Sources and seasonality of atmospheric methanol based on tall tower measurements in the US Upper Midwest

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Abstract. We present over one year (January 2010–February 2011) of continuous atmospheric methanol measurements from the University of Minnesota tall tower Trace Gas Observatory (KCMP tall tower; 244 m a.g.l.), and interpret the dataset in terms of constraints on regional methanol sources and seasonality. The seasonal cycle of methanol concentrations observed at the KCMP tall tower is generally similar to that simulated by a global 3-D chemical transport model (GEOS-Chem, driven with MEGANv2.0 biogenic emissions) except the seasonal peak occurs ∼1 month earlier in the observations, apparently reflecting a model underestimate of emission rates for younger versus older leaves. Based on a source tracer approach, which we evaluate using GEOS-Chem and with multiple tracers, we estimate that anthropogenic emissions account for approximately 40% of ambient methanol abundance during winter and 10% during summer. During daytime in summer, methanol concentrations increase exponentially with temperature, reflecting the temperature sensitivity of the biogenic source, and the observed temperature dependence is statistically consistent with that in the model. Nevertheless, summertime concentrations are underestimated by on average 35% in the model for this region. The seasonal importance of methanol as a source of formaldehyde (HCHO) and carbon monoxide (CO) is highest in spring through early summer, when biogenic methanol emissions are high but isoprene emissions are still relatively low. During that time observed methanol concentrations account for on average 20% of the total CO and HCHO production rates as simulated by GEOS-Chem, compared to 12% later in the summer and 12% on an annual average basis. The biased seasonality in the model means that the photochemical role for methanol early in the growing season is presently underestimated.

1 Introduction

Methanol (CH₃OH) is one of the most abundant non-methane organic gases in the atmosphere, with concentrations ranging from hundreds of parts per trillion (pptv) in the remote troposphere to tens of parts per billion (ppbv) near surface sources (Singh et al., 1995, 2004; Karl et al., 2003; Millet et al., 2004; Schade and Goldstein, 2006; Jordan et al., 2009). It affects background and boundary layer photochemistry, and is an important source of carbon monoxide (CO) and formaldehyde (HCHO) (Riemer et al., 1998; Tie et al., 2003; Millet et al., 2006; Duncan et al., 2007; Choi et al., 2010). Here we present over one full year of methanol concentration measurements from a tall tower observatory in the US Upper Midwest, and interpret the data in terms of the information they provide on seasonal methanol sources and natural versus anthropogenic contributions.

Estimates of the global methanol source range from 75 to 350 Tg yr⁻¹ (Singh et al., 2000, 2004; Galbally and Kirshtein, 2002; Tie et al., 2003; Jacob et al., 2005; Millet et al., 2008; Stavrakou et al., 2011), with the largest fluxes from terrestrial plant growth (MacDonald and Fall, 1993; Fall and Benson, 1996; Singh et al., 2000; Karl et al., 2003; Harley et al., 2007; Folkes et al., 2008; Stavrakou et al., 2011). Other methanol sources include emissions from decaying plant matter (Warneke et al., 1999), biomass and biofuel burning (Holzinger et al., 1999; Andreae and Merlet, 2001), urban and industrial activities (de Gouw et al., 2005).
and atmospheric production (mostly from methane oxidation) (Madronich and Calvert, 1990; Tyndall et al., 2001). The marine biosphere appears to be a large gross source but an even larger gross sink of atmospheric methanol, so that overall it acts as a net sink (Singh et al., 1995; Heikes et al., 2002; Williams et al., 2004; Millet et al., 2008; Stavrakou et al., 2011). Gas-phase oxidation by OH is the other major methanol sink, with deposition playing a smaller but still significant role (Atkinson et al., 2006; Karl et al., 2010). The overall atmospheric life time for methanol is an estimated 5–12 days (Heikes et al., 2002; Tie et al., 2003; Singh et al., 2004; Stavrakou et al., 2011).

The atmospheric methanol budget includes substantial uncertainties, most notably in terms of the magnitude and distribution of biogenic emissions, which are large but remain poorly constrained. This in part reflects the scarcity of long-term measurements to constrain methanol emission fluxes and their seasonal variability.

In this paper we report atmospheric methanol measurements made using a proton transfer reaction mass spectrometer (PTR-MS) over one year (January 2010–February 2011) at the University of Minnesota tall tower Trace Gas Observatory (KCMP tall tower), a 244 m AmeriFlux tall tower in the US Upper Midwest (Fig. 1). Measurements are made at a sampling height of 185 m above ground level, providing a high-resolution signal with regional-scale footprint. The tower is located near the intersection of the major North American ecosystems (eastern deciduous forest, northern coniferous forest, agriculture, and western prairie), and is at times downwind of the Twin Cities metropolitan area (2010 population: 3.3 M), thus affording information on emissions of methanol and other reactive gases from some of the most important US landscapes as well as from anthropogenic sources. We apply the tall tower methanol observations with a suite of other chemical measurements, and in combination with a global 3-D chemical transport model (GEOS-Chem), to better understand seasonal methanol sources, natural versus anthropogenic contributions, and the resulting impact on atmospheric formaldehyde and CO.

2 Methods

2.1 Field site description

The University of Minnesota tall tower Trace Gas Observatory is a Minnesota Public Radio communications tower (KCMP 89.3 FM, 44.689° N, 93.073° W; 244 m height, 534 m above sea level), located approximately 25 km south of Minneapolis-St. Paul, MN, US (Fig. 1). Land cover within a 50 km radius of the tall tower consists of approximately 26 % crops, 18 % grassland, 18 % woodland, 27 % developed land, 6 % wetland, and 5 % water (Corcoran, 2009). May through October is the freeze-free growing season near the site with a 30 yr normal (arithmetic mean) temperature of 16.6 °C (min 11.2 °C, max 23.6 °C). The 30 yr normal annual precipitation is 88 cm, about 70 % of which occurs between May and October each year (Corcoran, 2009). Winter is cold (November to January 30 yr mean −8.2 °C, min −12.2 °C, max −3.3 °C) and the snow cover period can span November to March (Corcoran, 2009; NOAA-MRCC, 2011).

The tower was instrumented with a variety of micrometeorological instruments in April 2007, and is part of the AmeriFlux network (Sitename KCMP tall tower) (Griffis et al., 2010). A three-dimensional sonic anemometer-thermometer (CSAT3, Campbell Scientific Inc., USA) installed near the air sample inlet measures the three components of wind speed and temperature at 10 Hz for eddy flux measurements of CO2, H2O, and sensible heat. A detailed description of meteorological measurements at the site is given by Griffis et al. (2010).

Figure 2 shows the seasonal wind speed and direction observed during 2010 at the KCMP tall tower. Mean wind speeds for each season ranged between 5.5 and 7.5 m s−1. Winds during spring were most commonly from south to southeast, while in the summer southerly and north-westerly winds were most frequent. In the autumn and winter, winds were predominantly from the northwest. The footprint sampled by the KCMP tall tower therefore varies through the year, due to these shifting wind patterns (as well as seasonal changes in boundary layer mixing and photochemical lifetimes).

2.2 Volatile organic compound (VOC) measurements

We measure methanol and an ensemble of other VOCs (including isoprene and its first-generation oxidation products methyl vinyl ketone and methacrolein; acetone; acetonitrile; and C6-C9 aromatics) at the KCMP tall tower using PTR-MS (HS-PTR-MS, Ionicon Analytik, Austria). In this technique, air flows continuously through a drift tube reaction chamber containing H3O+ ions produced by hollow
During measurement, the PTR-MS steps continuously through a series of 15 protonated VOC masses (including $m/z$ 33 for methanol), with 2–20 s dwell time on each. Additional masses are monitored for diagnostic purposes at 200 ms dwell time each (e.g., $H_2O^+$, $H_2O^+(H_2O)$), for a total measurement cycle of 26 s (October 2009 to July 2010) or 170 s (after August 2010). Drift tube pressure is maintained at 2.0 to 2.3 mbar, voltage at 600 V, and temperature at 60°C.

For all species there is a nonzero background signal because of impurities in the system and/or the presence of unwanted ions from the ion source (de Gouw et al., 2003), so regular background checks under ambient humidity are performed every 2–5 h. Background signals are monitored for 5–10 measurement cycles by passing ambient air flow through a 0.95 cm × 1.27 cm (I.D. × O.D.) stainless steel tube filled with platinum bead catalyst (3 mm in diameter; Shimadzu Corp., Japan) and heated to 450°C. After subtracting the interpolated background signals, the raw data are post-processed following de Gouw et al. (2003) to account for humidity effects.

The PTR-MS is calibrated by dynamic dilution of multi-component standards (Air Environmental Inc., USA; stated accuracy ±5%) into zero air (generated as above). The standard cylinders were originally filled December 2008 and were re-analyzed in May 2011. VOC concentrations in the standards range from tens to hundreds of ppbv depending on the compound (202 ppbv for methanol). Calibration curves were performed automatically every 23 h through August 2010; this was subsequently decreased to 47 h because of the observed instrument stability. Under most conditions, the $R^2$ values for 6-point calibration curves are >0.995 for reported compounds, with the relative standard deviation of residuals <7%. Detection limits, defined as 3× the precision, are ∼220 pptv for methanol (5 s dwell time), ∼18 pptv for benzene (10 s dwell time), and ∼30 pptv for toluene (10 s dwell time). Typical sensitivities during calibration are ∼10.9 ncps ppbv$^{-1}$ for methanol, ∼10.2 ncps ppbv$^{-1}$ for benzene, and ∼13.2 ncps ppbv$^{-1}$ for toluene for a drift tube pressure of 2.2 mbar and a drift tube voltage of 600 V. The overall uncertainty of measurement for methanol, based on quadrature addition of the individual error sources (flow controllers, standard accuracy, calibration fit, standard error of the 30 min averages, etc.), is estimated at 10% on average (and in nearly all cases ∼20%).

All analyses presented here were performed using the open-source tools R (www.r-project.org) and openair (www.openair-project.org).

### 2.3 CO measurements

We apply concurrent CO measurements at the KCMP tall tower in our interpretation of the methanol data. CO (and $H_2$) are measured by gas chromatography (GC) with a reducing compound photometer (Peak Performer 1; Peak Laboratories LLC., USA). CO and $H_2$ eluting from the GC column

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**Fig. 2.** Wind rose plots showing wind speed as a function of direction during 2010 at the KCMP tall tower. MAM: March, April, May; JJA: June, July, August; SON: September, October, November; DJF: December, January, February.
(Unibeads 1S and MS 13X) pass directly into a heated mercuric oxide bed (265°C). During the process, mercury vapour is liberated and subsequently measured via UV light absorption in the photometer cell. Compressed medical grade air is used as carrier gas. Multi-point calibrations are carried out daily by dynamic dilution of a ppmv-level standard (Scott Specialty Gases, USA; stated accuracy ±2%) into zero air generated by passing ambient air through a Sofnecat catalyst (Molecular Products Inc., USA) bed heated to 60°C. Under normal conditions, the $R^2$ values for 6-point calibration curves are >0.999 for both CO and H$_2$, with the relative standard deviation of residuals <3%. The detection limits for CO and H$_2$ are 300 and 800 pptv respectively for 3 min measurements.

### 2.4 GEOS-Chem chemical transport model

We use the GEOS-Chem 3-D model of atmospheric chemistry (Bey et al., 2001) as a tool to interpret the tall tower observations in terms of constraints on methanol source types. GEOS-Chem (v8, http://www.geos-chem.org) is a global Eulerian chemical transport model driven by assimilated meteorological observations from the NASA Goddard Earth Observing System (GEOS-5). The meteorological fields have $0.5^\circ \times 0.667^\circ$ horizontal resolution and 72 vertical layers. For our work here we reduce the spatial resolution to $2^\circ \times 2.5^\circ$ and 47 layers, of which 14 are below 2 km altitude, and use a 15 min transport timestep. For methanol, we apply here the simulation implemented in GEOS-Chem by Millet et al. (2008), with updates outlined below, while for CO and other tracers we apply the full-chemistry simulation described in Millet et al. (2010).

Biogenic emissions of methanol and other VOCs are computed online in GEOS-Chem using the Model of Emissions of Gases and Aerosols from Nature (MEGANv2.0; Guenther et al., 2006). This is change over the earlier work by Millet et al. (2008), which used a simple net primary productivity (NPP) based parameterization for estimating emission. Here, fluxes are estimated for each model grid square as a sum of contributions from four plant functional types (PFTs: broadleaf trees, needleleaf trees, shrubs, and herbaceous plants [crops + grasslands]):

$$E = \gamma \sum_{i=1}^{4} \epsilon_i \chi_i$$  

(1)

where $\epsilon_i$ is the canopy emission factor for PFT $i$ with fractional coverage $\chi_i$. Here we use $\epsilon_i = 800 \mu g m^{-2} h^{-1}$ for the four PFTs. The overall activity factor $\gamma$ is calculated from a set of individual activity factors accounting for the effects of temperature, photosynthetic photon flux density (PPFD), leaf age, and leaf area index (LAI) on emissions. Activity factors for LAI, PPFD and leaf age are computed as described by Stavrakou et al. (2011) and Guenther et al. (2006), but using a light-dependent fraction of 0.75 and relative emission rates of 3.0, 2.6, 0.85, and 1.0 for new, young, mature, and old leaves. Leaf age fractions are parameterized based on monthly changes in leaf area index as described by Guenther et al. (2006). The higher emission rate for old relative to mature leaves represents the methanol flux from dead and decaying foliage (Warneke et al., 1999). The temperature dependence of methanol emission is modelled using an exponential function,

$$\gamma_T = \exp[\beta (T - 303)]$$  

(2)

with $\beta = 0.08$ and $T$ the surface air temperature (K). The global annual flux of methanol from terrestrial plants for this simulation is 95 Tg in 2010. This is significantly lower than the base-case simulation presented by Millet et al. (2008) using the NPP-based parameterization of biogenic emissions (145 Tg y$^{-1}$), and comparable to their optimized estimates based on a reduced biogenic source (72–89 Tg y$^{-1}$).

Anthropogenic emissions over the US are from the US EPA NEI-99 (EPA, 1999), updated by Hudman et al. (2007; 2008) to account for recent CO and NOx emission reductions. The anthropogenic methanol flux is estimated based on a methanol:CO molar emission ratio of 0.012 (Goldan et al., 1995; de Gouw et al, 2005; Millet et al., 2005; Warneke et al., 2007). Methanol emissions from fires are from the Global Fire Emissions Database version 2 (GFEDv2, van der Werf et al., 2006) for the year 2008 (the most recent year for which these emissions are available) and measured species:species pyrogenic emission ratios (Andreae and Merlet, 2001; Andreae, unpublished data, 2006). Other methanol sources (biofuel burning, photochemical production, gross ocean emission) and sinks (oxidation by OH, gross ocean uptake) are implemented as described in our earlier work (Millet et al., 2008).

GEOS-Chem is not necessarily expected to fully resolve the fine-scale concentration fluctuations observed at the KCMP tall tower due to the $2^\circ \times 2.5^\circ$ model resolution. However, as we will show it does provide a useful way to leverage the observations in testing some basic assumptions in terms of how methanol sources are represented in current models.

### 3 Results and discussion

#### 3.1 Seasonality of methanol abundance and emissions in the US Upper Midwest

Figure 3 shows hourly mean methanol mixing ratios observed from January 2010 through February 2011 at the KCMP tall tower. Also shown are concurrent measurements of benzene, toluene, CO, and air temperature. For all compounds statistical outliers (>0.98 quantile for each month) have been removed prior to plotting. Table 1 shows the corresponding seasonal statistics for methanol along with data
from related sites where measurements covered multiple seasons.

Methanol is the most abundant VOC measured at the KCMP tall tower, with a 2010 annual mean mixing ratio of 3.8 ppbv (median 3.0 ppbv). The highest seasonal mixing ratios of methanol are observed during summer, reflecting biogenic emissions in the warm season. Lowest concentrations are seen during winter; spring and autumn show intermediate levels of similar magnitude (Table 1). This seasonality is broadly consistent with the few other in-situ observations spanning a full annual cycle (Table 1; Schade and Goldstein, 2006; Jordan et al., 2009; Lappalainen et al., 2009).

We also see in Fig. 3 some wintertime events with strongly elevated concentrations of methanol (as well as of benzene, toluene, and CO). These reflect local anthropogenic emissions within the shallow wintertime mixed layer, transported from the Twin Cities by northwesterly winds (predominant at that time of year).

Prior methanol observations in the US Midwest are sparse (Table 1). They include urban measurements in Pittsburgh PA (on average ∼1.6–2.4 times higher than observed at the KCMP tall tower) (Millet et al., 2005) and measurements over a mixed hardwood forest in northern Michigan (∼1.2–2.0 times higher than KCMP) (Karl et al., 2003). Table 1 shows that the methanol concentrations at the KCMP tall tower are generally in the range observed at rural sites elsewhere in the US and abroad.

Figure 3 also shows the methanol concentrations simulated by GEOS-Chem at the KCMP tall tower (red). The model consistently underestimates the observed concentrations year-round, with ∼35% low bias during summer. This can be compared to other recent findings by Millet et al. (2008) and Stavrakou et al. (2011). Using an NPP-based parameterization of biogenic emissions, and after scaling down this biogenic flux to account for a source overestimate, Millet et al. (2008) found that their simulation still had a slight high bias in the eastern US and a low bias in the west. Likewise, on the basis of satellite observations from the IASI sensor, Stavrakou et al. (2011) found that MEGAN overpredicts emissions in the eastern US while underpredicting in the west. More work is needed to better quantify methanol emission rates in different ecosystems and regions, but the observations reported here suggest a source underestimate for the US Upper Midwest.

The seasonality of methanol concentrations as seen in Fig. 3 is clearly distinct from that of CO, benzene, and toluene, which are predominantly emitted from anthropogenic sources (CO is also produced in the atmosphere from VOC oxidation, as we discuss later). Mixing ratios for these species are low in the summer and peak during winter, reflecting the seasonal change in OH radical concentrations. Below (Sect. 3.2) we will use these anthropogenic compounds as tracers to evaluate the relative importance of biogenic versus anthropogenic methanol sources for the region sampled by the KCMP tall tower.

Current understanding of the seasonality of methanol emissions is based on the premise that pectin biosynthesis during plant and leaf growth is a major source of atmospheric methanol, resulting in higher emissions from new and young leaves in spring and early summer as compared to mature and old leaves during later summer and autumn (MacDonald and Fall, 1993; Fall and Benson, 1996; Galbally and Kirstine, 2002; Fall, 2003; Karl et al., 2003). Long-term measurements from the KCMP tall tower, with its source footprint influenced by multiple PFTs, provide an opportunity to test how well this seasonal dependence is represented in current models.

Figure 4 shows the normalized seasonality in methanol abundance observed at the KCMP tall tower (black) compared to that simulated by GEOS-Chem (red). The simulated seasonality is generally similar to the observations, peaking in summer and with a winter minimum. However, the modeled seasonal amplitude is stronger than observed (simulated summer:winter ratio of 3 versus an observed value of 2). As we discuss later, this in part reflects a model underestimate of the anthropogenic methanol source for this region.

As Fig. 4 shows, the onset of biogenic methanol emissions is well-captured by the model, with measured and simulated concentrations rising in parallel during mid-April and May, and in concert with increasing regional leaf area. However, the seasonal peak in the model is phase-shifted relative to the observations, occurring substantially later in the summer (second week of August versus mid-July in the observations).
Table 1. Seasonality of atmospheric methanol abundance (ppbv) observed at the KCMP tall tower in the US Upper Midwest and comparison with other sites.

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</thead>
<tbody>
<tr>
<td>spring</td>
<td>mean 4.0, median 3.2, 10^4–9 x 10^4 (ppbv)</td>
<td>median 10.7</td>
<td>mean 8</td>
<td>median 10.9</td>
<td>median 1.3</td>
<td>mean 2.2</td>
<td>day-time median ~0.5</td>
</tr>
<tr>
<td>summer</td>
<td>7.0</td>
<td></td>
<td>10</td>
<td>2</td>
<td>2.7</td>
<td>3.2</td>
<td>~2</td>
</tr>
<tr>
<td>autumn</td>
<td>3.2</td>
<td></td>
<td>4</td>
<td>2.0</td>
<td>1.1</td>
<td>3.2</td>
<td>~1</td>
</tr>
<tr>
<td>winter</td>
<td>2.1</td>
<td></td>
<td>3.8</td>
<td>2</td>
<td>1.0</td>
<td>1.1</td>
<td>~0.4</td>
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<tr>
<td>land cover</td>
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<td>mixed forest</td>
<td>needleleaf</td>
<td>crop and mixed forest</td>
<td>urban</td>
<td>needleleaf</td>
</tr>
<tr>
<td>measurement height</td>
<td>185 m</td>
<td></td>
<td>34 m</td>
<td>12 m</td>
<td>12 m</td>
<td>14 m</td>
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1 Seasons here are: spring (March–May), summer (June–August), autumn (September–November), winter (December–February)

2 Pittsburgh, PA, USA; Millet et al. (2005)

3 University of Michigan Biological Station, MI, USA; Karl et al. (2003)

4 UC Berkeley Blodgett Forest Research Station, CA, USA; Schade and Goldstein (2006)

5 Thompson Farm, NH, USA; Jordan et al. (2009)

6 Zurich, Switzerland; Legreid et al. (2007)

7 SMEAR-II, Hyttiala, Finland; Lappalainen et al. (2009)

Later (Sect. 3.3) we assess how errors in the assimilated temperatures used to drive GEOS-Chem and MEGAN affect the model:measurement comparisons, but the temperature bias is small and cannot explain this seasonality offset. It appears likely that MEGAN is underestimating the relative methanol emission rate for younger versus older leaves, at least for this part of North America.

3.2 Anthropogenic versus biogenic sources of methanol

In this section we apply the KCMP tall tower measurements to evaluate the year-round importance of anthropogenic methanol emissions in the US Upper Midwest. Globally, anthropogenic methanol emissions are uncertain (2–11 Tg yr^{-1}) and are thought to make up only a small fraction of the total flux (2–5 %; Singh et al., 2000; Galbally and Kirstine, 2002; Heikes et al., 2002; Jacob et al., 2005; Millet et al., 2008; Stavarakou et al., 2011). On a regional scale, and in winter, the contribution from urban and industrial sources can be more substantial.

In winter, methanol concentrations at the KCMP tall tower are well-correlated with CO, benzene, and toluene (R = 0.67, 0.54, and 0.65, respectively during December-February), providing a fingerprint of regional anthropogenic emissions in the absence of any significant biogenic sources (or of photochemical production in the case of CO). The corresponding methanol:tracer major axis slopes are 0.034 (0.032–0.035) ppbv ppbv^{-1} for CO, 21.3 (20.2–22.4) ppbv ppbv^{-1} for benzene, and 26.0 (24.2–27.9) ppbv ppbv^{-1} for toluene (values in parentheses indicate 95 % confidence intervals, CI). For this analysis we remove data above the 0.8 quantile to ensure a regionally representative signal, and subtract a 7 day running 0.1 quantile to eliminate any influence from changing atmospheric background concentrations. No statistical difference in the slopes is seen for weekdays versus weekends, and no clear enhancements of acetonitrile
(a biomass burning tracer) were observed during winter. We therefore treat the observed methanol:tracer enhancement ratios as regional urban emission ratios, including contributions from fossil fuel combustion, industrial activity, and solvent use. The observed wintertime methanol:CO slope can be compared to previously reported methanol:CO enhancement ratios which range from 0.004 to 0.056 ppbv ppbv$^{-1}$ (Goldan et al., 1995; Singh et al., 1995; Salisbury et al., 2003; Jacob et al., 2005; Warneke et al., 2007; and references therein). The wintertime methanol:CO enhancement ratio observed at the KCMP tall tower is a factor of 2.8 higher than the global emission ratio employed in GEOS-Chem (0.012 ppbv ppbv$^{-1}$; Millet et al., 2008), indicating a significantly larger anthropogenic methanol flux for this area.

We apply here the observed tracer-tracer correlations to diagnose the anthropogenic methanol contribution $f_{\text{ANTH}}$ throughout the year as:

$$f_{\text{ANTH}} = \frac{E_X[X]_t}{[\text{CH}_3\text{OH}]_t}$$

(3)

where $E_X$ is the methanol enhancement ratio relative to tracer X (CO, benzene, or toluene) derived from the wintertime data, $[X]_t$ is the observed enhancement of X at time $t$, and $[\text{CH}_3\text{OH}]_t$ is the observed methanol abundance at time $t$. Potential biases in this method can arise from the assumption of a constant emission ratio, or because of differing atmospheric lifetimes for the compound of interest and the tracer. The secondary CO source from VOC oxidation (Hudman et al., 2008) can also complicate interpretation. We therefore evaluate the robustness of the approach in two ways. First, we apply it to the CO and methanol concentrations simulated by GEOS-Chem and compare the result with the actual anthropogenic methanol abundance in the model, which is known. Second, we assess the consistency of the findings using multiple anthropogenic tracers (CO, benzene, and toluene) in the observations. These analyses provide a means of estimating the uncertainty in the results.

Figure 5 (top panel) shows the source tracer approach applied to simulated methanol and CO concentrations from GEOS-Chem. The anthropogenic fraction $f_{\text{ANTH}}$ estimated using the methanol:CO correlation agrees well with the actual model value: major axis regression of the calculated versus actual model $f_{\text{ANTH}}$ yields a slope of 0.95 (95 % CI: 0.80–1.09) with $R = 0.83$ (Fig. 6). Likewise, applying CO, benzene and toluene as tracers to estimate $f_{\text{ANTH}}$ in the observations gives consistent results to within ±30 % (Fig. 6). Both tests thus provide support for the use of this method in interpreting the observations. Figure 5 (middle panel) shows the year-round anthropogenic contribution to the observed methanol abundance at the KCMP tall tower estimated on the basis of measured CO (black), benzene (red), and toluene (blue). In all cases we see a pronounced and consistent seasonal cycle, where regional anthropogenic sources account for approximately 10 % of observed levels during summer, and up to 70 % (on average 40 %) during winter. The seasonality is driven by the changing importance of biogenic emissions, rather than by any strong variation in the absolute anthropogenic flux itself: the bottom panel of Fig. 5 shows that the weekly median anthropogenic contribution is between 0.2–1.5 ppbv year-round. The source tracer approach does not provide a straightforward way to parse the residual, non-anthropogenic methanol abundance: biogenic tracers, such as isoprene, tend to be short-lived, and emissions may not correlate particularly well with those of methanol. In the GEOS-Chem simulation the corresponding non-anthropogenic fraction is almost exclusively biogenic during summer (with nearly 80 % from crops and grasslands), and during winter represents a mixture of transported biogenic emissions, biomass and biofuel burning, and photochemical production.

We repeated this analysis with a filter applied (30° < wind direction < 300°) to exclude air masses transported directly from Minneapolis-St. Paul; this resulted in only minor changes to the calculated anthropogenic component (to approximately 13 % of the total in summer and 44 % in winter).

### 3.3 Temperature control on methanol emissions

Biogenic methanol emissions increase exponentially with temperature (Harley et al., 2007; Folkers et al., 2008). In this section, we use the tall tower measurements during summer, when (as we have shown above) the land biosphere
is the dominant source, to test the temperature-flux relationship used in present models.

Figure 7 shows daytime summer (June–August, 07:00–18:00 LST) methanol concentrations as a function of temperature. A wind direction filter (30° < wind direction < 300°) is applied to exclude urban impacts from Minneapolis-St. Paul. An exponential fit to the observations in Fig. 7 yields a coefficient of 0.087 (95% CI: 0.082–0.094), in the range of previously published values (0.068–0.094) (Riemer et al., 1998; Schade and Goldstein, 2006; Lappalainen et al., 2009). The observed coefficient is also statistically consistent with the corresponding value from a fit of the simulated methanol abundance to the assimilated temperature used to drive GEOS-Chem (0.083, 95% CI: 0.078–0.088), and with the β value used in the emission algorithm (0.08; Eq. 2). We conclude that the model biogenic temperature response is accurate to within the constraints provided by the KCMP tall tower observations.

We analyzed the anthropogenic tracers CO, benzene, and toluene in a similar manner, and found no positive correlation with temperature (R ≤ 0.1), so that the strong observed methanol:temperature correlation is mainly providing information relevant to the biogenic source, rather than vertical mixing or evaporative emissions.

The fact that the simulated methanol concentrations are biased low by ~35% in the summer compared to the observations (Fig. 3) thus suggests that some emission parameters, other than the temperature dependence, are not adequately represented for this region in GEOS-Chem and MEGAnv2.0. The GEOS-5 temperatures used in GEOS-Chem have a slight low bias (on average −0.9 °C) compared to the observations, but we find that correcting this would increase model concentrations by only ~10%, significantly less than the model-measurement mismatch.
3.4 Methanol as a source of HCHO and CO

HCHO and CO are produced sequentially during the oxidation of methanol by OH and with essentially quantitative yield. The production rate of HCHO ($P_{\text{HCHO}}$) and CO ($P_{\text{CO}}$) from methanol oxidation is:

$$P_{\text{CO}} \approx k_{\text{OH}}Y[\text{OH}][\text{CH}_3\text{OH}]$$  \hspace{1cm} (4)

with $k_{\text{OH}}$ the rate constant for reaction of methanol with OH (Atkinson et al., 2006) and $Y$ the HCHO yield (1.0). The production rate of CO from methanol is essentially equivalent to the production rate of HCHO, since the HCHO lifetime is ~10 times shorter than that of methanol (Atkinson et al., 2006), and depositional sinks for HCHO are small (Dufour et al., 2009).

Figure 8 shows the CO and HCHO production rates computed from the methanol measurements at the KCMP tall tower. We plot these also as a fraction of the total CO and HCHO production rates simulated by GEOS-Chem (right panel), to gauge methanol’s importance as a source of these two important compounds in the US Upper Midwest (OH concentrations in Eq. (4) are likewise taken from the model).

We see from Fig. 8 that $P_{\text{HCHO}}$ and $P_{\text{CO}}$ estimated from the methanol observations at the KCMP tall tower range from $1.4 \times 10^4$ to $8.1 \times 10^5$ molec cm$^{-3}$ s$^{-1}$, with an annual mean of $2.3 \times 10^5$ molec cm$^{-3}$ s$^{-1}$. Observed methanol concentrations thus account for between 4 and 25 %, and on average 12 %, of the total CO and HCHO production rates as simulated by the GEOS-Chem model. Methanol’s contribution in late summer (~12 %) is comparable to that seen at Blodgett Forest Research Station in California (13–18 % during daytime in late summer) (Choi et al., 2010). At the KCMP tall tower, the highest fractional contribution to tropospheric HCHO and CO (~20 %) occurs during April, May, and June, at a time when methanol emissions and concentrations are high, but prior to the seasonal peak in isoprene emissions later in the summer. Thus the seasonal phase shift for methanol compared to isoprene manifests in terms of a pronounced photochemical role early in the growing season. This role is underestimated in the model because the simulated seasonal peak occurs too late in the summer.

4 Conclusions

We applied measurements of a suite of atmospheric VOCs and CO over one year (January 2010–February 2011) from the KCMP tall tower in the US Upper Midwest to test our understanding of methanol sources and seasonal impacts as represented in a current chemical transport model.

Measured methanol concentrations at the KCMP tall tower ranged from 0.1 ppbv to 25 ppbv with an annual mean of
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