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1 **Role of OH variability in the stalling of the global atmospheric CH₄ growth**
2 **rate from 1999 to 2006**

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17 **Abstract**

18 The growth in atmospheric methane (CH₄) concentrations over the past two decades has shown
19 large variability on a timescale of many years. Prior to 1999 the globally averaged CH₄
20 concentration was increasing at a rate of 6.0 ppb/yr, but during a stagnation period from 1999
21 to 2006 this growth rate slowed to 0.6 ppb/yr. Since 2007 the growth rate has again increased
22 to 4.9 ppb/yr. These changes in growth rate are usually ascribed to variations in CH₄ emissions.
23 We have used a 3-D global chemical transport model, driven by meteorological reanalyses and
24 variations in global mean hydroxyl (OH) concentrations derived from CH₃CCl₃ observations
25 from two independent networks, to investigate these CH₄ growth variations. The model shows
26 that between 1999 and 2006, changes in the CH₄ atmospheric loss contributed significantly to
27 the suppression in global CH₄ concentrations relative to the pre-1999 trend. The largest factor
28 in this is relatively small variations in global mean OH on a timescale of a few years, with
29 minor contributions of atmospheric transport of CH₄ to its sink region and atmospheric
30 temperature. Although changes in emissions may be important during the stagnation period,
31 these results imply a smaller variation is required to explain the observed CH₄ trends. The
32 contribution of OH variations to the renewed CH₄ growth after 2007 cannot be determined with
33 data currently available.



34 1. Introduction

35 The global mean atmospheric methane (CH_4) concentration has increased by a factor of 2.5
36 since the pre-industrial era, from approximately 722 ppb in 1750 to 1803.2 ± 0.7 ppb in 2011
37 (Etheridge et al., 1998; Dlugokencky et al., 2005). Over this time period methane has accounted
38 for approximately 20% of the total direct anthropogenic perturbation of radiative forcing by
39 long-lived greenhouse gases ($0.48 \pm 0.05 \text{ W/m}^2$), the second largest contribution after CO_2
40 (Cicerone et al., 1988; Myhre et al., 2013). This long-term methane increase has been attributed
41 to a rise in anthropogenic emissions from fossil fuel exploitation, agriculture, waste
42 management and biomass burning (Dlugokencky et al., 2011). Predictions of future CH_4 levels
43 require a complete understanding of processes governing emissions and atmospheric removal.

44 Since the mid-1980s measurements of CH_4 in discrete atmospheric air samples collected at
45 surface sites have been used to observe changes in the interannual growth rate of CH_4 (Rigby
46 et al., 2008; Dlugokencky et al., 2011, Kirschke et al., 2013). Nisbet et al. (2014) showed that
47 between 1984 and 1992 atmospheric CH_4 increased at ~ 12 ppb/yr, after which the growth rate
48 slowed to ~ 3 ppb/yr. In 1999 a period of near-zero growth began which continued until 2007.
49 In 2007 this stagnation period ended and since then average growth has increased again to ~ 6
50 ppb/yr (Rigby et al., 2008; Dlugokencky et al., 2011).

51 The reasons for the pause in CH_4 growth are not well understood. Bousquet et al. (2006)
52 performed an atmospheric transport inversion study to infer an increase in anthropogenic
53 emissions since 1999. Similarly, the EDGAR v3.2, bottom-up anthropogenic emission
54 inventory, an updated inventory to that used as an a priori by Bousquet et al. (2006), shows a
55 year-on-year increase in anthropogenic CH_4 emissions between 1999 and 2006 (Olivier et al.,
56 2005). This would suggest that a decrease in anthropogenic emissions is not the likely cause of
57 the pause in growth during this period. A second explanation is a reduction in wetland
58 emissions between 1999 and 2006, which is in part compensated by an increase in
59 anthropogenic emissions (Bousquet et al., 2006). However, more recently, Pison et al. (2013)
60 used two atmospheric inversions and found much more uncertainty in the role wetlands played
61 in the pause in growth over this period.

62 Dlugokencky et al. (2003) argued that the behaviour of global mean CH_4 up to around 2002
63 was characteristic of the system approaching steady state, accelerated by decreasing emissions
64 at high northern latitudes in the early 1990s and fairly constant emissions elsewhere. However,
65 since then there have been notable perturbations to the balance of sources and sinks (Rigby et
66 al., 2008). This has been, at least partly, attributed to increases in wetland (Bousquet et al.,
67 2011) and anthropogenic emissions (Bousquet et al., 2006). Recent changes in emissions are
68 not well constrained and the reasons for the renewed growth are also not fully understood.

69 Atmospheric chemistry has also been hypothesised to play a role in past variations in CH_4
70 growth rates. The major (90%) sink of atmospheric CH_4 is via reaction with the hydroxyl
71 radical, OH. Variations in the concentration of OH ($[\text{OH}]$), or changes to the reaction rate
72 through changes in temperature, therefore have the potential to affect CH_4 growth. Previous
73 studies have suggested that an increase in atmospheric OH concentration may have been at
74 least partly responsible for a decrease in the CH_4 growth rate (Lelieveld et al., 2004; Fiore et



75 al., 2006). This rise in OH has been attributed to an increase in lightning NO_x (Fiore et al.,
76 2006). The abundance of other species such as H₂O, O₃, CO and CH₄ also determine the
77 concentration of OH (Leliveld et al., 2004). Prinn et al. (2005) suggested that major global
78 wildfires and El Nino Southern Oscillation (ENSO) events could influence [OH] variability. A
79 reduced number of small- to moderate-magnitude volcanic eruptions during the CH₄ stagnation
80 period (Carn et al., 2015; Mills et al., 2015) may have increased [OH], due to increased
81 downward UV radiation. Recently, Patra et al. (2014) analysed global CH₃CCl₃ observations
82 for 2004-2011 to derive the interhemispheric ratio of OH. In contrast to many model results
83 which suggest higher mean [OH] in the north, they derived similar values for both hemispheres.

84 Warwick et al. (2002) investigated the impact of meteorology on atmospheric CH₄ growth rates
85 from 1980 to 1998, i.e. well before the observed recent pause. They concluded that atmospheric
86 conditions could be an important driver in the interannual variability (IAV) of atmospheric
87 CH₄. In similar studies a combination of atmospheric dynamics and changes in emissions were
88 shown to explain some of the earlier past trends in atmospheric CH₄ (Fiore et al., 2006; Patra
89 et al., 2009). This paper builds on these studies to investigate the chemical and non-chemical
90 atmospheric contribution to the recent variations in CH₄ growth. By ‘non-chemical’ we mean
91 transport-related influences, although the loss of CH₄ is ultimately due to chemistry as well.
92 We use a global chemical transport model to simulate the period from 1993 to 2011 and to
93 quantify the impact of variations in [OH] and meteorology on atmospheric CH₄ growth.

94 2. Data and Models

95 2.1 NOAA and AGAGE CH₄ Data and Derived OH

96 We have used surface CH₄ observations from 19 National Oceanographic and Atmospheric
97 Administration/Earth System Research Laboratory (NOAA/ESRL) cooperative global air
98 sampling sites (Dlugokencky et al., 2014) over 1993-2009 (see Table 1). To calculate the global
99 average concentration, measurements were interpolated across 180 latitude bins, which were
100 then weighted by surface area. We have also used the same method to derive global mean CH₄
101 based on 5 sites from the Advanced Global Atmospheric Gases Experiment (AGAGE) network
102 (Prinn et al., 2000; Cunnold et al., 2002; Prinn et al., 2015).

103 Montzka et al. (2011) used measurements of methyl chloroform (CH₃CCl₃) from an
104 independent set of flasks sampled at a subset of NOAA air sampling sites to derive global [OH]
105 anomalies from 1997 to 2007. They argued that uncertainties in emissions are likely to limit
106 the accuracy of the inferred inter-annual variability in global [OH], particularly before 1997.
107 At this time the emissions were large but decreasing rapidly due to the phaseout of CH₃CCl₃
108 production and consumption, and the large atmospheric gradients were also more difficult to
109 capture accurately with only few measurement sites. Instrument issues caused an interruption
110 to their CH₃CCl₃ time series in 2008/9. We have averaged these (based on the red curve in
111 Figure 3 of Montzka et al.) into yearly anomalies to produce relative interannual variations in
112 the mean [OH]. Similarly, Rigby et al. (2013) used CH₃CCl₃ measurements from the 5-station
113 AGAGE network in a 12-box model to produce yearly global [OH] anomalies from 1995 (the
114 date from which data from all 5 stations is available) to 2010. These two timeseries, which
115 convert anomalies in the CH₃CCl₃ decay rate into anomalies in [OH] using constant



116 temperature, correspond to the best estimate of [OH] variability from the two measurement
117 networks by the groups who operate them. We then applied these two series of yearly anomalies
118 uniformly to the global latitude-height [OH] field used in the recent TransCom CH₄ model
119 intercomparison (see Patra et al., 2011), which itself was derived from a combination of semi-
120 empirically calculated tropospheric OH distributions (Spivakovsky et al. 2000; Huijnen et al.,
121 2010) and 2-D model simulated stratospheric loss rates (Velders, 1995). For consistency
122 between the model experiments, both sets of yearly anomalies were scaled so that the mean
123 [OH] between 1997 and 2007 (the overlap period where NOAA and AGAGE anomalies are
124 both available) equalled the TransCom [OH] value. In the rest of this paper we refer to these
125 two OH datasets as ‘NOAA-derived’ and ‘AGAGE-derived’.

126 These two calculations of yearly [OH] anomalies use slightly different assumptions for
127 CH₃CCl₃ emissions after 2002. Before that date they use values from Prinn et al. (2005). The
128 NOAA data then assumed a 20% decay in emission for each subsequent year (Montzka et al.,
129 2011), while AGAGE used United Nations Environment Programme (UNEP) consumption
130 values (UNEP, 2015). Holmes et al. (2013) suggested that inconsistencies in CH₃CCl₃
131 observations between the AGAGE and NOAA networks also limit understanding of OH
132 anomalies for specific years due to an unexplained phasing difference of up to around 3 months.
133 As we are interested in the impact of [OH] changes over longer time periods (e.g. 2000 – 2006),
134 this phase difference will be less important. We have investigated the impact of the different
135 CH₃CCl₃ observations and assumed emissions on the derived [OH] anomalies (see Section
136 3.1).

137 2.2 TOMCAT 3-D Chemical Transport Model

138 We have used the TOMCAT global atmospheric 3-D off-line CTM (Chipperfield, 2006) to
139 model atmospheric CH₄ and CH₃CCl₃ concentrations. The TOMCAT simulations were forced
140 by winds and temperatures from the 6-hourly European Centre for Medium-Range Weather
141 Forecasts (ECMWF) ERA-Interim reanalyses (Dee et al., 2011). They covered the period 1993
142 to 2011 with a horizontal resolution of 2.8° × 2.8° and 60 levels from the surface to ~60 km.

143 The TOMCAT simulations use annually repeating CH₄ emissions, which have been scaled to
144 previous estimates of 553 Tg/yr (Ciais et al., 2013), taken from various studies (Fiore et al.,
145 2006; Curry et al., 2007; Bergamaschi et al., 2009; Pison et al., 2009; Spahni et al., 2011; Ito
146 et al., 2012). Annually-repeating anthropogenic emissions (except biomass burning) were
147 calculated from averaging the EDGAR v3.2 (2009) inventory from 1993 to 2009 (Olivier and
148 Berowski, 2001). Biomass burning emissions were calculated using the GFED v3.1 inventory
149 and averaged from 1997 to 2009 (van der Werf et al., 2010). The Joint UK Land Environment
150 Simulator (JULES) (Best et al., 2011; Clark et al., 2011; Hayman et al., 2014) was used to
151 calculate a wetland emission inventory between 1993 and 2009, which was then used to
152 produce a scaled mean annual cycle. Annually-repeating rice (Yan et al., 2009), hydrate, mud
153 volcano, termite, wild animal and ocean (Matthews et al., 1987) emissions were taken from the
154 TransCom CH₄ study (Patra et al., 2011). The methane loss fields comprised an annually
155 repeating soil sink (Patra et al., 2011), an annually repeating stratospheric loss field (Velders,



156 1995) and a specified [OH] field. The model was spun up for 15 years prior to initialising the
157 simulations, using emission data from 1977 where available and annual averages otherwise.

158 Fifteen TOMCAT simulations were performed each with a CH₄ tracer and a CH₃CCl₃ tracer.
159 The runs had differing treatments of meteorology (winds and temperature) and [OH] (see Table
160 2). Simulations with repeating [OH] fields (RE_xxxx) used the TransCom dataset. The other
161 runs with varying [OH] used the NOAA-derived or AGAGE-derived [OH] fields based on the
162 original published work or our estimates (see Section 3.1). For these runs, the mean [OH] field
163 is used where the respective NOAA or AGAGE-derived [OH] is unavailable or uncertain
164 (before 1997 / after 2007 for NOAA and before 1997 / after 2009 for AGAGE). The five
165 simulations with fixed wind and temperature fields (with labels ending in FTFW) used the
166 ERA-Interim analyses from 1996 repeated for all years. The five simulations with varying
167 winds and fixed temperature (with labels ending in FTVW) used zonal mean temperature fields
168 averaged from 1993-2009. The OH anomalies are derived from the anomaly in the CH₃CCl₃
169 loss rate, which combines variations in atmospheric OH concentration with variations in
170 temperature which affect the rate constant of the CH₃CCl₃ + OH reaction. To quantify the
171 importance of this temperature effect we also performed 5 model runs which allow both winds
172 and temperature to vary interannually according to ERA-Interim data (labels ending VTVW).

173 3. Results

174 3.1 Correlation of CH₄ variations with OH and temperature

175 We first investigate the extent to which variations in the observed CH₄ growth rate correlate
176 with variations in derived [OH]. Figure 1a shows the published NOAA-derived and AGAGE-
177 derived global [OH] anomalies along with the annual CH₄ growth rate estimated from the
178 NOAA and AGAGE measurements. The two [OH] series show the similar behaviour of
179 negative anomalies around 1997 and 2006/7, and an extended period of more positive
180 anomalies in between. For the time periods covered by the NOAA (1997-2007) and AGAGE
181 (1997-2009) CH₃CCl₃ observations, the two derived [OH] timeseries show negative correlations
182 with the CH₄ growth from NOAA (regression coefficient, R = -0.32) and AGAGE (R = -0.64).
183 Only the AGAGE [OH] correlation, from the longer timeseries, is statistically significant at the
184 90% level.

185 We can use a simple ‘global box model’ (see Supplement S1) to estimate the [OH] variations
186 required to fit the observed CH₄ growth rate variations assuming constant CH₄ emissions and
187 temperature (black line in Figure 1b). This provides a crude guide to the magnitude of OH
188 variations which could be important for changes in the CH₄ budget. Our results are consistent
189 with those of Montzka et al. (2011) who performed a similar analysis on the NOAA CH₄ data.
190 The required [OH] rarely exceeds their CH₃CCl₃-derived interannual variability (IAV) range
191 of [OH] (±2.3%, shown as shading in the figure). Also shown in Figure 1b are the published
192 estimates of the global mean OH anomalies from Figure 1a, converted to concentration units
193 (see Section 2.1). The relative interannual variations in [OH] required to fit the CH₄
194 observations match the CH₃CCl₃-derived [OH] variations in many years, for example from
195 1998-2002 (see Montzka et al., 2011). Some of the derived variations in [OH] exceed that



196 required to match the CH₄ growth rate, with larger negative anomalies in the early and later
197 years and some slightly larger positive values in middle of the period.

198 Figures 1c and 1d show our estimates of [OH] using NOAA and AGAGE observations and
199 two assumptions of post-2000 CH₃CCl₃ emissions (see Section 2.1) in a global box model. The
200 figures also compare our OH estimates with the NOAA-derived and AGAGE-derived [OH]
201 anomalies based on the work of the observation groups (Figure 1a). Our results demonstrate
202 the small impact of using different observations and post-2000 emission assumptions (compare
203 filled and open red circles for the two panels). For these box model results there is also only a
204 very small effect of using annually varying temperature (compare red and blue lines). In later
205 years the choice of observations has a bigger impact than the choice of emissions on the derived
206 [OH]. For AGAGE-derived values (Figure 1d) our estimates agree well with the published
207 values of Rigby et al., (2013), despite the fact we use a global box model while they used a
208 more sophisticated 12-box model. In contrast, there are larger differences between our values
209 and the NOAA-derived OH variability published by Montzka et al. (2011) (Figure 1c), despite
210 both studies using box models. In particular, around 2002-2003 we overestimate the positive
211 anomaly in [OH]. We also estimate a much more negative OH anomaly in 1997 than Montzka
212 et al., though we slightly underestimate the published AGAGE-derived anomaly in that year
213 (Figure 1d). Tests show that differences between our results and the NOAA box model are due
214 to the treatment of emissions. This suggests a larger uncertainty in the inferred low 1997 [OH]
215 value, when emissions of CH₃CCl₃ were decreasing rapidly, although reasons why atmospheric
216 [OH] might have been anomalously low were discussed by Prinn et al. (2005). In the subsequent
217 analysis we use the OH variability from the published NOAA and AGAGE studies as input to
218 the 3-D model.

219 3.2 TOMCAT Simulations

220 Overall, Figure 1 shows the potential importance of small, observationally derived variations
221 in OH concentrations to impact methane growth. We now investigate this quantitatively in the
222 framework of a 3-D CTM.

223 3.2.1 Methyl Chloroform

224 The TOMCAT simulations include a CH₃CCl₃ tracer. This allows us to verify that our approach
225 of using a global OH field, scaled by derived anomalies, allows the model to reproduce the
226 observed magnitude and variability of CH₃CCl₃ decay accurately. Figure 2a shows that the
227 model, with the imposed [OH] field, does indeed simulate the global decay of CH₃CCl₃ very
228 well. This justifies our use of the ‘offline’ [OH] field, as models with interactive tropospheric
229 chemistry can produce a large range in absolute global mean [OH] and therefore in lifetimes
230 of gases such as CH₃CCl₃. For example, Voulgarakis et al., (2013) analysed the global mean
231 [OH] from various 3D models and found a range of 0.55×10^6 to 1.34×10^6 molecules cm⁻³.
232 Figure 2a also shows that the global mean CH₃CCl₃ from the NOAA and AGAGE networks
233 differ by ~2.5ppt around 1993-1996, but since then this difference has become smaller.

234 The observed and modelled CH₃CCl₃ decay rate anomalies (calculated using the method of
235 Holmes et al., (2013) with a 12-month smoothing) are shown in Figures 2b and 2c (different
236 panels are used for AGAGE and NOAA comparisons for clarity). The model and observation-



237 derived results both tend to show a faster CH_3CCl_3 decay (more positive anomaly) in the middle
238 of the period, with slower decay at the start and end. The anomalies for the NOAA and
239 AGAGE-derived OH show periodic variations on a timescale of 2-3 yrs but with a phase shift
240 between the two datasets of 3 months, as noted by Holmes et al., (2013). The model runs with
241 OH variability prescribed from the observations and varying winds also show these periodic
242 variations with correlation coefficients ranging from 0.71 – 0.90. The correlation values for
243 these runs using varying OH are all larger than the run using repeating OH (for RE_FTVW
244 $R=0.62$ compared to AGAGE data and 0.67 compared to NOAA data). Note that for CH_3CCl_3
245 decay there are only small differences between the 3-D simulations which use varying
246 temperatures and the corresponding runs which use fixed temperature (e.g. simulation
247 RE_VTVW versus RE_FTVW). This agrees with the results of Montzka et al (2011) based on
248 their box model. This shows that the largest contribution from the CH_3CCl_3 decay rate anomaly
249 comes from variations in atmospheric OH concentration, rather than atmospheric temperature.
250 The simulations with repeating winds show less variability in the CH_3CCl_3 decay rate,
251 particularly in the period 1999-2004, but the small difference suggests that the interannual
252 variability in the observed CH_3CCl_3 decay rate is driven primarily by the variations in the OH
253 concentration. The remaining interannual variability in run RE_FTVW is due to variations in
254 emissions.

255 Figure 3 shows the CH_3CCl_3 decay and decay rate anomalies at four selected stations, two from
256 the NOAA network and two from the AGAGE network. The good agreement in the global
257 CH_3CCl_3 decay in Figure 2 is also seen at these individual stations. At the AGAGE stations of
258 Mace Head and Gape Grim, the model runs with varying OH perform better in capturing the
259 decay rate anomalies than the runs with repeating OH. However, the impact of variability in
260 the winds (solid lines versus dotted lines) is more apparent at these individual stations
261 compared to the global means. At the NOAA station of Mauna Loa the model run with varying
262 OH and varying winds also appears to perform better in capturing the observed variability in
263 CH_3CCl_3 decay. At the South Pole the observed variability is small, except in 2000-2002. This
264 feature is not captured by the model.

265 In summary, Figures 2 and 3 show that the global OH fields that we have constructed from
266 different datasets can perform well in capturing the decay of CH_3CCl_3 and its anomalies both
267 globally and at individual stations. Although, the interannual variability in global mean OH has
268 been derived from these CH_3CCl_3 observations, the figures do show that the reconstructed
269 model OH fields (which also depend on the methodology discussed in Section 2) perform well
270 in simulating CH_3CCl_3 within the 3D model. Therefore, we would argue that these fields are
271 suitable for testing the impact of OH variability on the methane growth rate. Even so, it is
272 important to bear in mind that these fields may not represent the true changes in atmospheric
273 OH, particularly if the interannual variability in CH_3CCl_3 emissions was a lot different to that
274 assumed here. However, we would again note that we are focussing on the impact of multi-
275 year variability which appears more robustly determined by the networks under differing
276 assumptions of temperature and emissions than year-year variability.

277

278 3.2.2 Methane



279 Figure 4 shows deseasonalised modelled surface CH₄ from the 3-D CTM simulations compared
280 with in-situ observations from a northern high-latitude station (Alert), two tropical stations
281 (Mauna Loa and Tutuila), a southern high-latitude station (South Pole) and the global average
282 of the NOAA and AGAGE stations. The global comparisons are shown for simulations both
283 with varying and repeating meteorology. Figure 5 shows the global annual CH₄ growth rates
284 with a 12-month smoothing (panel a) and differences between the model and NOAA and
285 AGAGE observations (panels b and c). The changes in the modelled global mean CH₄ over
286 different time periods are given in Table 3.

287 Figure 4 shows that in 1993, at the end of the model spin-up, the simulations capture the global
288 mean CH₄ level well, along with the observed values at a range of latitudes. The exception is
289 at high northern latitudes. However, these differences are not important when investigating the
290 change in the global growth rate. The global change in atmospheric CH₄ in the simulations
291 with varying winds for 1993 to the end of 2009 is between 75 and 104 ppb, compared to 56
292 and 66 ppb in the observations.

293 Model run RE_FTFW does not include interannual variations in atmospheric transport or CH₄
294 loss. Therefore, the modelled CH₄ gradually approaches a steady state value of ~1830 ppb
295 (Figure 4f). The rate of CH₄ growth decreases from 7.9 ppb/yr (1993-1998) to 1.4 ppb/yr (2007-
296 2009). Compared to run RE_FTFW, the other simulations introduce variability on this CH₄
297 evolution.

298 Run RE_FTVW includes interannual variability in wind fields which may alter the transport
299 of CH₄ from the source (emission) to the sink regions. The largest difference between runs
300 RE_FTFW and RE_FTVW occurs after 2000 (Figure 4f). During the stagnation period (1999-
301 2006) run RE_FTVW has a smaller growth rate of 3.5 ppb/yr compared to 4.1 ppb/yr in run
302 RE_FTFW, showing that variations in atmospheric transport made a small contribution to the
303 slowdown in global mean CH₄ growth.

304 Compared to run RE_FTVW, runs AP_FTVW, AL_FTVW, NP_FTVW and NL_FTVW
305 include CH₃CCl₃-derived interannual variations in [OH] which introduce large changes in
306 modelled CH₄, which are more in line with the observations (Figure 4e and 5). These runs
307 produce turnarounds in the CH₄ growth in 2001/2 (becomes negative) and 2005/6 (returns to
308 being positive). For AGAGE-derived [OH] (runs AP_FTVW, AL_FTVW) the large negative
309 anomaly in OH in 1997 produces a significant increase in CH₄ prior to the turnaround in 2001.

310 Table 3 summarises the change in global mean CH₄ over different time periods. These periods
311 are defined by the key dates in the observed record, i.e. 1999 and 2006 as the start and end
312 dates of the stagnation period. Comparison of Figure 4e and Table 3 shows, however, that the
313 timing of the largest modelled change in growth rate do not necessarily coincide with those
314 dates. That is understandable if other factors not considered here, e.g. emission changes, are
315 contributing to the change in global CH₄ concentration. It does mean that the summary values
316 in Table 3 do not capture the full impact of the changes in [OH] and winds within the stagnation
317 period. Figure 4e shows that model runs with varying OH perform better in simulating the
318 relative CH₄ trend 1999 to around 2004.



319 Table 3 shows that runs NP_FTVW and NL_FTVW (NOAA-derived [OH]) produce a small
320 modelled CH₄ growth of 2.5-3.1 ppb/yr during the stagnation period 1999-2006, compared to
321 1.0 ppb/yr for run AP_FTVW (AGAGE-derived [OH]). The AGAGE results are slightly larger
322 than the observed growth rate of 0.6-0.7 ppb/yr. Runs AL_FTVW, AP_FTVW, NL_FTVW
323 and NP_FTVW capture the observed strong decrease in the CH₄ growth rate. Clearly, these
324 runs demonstrate the significant potential for relatively small variations in mean [OH] to affect
325 CH₄ growth. Excluding the stagnation period the mean modelled CH₄ lifetime in run
326 NP_FTVW is 9.4 years, but this decreases slightly by 0.01 years during the stagnation period.
327 For run AP_FTVW there is a decrease of 0.18 years from 9.6 years between the same intervals.
328 The results from all the CTM simulations during 1999-2006 indicate that the accuracy of
329 modelled CH₄ growth is improved by accounting for interannual variability in [OH] as derived
330 from CH₃CCl₃ observations, and interannual variability in meteorology.

331 The variation of [OH] after 2007 cannot be determined from the available NOAA data so run
332 NP_FTVW used the mean [OH] field for all subsequent years. The modelled CH₄ increase of
333 3.5 ppb/yr underestimates the observations (4.9 ppb/yr). Should the lower [OH] of 2007 have
334 persisted then the model would have produced a larger increase in CH₄, in better agreement
335 with the observations. The AGAGE-derived [OH] for 2007-2009 (run AP_FTVW) produces a
336 larger CH₄ growth relative to the previous years (8.8 ppb/yr). Runs RE_FTFW (1.4 ppb/yr) and
337 RE_FTVW (1.8 ppb/yr) both show a decreased rate of growth during the final 5 years,
338 consistent with a system approaching steady state.

339 Figure 5a shows the global CH₄ growth rate derived from the AGAGE and NOAA networks
340 together with selected model simulations. Figures 5b and c show the differences between the
341 model simulations and the NOAA and AGAGE observations, respectively. The runs which
342 include variations in [OH] agree better with the observed changes, i.e. larger R values in panel
343 (a) and the model lines are closer to the y=0 line in panels (b) and (c), especially in the first 5
344 years of the stagnation period. It is interesting to note that the relative impacts of wind and
345 temperature variations are larger for CH₄ than for CH₃CCl₃ (compare simulations RE_FTFW,
346 RE_FTVW and RE_VTVW in Figures 2 and 5a). The temperature dependences of the OH loss
347 reactions are similar for the two species (see Supplement S1) but the impact of transport from
348 emissions regions to chemical loss regions is more variable for CH₄. This needs to be considered
349 when applying results derived from CH₃CCl₃ to CH₄.

350 4. Discussion and Conclusions

351 Our model results suggest that variability in atmospheric [OH] and transport played key roles
352 in the observed recent variations in CH₄ growth, particularly during the CH₄ stagnation period
353 between 1999 and 2006. The 3-D CTM calculations show that during the stagnation period,
354 variations in atmospheric conditions in the tropical lower to mid-troposphere could potentially
355 account for an important component of the observed decrease in global CH₄ growth. Within
356 this, small increases in [OH] were the largest factor, while variations in transport made a
357 smaller contribution. Note again, however, that the ultimate loss of CH₄ is still due to
358 chemistry. The role of atmospheric temperature variations is factored into the observationally
359 derived OH, but model experiments show that changes in the OH concentration itself is most



360 important. The remainder of the variation can be ascribed to other processes not considered in
361 our runs such as emission changes. There are also measurement uncertainties to consider and
362 the possible underrepresentation of the global mean CH_3CCl_3 which will affect the derived OH
363 concentration. Our results are consistent with an earlier budget study which analysed 1991 to
364 2004 and found that variations in [OH] were the main control of variations in atmospheric CH_4
365 lifetime (65%), with temperature accounting for a smaller fraction (35%) (Fiore et al., 2006).
366 As we have noted here the CH_4 lifetime can also be affected by emissions distributions which
367 affects transport to the main loss regions.

368 Prior to the stagnation period the simulation using AGAGE-derived [OH] overestimates CH_4
369 growth when compared to observations which degrades the agreement with the observed CH_4
370 variations. A likely cause of this is inaccuracies in derived [OH] in 1997 when emissions still
371 played a large role in the observed CH_3CCl_3 and the e-fold decay had not yet stabilised
372 (Montzka et al., 2011).

373 We have not accounted for expected variations in CH_4 emissions in this study. We can conclude
374 that although global CH_4 emissions do vary year-to-year, the observed trend in CH_4 growth
375 between 1999 and 2006 was impacted by changing atmospheric processes that affected CH_4
376 loss. Changes in emissions are still important and likely still dominate CH_4 variations over
377 other time periods. The observed changes in growth rates during ENSO events in e.g. 1998 are
378 poorly captured by the meteorological changes considered here and can be attributed to changes
379 in emissions through changing precipitation and enhanced biomass burning (Hodson et al.,
380 2011). The renewed growth of CH_4 in 2007 is also poorly captured by all model simulations
381 without varying [OH]. The observed decrease in AGAGE and NOAA-derived [OH] coincides
382 with the increase in CH_4 growth in 2007, although the currently available data do not allow for
383 a more detailed investigation of the possible contribution of [OH] changes in this recent
384 increase.

385 Despite the differences in year-to-year variability in [OH] derived from CH_3CCl_3 observations
386 (Holmes et al., 2013), we find that [OH] variability derived from two different networks of
387 surface CH_3CCl_3 observations over multi-year periods provide insights into atmospheric CH_4
388 variations. Improved quantification of the role of OH variability will require efforts to reduce
389 uncertainties associated with estimating [OH]. Estimates of global mean [OH] in recent years
390 from CH_3CCl_3 observations is becoming increasingly difficult because CH_3CCl_3 levels are
391 currently <5 ppt; hence this may limit the accuracy of derived [OH] and its variability in future
392 years (Lelieveld et al., 2006). Wennberg et al. (2004) also noted that there can be time
393 variations in the small uptake of CH_3CCl_3 by the oceans, which can also affect the derived
394 [OH] concentrations and are not considered here. Overall our study suggests that future
395 atmospheric trends in CH_4 are likely to be strongly influenced by not only emissions but also
396 changes in processes that affect atmospheric loss. The accuracy of predictions would therefore
397 be improved by including variations in [OH] and meteorology.

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- 600



601 **Tables**

602

603 **Table 1.** List of NOAA and AGAGE stations which provided CH₄ and CH₃CCl₃ observations.

Site Code	Site Name	Lat. (°N)	Lon. (°N)	Altitude (km)	CH ₄	CH ₃ CCl ₃	Start Date ⁺⁺	End Date
ABP	Arembepe, Brazil	-12.77	-38.17	0	NOAA		27/10/2006	12/01/2010
ALT	Alert, Canada	82.45	-62.51	0.2	NOAA	NOAA	10/06/1985	Ongoing
ASC	Ascension Island, UK	7.97	-14.4	0.09	NOAA		11/05/1983	Ongoing
BRW	Barrow, USA	71.32	-156.61	0.01	NOAA	NOAA	06/04/1983	Ongoing
CGO	Cape Grim, Australia	-40.68	144.69	0.09	NOAA/AGAGE	NOAA/AGAGE	19/04/1984	Ongoing
HBA	Halley Station, UK	-75.61	-26.21	0.03	NOAA		17/01/1983	Ongoing
ICE	Storhofdi, Iceland	63.4	-20.29	0.12	NOAA		02/10/1992	Ongoing
KUM	Cape Kumukahi, USA	19.5	-154.8	0.02	-	NOAA	-	-
LEF	Park Falls, USA	45.9	-90.3	0.47	-	NOAA	-	-
MHD	Mace Head, Ireland	53.33	-9.9	0.01	NOAA/AGAGE	AGAGE**	03/06/1991	Ongoing
MLO	Mauna Loa, USA	19.54	-155.58	3.4	NOAA	NOAA	06/05/1983	Ongoing
NWR	Niwot Ridge, USA	40.05	-105.59	3.52	NOAA	NOAA	21/06/1983	Ongoing
PAL	Pallas-Sammaltunturi, Finland	67.97	24.12	0.56	NOAA		21/12/2001	Ongoing
PSA	Palmer Station, USA	-64.92	-64	0.01	NOAA	**	01/01/1983	Ongoing
RPB	Ragged Point, Barbados	13.17	-59.43	0.02	NOAA/AGAGE	AGAGE	14/11/1987	Ongoing
SEY	Mahe Island, Seychelles	-4.68	55.53	0	NOAA		12/05/1983	Ongoing
SMO	Tutuila, American Samoa	-14.25	-170.56	0.04	NOAA	NOAA/AGAGE	23/04/1983	Ongoing
SPO	South Pole, USA	-89.98	-24.8	2.81	NOAA	NOAA	20/02/1983	Ongoing
STM	Ocean Station M, Norway	66	2	0	NOAA		29/04/1983	27/11/2009
SUM	Summit, Greenland	72.6	-38.42	3.21	NOAA	**	23/06/1997	Ongoing
THD	Trinidad Head, USA	41.1	-124.1	0.1	AGAGE	AGAGE**	09/1995	Ongoing
ZEP	Ny-Alesund, Norway/Sweden	78.91	11.89	0.47	NOAA		11/02/1994	Ongoing

604

605 ⁺⁺For NOAA CH₃CCl₃ data the record starts in 1992 at 7 of the 9 stations used here. It started
 606 in 1995 for KUM and 1996 for LEF.

607 **NOAA flask data from these sites was not used in the present study or in Montzka et al.,

608 (2011).

609 **Table 2.** Summary of the fifteen TOMCAT 3-D CTM simulations.

Run	OH time variation	Meteorology ^b	
		Winds ^c	Temperature ^d
RE_FTFW	Repeating ^a	Fixed	Fixed
RE_FTVW	Repeating ^a	Varying	Fixed
RE_VTVW	Repeating ^a	Varying	Varying
AP_FTFW	AGAGE (Rigby et al., 2013)	Fixed	Fixed
AP_FTVW	AGAGE (Rigby et al., 2013)	Varying	Fixed
AP_VTVW	AGAGE (Rigby et al., 2013)	Varying	Varying
AL_FTVT	AGAGE (this work)	Fixed	Fixed
AL_FTVW	AGAGE (this work)	Varying	Fixed
AL_VTVW	AGAGE (this work)	Varying	Varying
NP_FTFW	NOAA (Montzka et al., 2011)	Fixed	Fixed
NP_FTVW	NOAA (Montzka et al., 2011)	Varying	Fixed
NP_VTVW	NOAA (Montzka et al., 2011)	Varying	Varying
NL_FTFW	NOAA (this work)	Fixed	Fixed
NL_FTVW	NOAA (this work)	Varying	Fixed
NL_VTVW	NOAA (this work)	Varying	Varying

610 a. Annually repeating [OH] taken from Patra et al. (2011).

611 b. Varying winds and temperatures are from ERA-Interim.

612 c. Fixed winds using repeating ERA-Interim winds from 1996.

613 d. Fixed temperatures use zonal mean ERA-Interim temperatures averaged over 1993-2009.



614 **Table 3.** Calculated methane changes over different time periods from selected TOMCAT
 615 experiments and the NOAA and AGAGE observation networks.

Model run or observation network	Global mean ΔCH_4 /ppb (ppb/yr)			
	2009-1993	1998-1993	2006-1999	2009-2007
RE_FTFW	85.0 (5.0)	47.2 (7.9)	32.9 (4.1)	4.3 (1.4)
RE_FTVW	82.2 (4.8)	48.2 (8.0)	27.8 (3.5)	5.4 (1.8)
RE_VTVW	74.6 (4.4)	45.6 (7.6)	23.1 (2.9)	5.3 (1.8)
AP_FTVW ^a	97.7 ^e (5.7)	62.3 ^e (10.4)	8.2 (1.0)	26.4 (8.8)
AL_FTVW ^b	104.2 ^e (6.1)	58.4 ^e (9.7)	17.3 (2.2)	27.5 (9.2)
NP_FTVW ^c	86.2 ^f (5.1)	49.7 ^f (8.3)	24.8 (3.1)	10.6 ^f (3.8)
NL_FTVW ^d	91.4 ^f (5.4)	58.8 ^f (9.8)	20.1 (2.5)	11.3 ^f (3.8)
NOAA obs.	56.1 (3.3)	36.0 (6.0)	4.8 (0.6)	14.7 (4.9)
AGAGE obs.	66.3 (3.9)	42.6 (7.1)	5.6 (0.7)	17.4 (5.8)

616

617 a. Taken from Rigby et al. (2013) and Patra et al. (2011).

618 b. Using 1997-2009 relative annual changes in mean [OH] derived from AGAGE data
619 (Cunnold et al., 2002).

620 c. Taken from Montzka et al. (2011) and Patra et al. (2011).

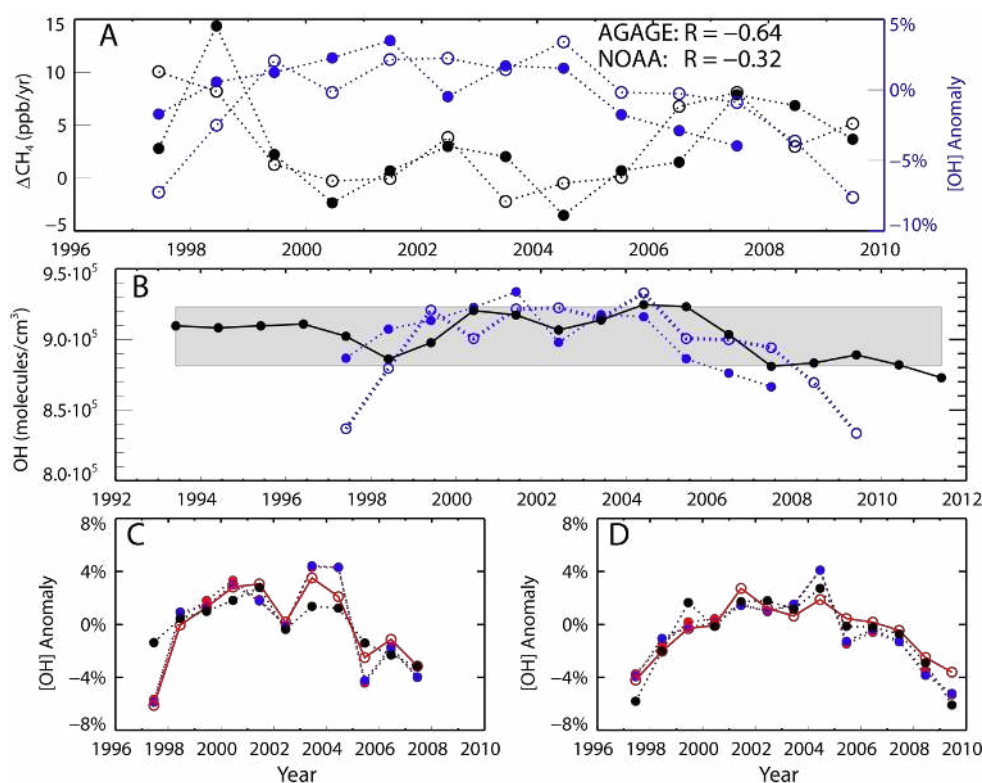
621 d. Using 1997-2007 relative annual changes in mean [OH] derived from NOAA data (Prinn
622 et al., 2015).

623 e. Value using mean [OH] from 1993-1996.

624 f. Value using mean [OH] from 1993-1996 and 2008-2011.



625 **Figures**



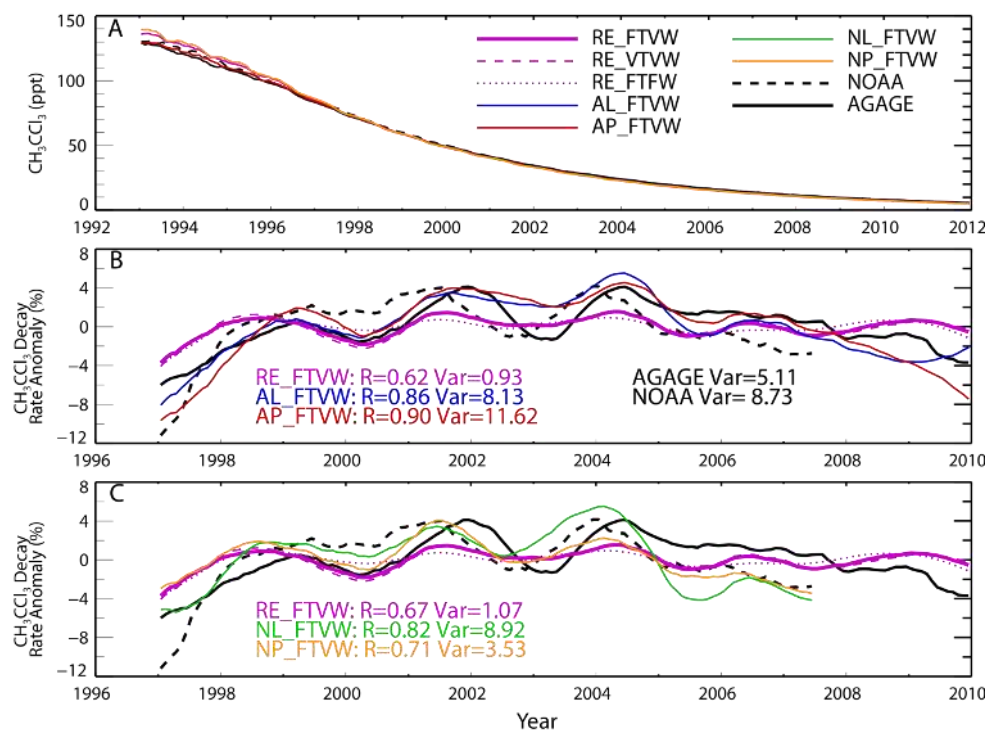
626

627 **Figure 1.** (a) Annual global CH₄ growth rate (ppb/yr) derived from NOAA (filled black circles)
628 and AGAGE (open black circles) data (left hand y-axis), and published annual global [OH]
629 anomalies derived from NOAA (filled blue circles, 1997-2007) and AGAGE (open blue
630 circles, 1997-2009) CH₃CCl₃ measurements (right hand y-axis) (see text). (b) Annual mean
631 [OH] (molecules/cm³) required for global box model (see Supplement S1) to fit yearly
632 variations in NOAA CH₄ observations assuming constant emissions and temperature ($E=553$
633 Tg/yr, $T=272.9$ K), based on Montzka et al. (2011) (solid black line). The shaded region
634 denotes [OH] deviation of $\pm 2.3\%$ from the 1993-2011 mean. Also shown are the NOAA- and
635 AGAGE-derived anomalies from panel (a) for an assumed mean OH (see Section 2.1). (c) Our
636 estimates of [OH] derived from NOAA CH₃CCl₃ calculated using a global box model
637 (Supplement S1) using repeating (blue) and varying (red) annual mean temperature and the
638 CH₃CCl₃ emission scenario from UNEP (2015) (filled circles and dashed lines). Also shown
639 for varying temperatures are results using the emissions of Montzka et al (2011) (red open
640 circles and solid line) based on (Prinn et al. 2005) and the NOAA-derived values from panel
641 (a) (black dashed line and circles). (d) As panel (c) but for OH derived from AGAGE CH₃CCl₃
642 observations.

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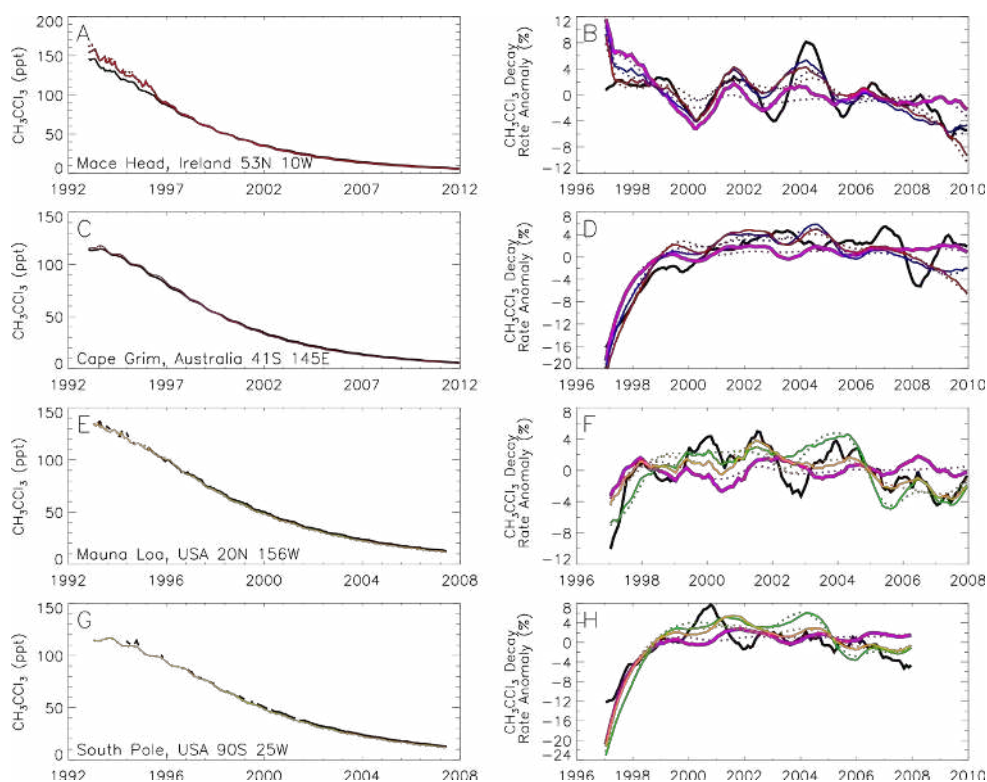


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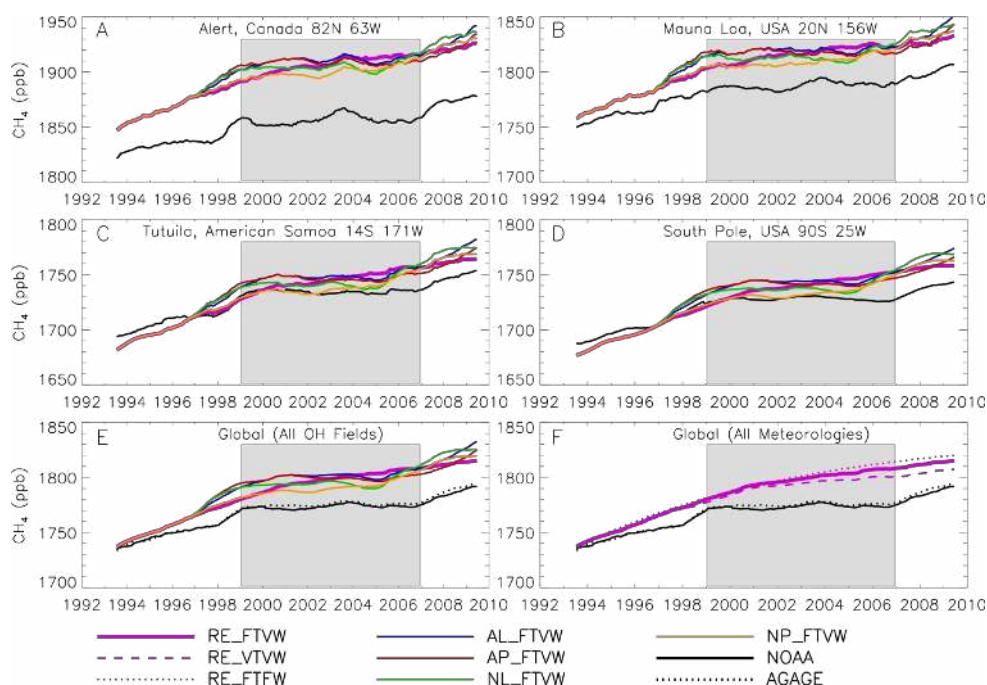
645

646 **Figure 2.** (a) Global mean surface CH_3CCl_3 (ppt) from NOAA (black dashed) and AGAGE
647 (black solid) observations from 1993 to 2012. Also shown are results from five TOMCAT
648 simulations with fixed temperatures and varying winds (see Table 1). (b) Global surface
649 CH_3CCl_3 decay rate anomalies from NOAA and AGAGE along with model runs RE_FTVW,
650 AL_FTVW and AP_FTVW (solid lines). Results from runs RE_FTFW and RE_VTVW are
651 shown as a purple dotted line and dashed line, respectively. Observation and model anomalies
652 are smoothed with a 12-month running average. Values given represent correlation coefficient
653 when compared to AGAGE observations and variance. The decay rate anomaly is calculated
654 from global mean CH_3CCl_3 values using equation (1) from Holmes et al., (2013), expressed as
655 a percentage of the typical decay with a 12-month smoothing. (c) As panel (b) but for model
656 runs NL_FTVW and NP_FTVW, along with RE_FTVW, RE_VTVW and RE_FTFW, and
657 correlation coefficients for comparison with NOAA observations. The model results are split
658 across panels (b) and (c) for clarity.



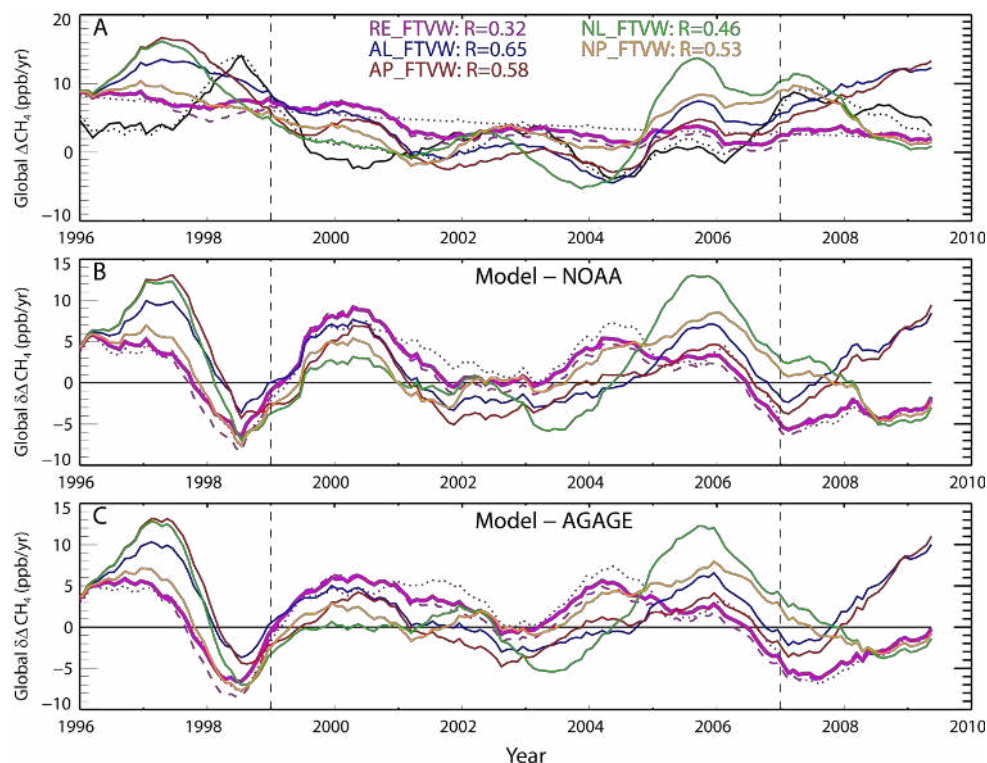
659

660 **Figure 3.** (Left) Observed mean surface CH_3CCl_3 (ppt) (black line) from (a) Mace Head
661 (AGAGE), (c) Cape Grim (AGAGE), (e) Mauna Loa (NOAA) and (g) South Pole (NOAA).
662 Also shown are results from five TOMCAT simulations with fixed temperatures and varying
663 winds (FTVW, for legend see Figure 2a). (Right): Surface CH_3CCl_3 decay rate anomalies at the
664 same station as the corresponding left column plot for observations (black), TOMCAT
665 simulations with varying winds (FTVW, solid coloured lines) and TOMCAT simulations with
666 fixed winds (FTFW, dotted lines). Comparisons at NOAA (AGAGE) stations show only
667 comparisons with runs using NOAA (AGAGE)-derived OH, along with runs RE_FTVW and
668 RE_FTFW in all panels.



669

670 **Figure 4.** (a, b, c and d) Deseasonalised surface CH₄ (ppb) from 4 NOAA sites (black solid line)
 671 from 1993 to 2009. Also shown are results from five TOMCAT 3-D CTM simulations with
 672 fixed temperatures and varying winds (FTVW, see **Table 2**). (e) Deseasonalised global mean
 673 surface CH₄ from NOAA (black solid) and AGAGE (black dashed) observations along with
 674 five TOMCAT simulations with different treatments of OH. (f) Same as (e) but for TOMCAT
 675 simulations using repeating OH (RE) and different treatments of winds and temperature. All
 676 panels use observation and model values which are smoothed with a 12-month running
 677 average. The shaded region marks the stagnation period in the observed CH₄ growth rate.



678

679 **Figure 5.** (a) The smoothed variation in the global annual CH₄ growth rate (ppb/yr) derived from NOAA (black solid) and AGAGE (black dashed) observations. Also shown are the smoothed growth rates from five TOMCAT 3-D CTM simulations with fixed temperatures and varying winds (FTVW, see Table 1). Values in legend give correlation coefficient between model run and NOAA observations. Also shown are results from runs RE_FTVW and RE_VTVW as a purple dotted line and dashed line, respectively (b) The difference in smoothed growth rate between TOMCAT simulations and NOAA observations shown in panel (a). (c) Same as (b) except using differences compared to AGAGE observations. The vertical dashed lines mark the start and end of the stagnation period in the observed CH₄ growth rate (1999 – 2006).