistent with those of HCN and
176 CO for both the SH Pacific and Atlantic Oceans (Figure 2a-b). Despite the wide geographic coverage of the data, a
177 strong correlation ($r^2 > 0.7$) between PAN and HCN was notable over the SH Oceans (Figure 2c), highlighting the
178 significant contribution of biomass burning to PAN in these regions. This chemical tracer analysis was further
179 facilitated by TOGA measurements of two anthropogenic tracers, C_2Cl_4 and CH_2Cl_2 , and a biomass burning tracer
180 CH_3CN (Apel et al., 2015; Roozitalab et al., 2024). The analysis using TOGA measurements agreed well with
181 analysis that used the faster time response measurements. The correlations of PAN, CO, and PPN with the biomass
182 burning tracers (HCN and CH_3CN) were significantly higher than those with the anthropogenic tracers measured by
183 TOGA. For this reason, this work suggests that biomass burning is the primary contributor to PAN and pollutant
184 distribution in the SH.

185 The seasonal trends of PAN were more comparable to those of HCN than those of CO for both the NH
186 Pacific and Atlantic Oceans (Figure 2d–e). PAN showed moderate correlations with HCN over the NH Pacific and
187 Atlantic Oceans ($r^2 = 0.5$ and 0.3 , respectively), indicating a notable contribution of biomass burning transport to
188 PAN in the NH (Figure 2f). Over the NH Pacific Ocean, the correlations of PAN and CO with C_2Cl_4 and CH_2Cl_2
189 were similar to those with HCN, suggesting that anthropogenic emissions are also important contributors to PAN
190 over this region. PPN, derived largely from anthropogenic precursors, showed moderate correlations with
191 anthropogenic (r^2 of 0.4 and 0.3 with C_2Cl_4 and CH_2Cl_2 , respectively) and burning ($r^2 > 0.3$ with CH_3CN) tracers
192 only over the NH Pacific Ocean and SH Oceans, respectively. While PAN and PPN showed a negligible relationship
193 with anthropogenic tracers including CO over the NH Atlantic Ocean, increasing values of PAN mixing ratios were
194 often coincident with elevations in the observed propane mixing ratios and the fractional contribution of CO from
195 European non-biomass burning source ($CO_{non-BB\ EU}$) in GEOS-5 simulations in the extratropical regions (latitude $>$
196 30°) (Figure S7). The highest levels of PAN and propane were observed up to 400 pptv and 1.3 ppbv, respectively,
197 along with greater than 60% combined contribution of $CO_{non-BB\ EU}$ and CO from Asian non-biomass burning sources
198 ($CO_{non-BB\ Asia}$) in GEOS-5. Consequently, this suggests that enhancements of PAN in the NH Atlantic extratropical
199 regions may be more associated with anthropogenic emissions from Europe and Asia than from North America.

200 The observed relationships between PAN and chemical tracers point out the important contribution of both
201 anthropogenic and biomass burning emissions to PAN levels over the NH Pacific Ocean. An important
202 anthropogenic pollution source in the NH Pacific Ocean is transpacific transport from East Asia (Wang et al., 2003;
203 Roberts et al., 2004), with its strongest influence in spring (Stohl et al., 2002; Liu et al., 2005). Figure 3 shows
204 cumulative distribution functions (CDFs) of PAN and related tracers with and without EAI, determined from the
205 back trajectories, using mid- and upper tropospheric data for the NH. On the basis of the CDFs in Figure 2g, data
206 with EAI contained elevated levels of PAN and tracers, with mixing ratios of PAN, CO, HCN, and PPN being 2.7 ,
207 1.3 , 1.3 and 2.3 times higher, respectively, compared to those without EAI. The elevated levels of CO are consistent
208 with a 26% higher fractional contribution of $CO_{non-BB\ Asia}$ to the total CO in GEOS-5 simulations. Similar systematic
209 differences in PAN levels have been reported by previous observations. For example, airborne measurements during
210 ITCT 2K2 (Intercontinental Transport and Chemical Transformation 2002) campaign observed up to 650 pptv of
211 PAN accompanied with CO mixing ratios above 150 ppbv (Roberts et al., 2004; Nowak et al., 2004). In accordance
212 with these findings, Figure 2h illustrates data with EAI contains higher levels of both PAN and CO, along with
213 larger slopes and r^2 for linear regressions between PAN and CO, as well as the higher fractional contribution of
214 $CO_{non-BB\ Asia}$. The median mixing ratios of PAN and CO with significant EAI peaked in spring with a value $\sim 70\%$
215 greater than the annual median, whereas those without EAI peaked in summer. This seasonal trend of PAN also
216 agrees with the stronger impacts of transpacific transport on pollution over the Pacific Ocean during springtime
217 (Jaffe et al., 1999; Zhang et al., 2008). In summary, this work suggests that PAN over the NH is largely influenced
218 by complicated sources such as biomass burning and anthropogenic emissions including transpacific transport,
219 whereas PAN over the SH can be largely accounted for by biomass burning emissions.

220 3.3 Diagnostic evaluation of modeled PAN using observations

221 Figure 3a illustrates the probability distribution function (PDF) of observed and modeled acetaldehyde and
222 β_{NO_2} , where β_{NO_2} represents the ratio of PA reacting with NO_2 to the total reacting with NO , NO_2 , HO_2 and RO_2 .
223 When PAN approaches steady state (SS) due to rapid thermal decomposition over the remote lower troposphere, its
224 concentration depends linearly on the production rate of PA, primarily driven by the oxidation of acetaldehyde by
225 the OH radical (e.g., Figure S5), but inversely on the product of the thermal decomposition rate and $1/\beta_{NO_2} - 1$
226 (Sillman et al., 1990). On average, the model underestimated both acetaldehyde and β_{NO_2} when using steady-state
227 NO_2 (β_{SS-NO_2}). In this section, NO_2 from pseudo-steady state (SS- NO_2) calculations is used due to potential
228 measurement interferences from thermally labile species and larger discrepancy between the observed NO_2 and the
229 estimated values based on SS calculations and model simulations at higher altitudes (Nault et al., 2015; Shah et al.,
230 2023). The model underestimation of acetaldehyde was persistent across altitudes, in accordance with the findings of
231 Wang et al. (2019) during ATom 1–2. In contrast, the model underestimated β_{NO_2} more significantly in the lower
232 troposphere, which is consistent with the underestimation in modeled NO_x using both GMI and GEOS-Chem
233 without photolysis of particulate nitrate at altitudes below 6 km, as discussed by Shah et al. (2023).

234 Figure 3b compares PA concentrations directly from the 3-D model output and those derived from SS
235 calculations using modeled and observed constraints. To be consistent with SS calculations for PAN (Figure 3c),
236 only data with temperatures >290 K were included. On average, PA concentrations from both the model output and
237 SS calculations using modeled PAN precursors were approximately 90% lower than the estimated values based on
238 observations and equations in the SI (SE1-2). Improved agreement between the modeled and observed PA was
239 found when the relative difference between modeled and observed acetaldehyde was lower.

240 The remote lower tropospheric conditions observed during ATom facilitate the diagnosis of the model-
241 simulated PAN by comparison with SS calculations of PAN. In these conditions, the short τ_{PAN} , persistent
242 acetaldehyde levels and limited perturbation from transport improve validity of the SS calculations (Cleary et al.,
243 2007, SI section 4 and Figures S8 and 9). Figure 3c illustrates that observed PAN is in better agreement with the SS
244 estimation than the 3-D model. The agreement between the modeled and observed PAN improves when β_{NO_2} and
245 PA from the 3-D model and observations are consistent. Consequently, this suggests that model simulations of PAN
246 in the remote lower troposphere can be improved by addressing the model bias in acetaldehyde and β_{NO_2} .

247 Biomass burning emissions had significant impact on the remote tropospheric PAN distribution during
248 ATom. In particular, strong linear relationships between PAN and HCN were notable in the SH. Figure S10
249 illustrates this linear relationship across altitude bins for both the NH and SH observations. To investigate the impact
250 of biomass burning on the observation-model discrepancies in PAN, we focused on the percent difference of PAN
251 between observations and GMI model simulations in the SH due to the relatively simple emission source
252 characteristics dominated by biomass burning. Figure 3d illustrates the percent difference of PAN increased with
253 fractional contribution of CO from global biomass burning in GEOS-5, which is consistent with increasing percent
254 difference of PAN with observed HCN levels (Figure S10). The increase in the percent difference of PAN was
255 found while CO from GEOS-5 agrees with both CO from GMI and observations within 30%. For this reason, we
256 suggest that model simulations of the global PAN distribution will depend on accurate treatment of chemical
257 evolution in biomass burning plumes as well as emissions and transport. Measurements of PAN in the SH, presented

258 in this work, may be useful to evaluate the simulated influence of biomass burning on PAN due to the dominant
259 contribution of biomass burning to PAN distributions in these regions.

260 **4 Discussion**

261 This work presents observations of the remote tropospheric PAN distribution. Biomass burning has a large
262 and persistent impact on PAN across the globe, and is a dominant source of PAN in the SH. The significant global
263 effects of biomass burning on PAN are likely due to the injection of its precursors in biomass burning plumes that
264 are often transported at higher altitudes, where thermolysis is slower and wind speeds are higher (e.g., Ke et al.,
265 2021). Global satellite observations have attributed a significant portion of upper tropospheric and lower
266 stratospheric PAN to biomass burning with pyro-convective transport, which is in accordance with our findings.
267 (Moore & Remedios, 2010; Tereszchuk et al., 2013; Pope et al., 2016). Notably, the agreement between modeled
268 and observed vertical distributions of PAN and CO was greatly improved for North American boreal regions with
269 the implementation of a new fire plume injection scheme based on satellite data into GEOS-Chem (Zhu et al., 2018).

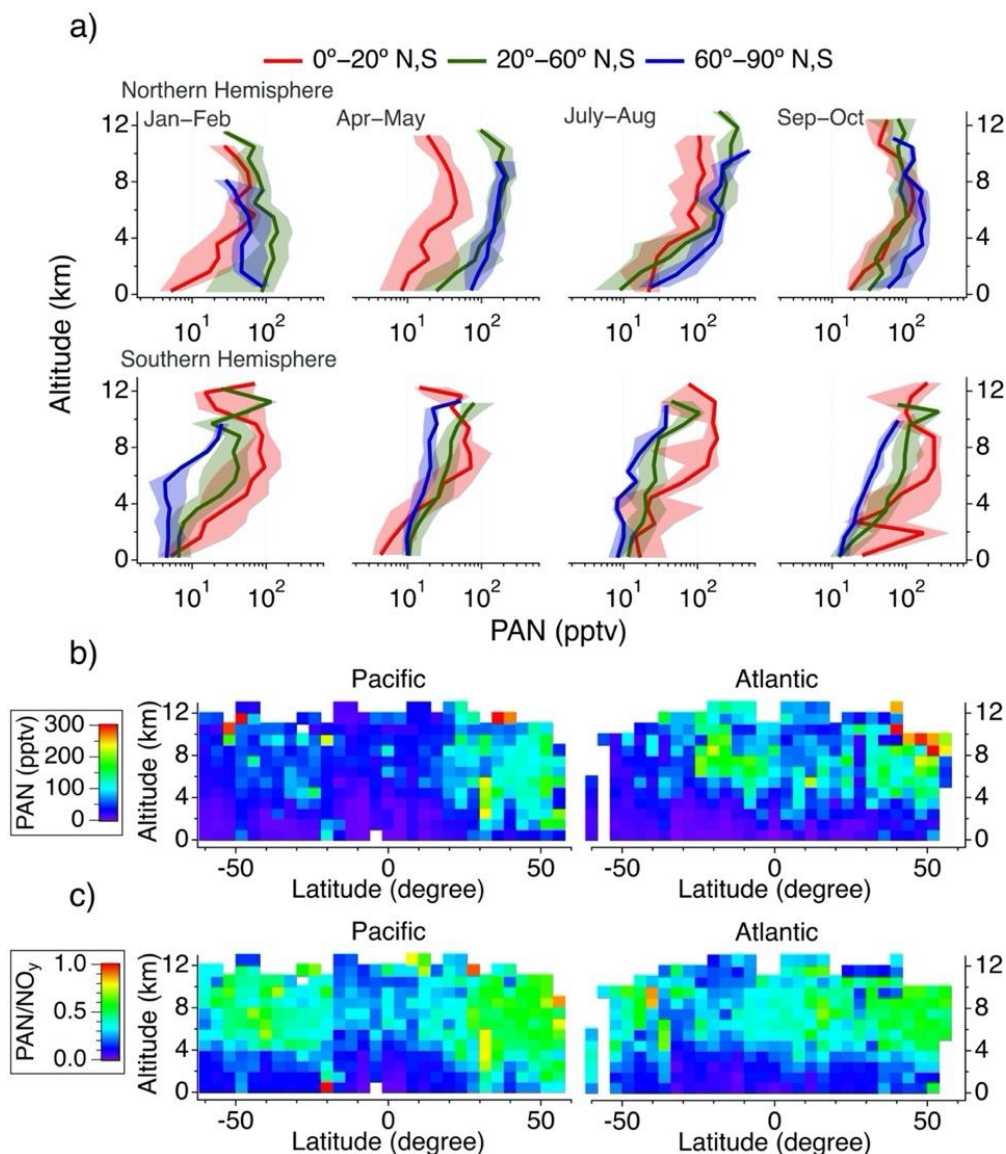
270 Efficient PAN production is commonly observed in biomass burning plumes (Alvarado et al., 2010; Liu et
271 al., 2016; Wolfe et al., 2022), but simulating this process using a CTM has proven challenging due to complex
272 factors such as emissions, plume injection, mixing, and photochemistry (Andreae, 2019; Liu et al., 2016; Wang et
273 al., 2021). Based on our recent PAN measurements during the 2019 NASA/NOAA FIREX-AQ (Fire Influence on
274 Regional to Global Environments of Air Quality) campaign, PAN often constitutes 40–50% of NO_y a few hours
275 downwind of biomass burning, which is consistent with findings from previous observations (Alvarado et al., 2010;
276 Liu et al., 2016). The relatively consistent contribution of observed PAN to NO_y may result from a combined effect
277 of these factors. While a detailed assessment of PAN chemistry is crucial for understanding the chemical evolution
278 of biomass burning emissions, assigning ~40% of biomass burning NO_x emissions to PAN seems reasonable for a 3-
279 D CTM focusing on the global distribution of PAN (e.g., Alvarado et al., 2010; Fischer et al., 2014). However,
280 different fire types (e.g., wildfire and prescribed burning) exhibit markedly different conversion rates of NO_x to
281 PAN (e.g., Liu et al., 2016, 2017). This suggests that implementation of different conversion rates to attribute initial
282 NO_x emissions to PAN based on fire types could be tested as a more effective modeling strategy. Recently, retrieval
283 products from the TROPospheric Monitoring Instrument (TROPOMI) were used by Jin et al. (2021) to estimate
284 global biomass burning NO_x emission factors and lifetimes for different fuel types. Their satellite-based NO_2
285 analysis illustrated decreasing NO_x lifetime with fire emissions due to changing chemical regimes (from NO_x limited
286 to saturated regimes). However, the ability to understand the impacts of PAN on the evolution of NO_x using an
287 exponentially modified Gaussian analysis on TROPOMI NO_2 columns was limited, highlighting the complexity of
288 NO_y processing in fire plumes. Consequently, our work suggests that the model treatments of plume transport and
289 conversion of biomass burning NO_x emissions to PAN are an important focus for reliably simulating the impacts of
290 biomass burning on global PAN distributions.

291 **5 Conclusion**

292 This work reports global-scale airborne observations of remote tropospheric PAN during the ATom
293 campaign. PAN concentrations over the NH oceans were impacted by a complex mix of sources such as

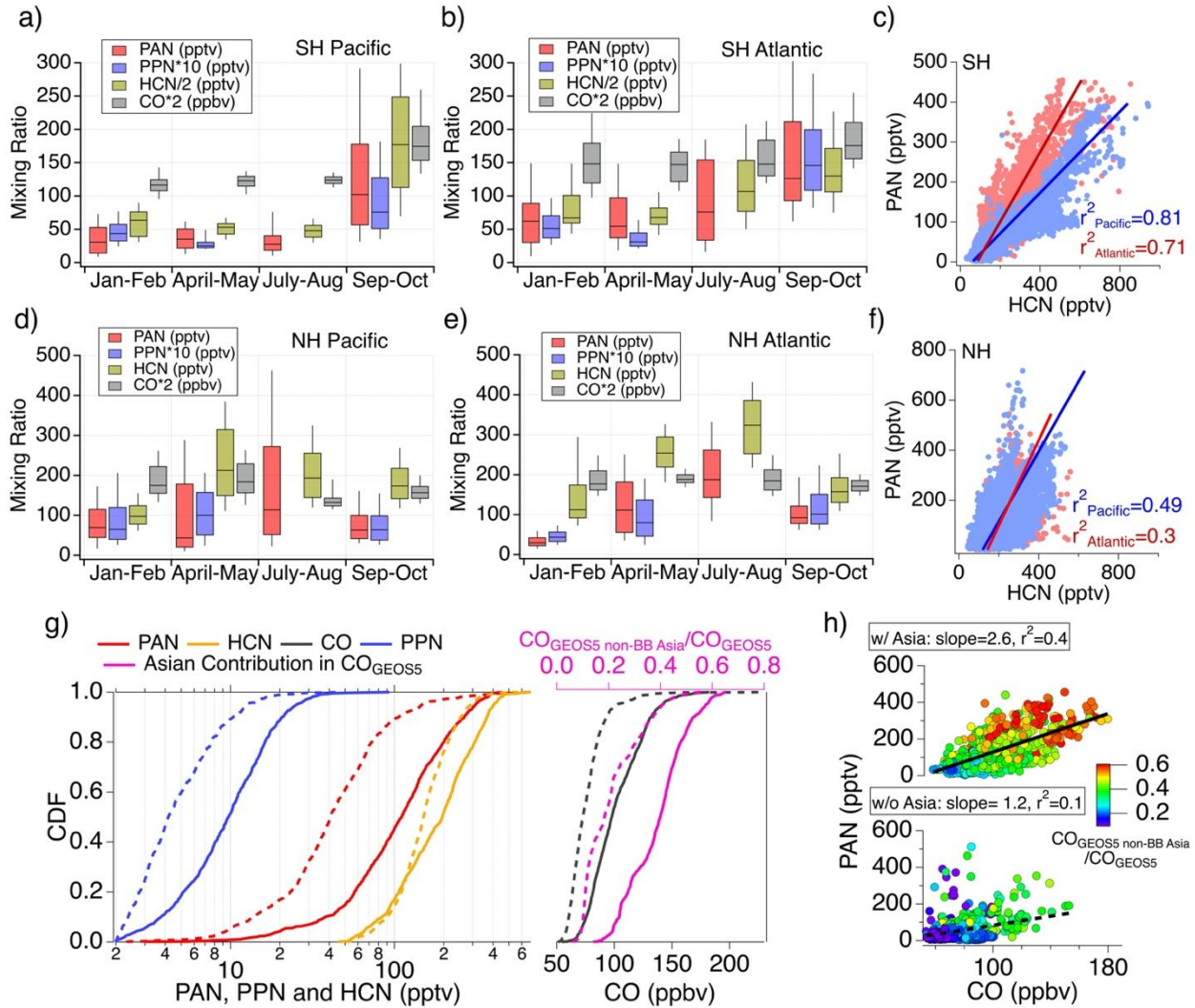
294 anthropogenic emissions, transpacific transport from East Asia, and biomass burning. Biomass burning has a
 295 dominant and persistent impact on PAN in the SH. Our diagnostic evaluation of modeled PAN and its precursors
 296 based on observations suggests the GMI model should account for the underestimations of acetaldehyde and β_{NO_2} to
 297 improve PAN simulations in the lower troposphere. In addition, we suggest that accurate treatment of biomass
 298 burning should be a focus for model simulations of PAN in the remote troposphere. The remote tropospheric PAN
 299 observations in this work provide a valuable data set for model evaluations and the analyses and evaluation of
 300 satellite observations.

301 **Figures**



302
 303 **Figure 1.** (a) Vertical profiles of median PAN mixing ratios in 1-km altitude bins over the tropics (red), mid-
 304 latitudes (green), and polar regions (blue) of the NH (top panels) and SH (bottom panels), with shaded areas
 305 indicating 25th and 75th percentiles. Altitude plots of (b) PAN and (c) PAN to NO_y ratios as a function of latitude.

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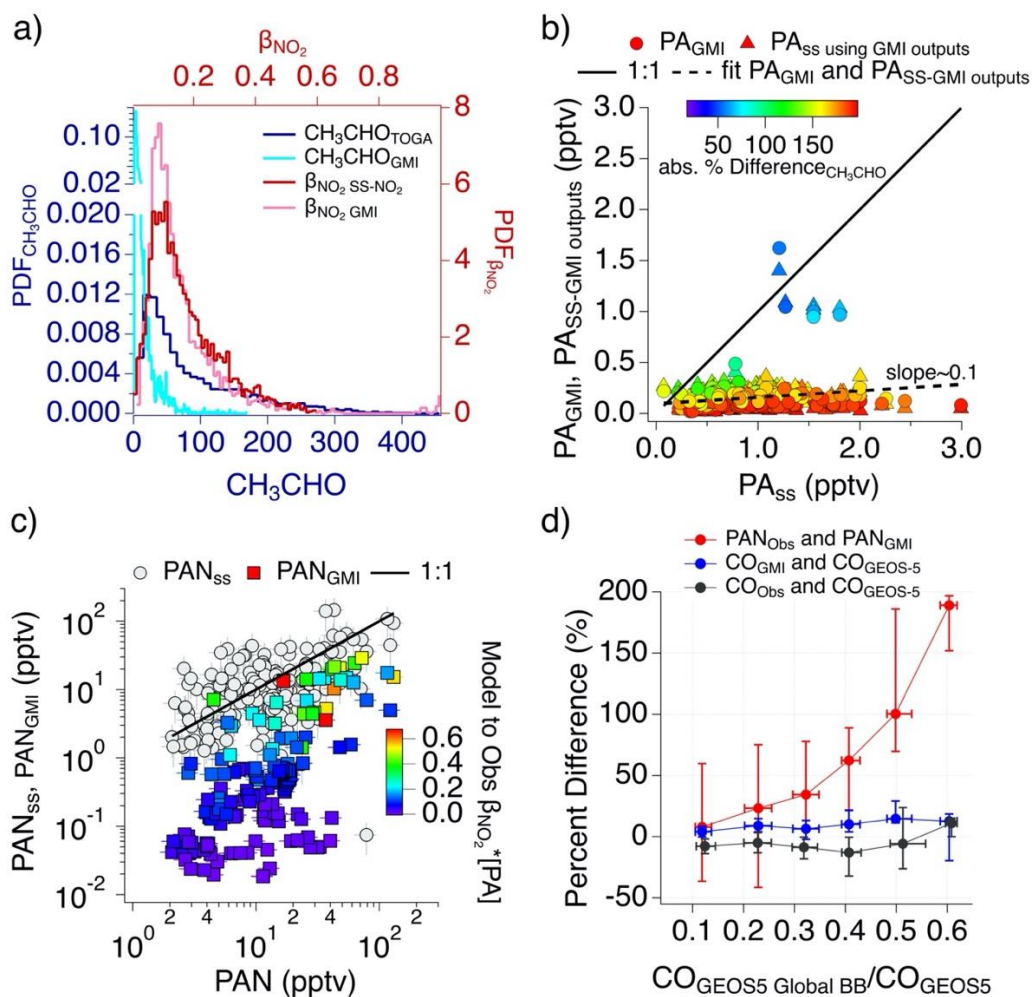
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308 **Figure 2.** Box and whisker plots of selected tracers observed over the SH (a) Pacific and (b) Atlantic, and (c) scatter
 309 plot of observed PAN versus HCN for the SH oceans in all four seasons. Box and whisker plots of selected tracers
 310 observed over the NH (d) Pacific and (e) Atlantic, and (f) scatter plot of observed PAN versus HCN for the NH
 311 oceans in all four seasons. (g) Cumulative Distribution Functions (CDFs) of observed PAN, PPN, HCN and CO and
 312 the fractional contributions of modeled CO from non-biomass burning Asian sources in GEOS-5 ($CO_{GEOS5 \text{ non-BB Asia}}$)
 313 over the NH Pacific Ocean. Solid and dashed lines represent data with and without East Asian influence,
 314 respectively. (h) Scatter plots of observed PAN versus CO (top) with and (bottom) without East Asian influence
 315 colored using the ratios of $CO_{GEOS5 \text{ non-BB Asia}}$ to the total CO in GEOS-5.

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 320 **Figure 3.** (a) Probability distribution functions (PDFs) of observed and modeled CH_3CHO and β_{NO_2} . (b) Scatter
 321 plots of PA concentrations from both the model outputs (PA_{GMI}) and steady state (SS) calculations using modeled
 322 PA precursors ($\text{PA}_{\text{SS-GMI}}$) against SS calculations using observations. Data points are color-coded based on the
 323 absolute percent difference between modeled and observed CH_3CHO . (c) Scatter plots of PAN from SS calculations
 324 (PAN_{SS}) and model simulations (PAN_{GMI}) versus observed PAN. (d) Percent differences between observed and
 325 GMI-modeled PAN (red), GMI-modeled and GEOS5 modeled CO (blue), and observed and GEOS5-modeled CO
 326 (gray), which are illustrated as a function of fractional contribution of CO from global biomass burning ($\text{CO}_{\text{GEOSS5}}$
 327 Global BB) to total CO in GEOS5 in the SH.

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335 **Data Availability**

336 The data used in this work is available from the ATom campaign archive at
337 <https://doi.org/10.3334/ORNLDAAC/1925> (Wofsey et al., 2021). The GMI modelling results for the ATom
338 campaign are available at <https://doi.org/10.3334/ORNLDAAC/1897> (Strode et al., 2021).

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