New emission factors for Australian vegetation fires measured using open-path Fourier transform infrared spectroscopy – Part 2: Australian tropical savanna fires

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Abstract. Savanna fires contribute approximately 40–50% of total global annual biomass burning carbon emissions. Recent comparisons of emission factors from different savanna regions have highlighted the need for a regional approach to emission factor development, and better assessment of the drivers of the temporal and spatial variation in emission factors. This paper describes the results of open-path Fourier transform infrared (OP-FTIR) spectroscopic field measurements at 21 fires occurring in the tropical savannas of the Northern Territory, Australia, within different vegetation assemblages and at different stages of the dry season. Spectra of infrared light passing through a long (22–70 m) open-path through ground-level smoke released from these fires were collected using an infrared lamp and a field-portable FTIR system. The IR spectra were used to retrieve the mole fractions of 14 different gases present within the smoke, and these measurements used to calculate the emission ratios and emission factors of the various gases emitted by the burning. Only a handful of previous emission factor measures are available specifically for the tropical savannas of Australia and here we present the first reported emission factors for methanol, acetic acid, and formic acid for this biome. Given the relatively large sample size, it was possible to study the potential causes of the within-biome variation of the derived emission factors. We find that the emission factors vary substantially between different savanna vegetation assemblages; with a majority of this variation being mirrored by variations in the modified combustion efficiency (MCE) of different vegetation classes. We conclude that a significant majority of the variation in the emission factor for trace gases can be explained by MCE, irrespective of vegetation class, as illustrated by variations in the calculated methane emission factor for different vegetation classes using data sub-set by different combustion efficiencies. Therefore, the selection of emission factors for emissions modelling purposes need not necessarily require detailed fuel type information, if data on MCE (e.g. from future spaceborne total column measurements) or a correlated variable were available.

From measurements at 21 fires, we recommend the following emission factors for Australian tropical savanna fires (in grams of gas emitted per kilogram of dry fuel burned), which are our mean measured values:

- 1674 ± 56 g kg⁻¹ of carbon dioxide
- 87 ± 33 g kg⁻¹ of carbon monoxide
- 2.1 ± 1.2 g kg⁻¹ of methane
- 0.11 ± 0.04 g kg⁻¹ of acetylene
- 0.49 ± 0.22 g kg⁻¹ of ethylene
- 0.08 ± 0.05 g kg⁻¹ of ethane
- 1.57 ± 0.44 g kg⁻¹ of formaldehyde
- 1.06 ± 0.87 g kg⁻¹ of methanol
- 1.54 ± 0.64 g kg⁻¹ of acetic acid
- 0.16 ± 0.07 g kg⁻¹ of formic acid
- 0.53 ± 0.31 g kg⁻¹ of hydrogen cyanide
- 0.70 ± 0.36 g kg⁻¹ of ammonia.

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similar techniques are used to characterise the emissions from Australian temperate forest fires.

1 Introduction

Estimates of mean annual carbon emissions from worldwide savanna and grassland fires amount to ~1 Pg C per year (2000–2009), or approximately a mean of 40–50% of total global annual biomass burning carbon emissions (van der Werf et al., 2010). In terms of the Australian savanna, the direct gross carbon emissions from fires have a similar, if not greater, magnitude than those from Australian fossil fuel combustion (Haverd et al., 2013). However, in tropical Australia, net carbon emissions from fires are considered negligible as the carbon lost to burning is approximately balanced by the carbon sequestered during regrowth (Haverd et al., 2013). Akagi et al. (2011) present summaries of emission factors (EFs) that can be used to convert estimates of total biomass (or carbon) consumed in fires into estimates of individual trace gas species release. Emissions of non-CO$_2$ gases from savanna fires, such as nitrous oxide and methane, whose emissions are not sequestered by regrowth, comprise 2–4% of total Australian national greenhouse gas emissions (DCCEE, 2010), and are included in the accountable greenhouse gas inventory. At state or territory scales, pyrogenic emissions of CO$_2$, CH$_4$, and N$_2$O can be far more significant, with emissions from fires comprising nearly 50% of the Northern Territory’s accountable carbon emissions (DCCEE, 2010).

A clear understanding of the emissions characteristics of Australia’s tropical savannas is important for regional emissions modelling and for associated greenhouse gas accounting schemes. As with the fire regimes of other savanna regions, that in Australia’s tropical savanna is heavily influenced by human fire use and fire management (Garde, 2009; Cook et al., 2010). Currently, policy in the Northern Territory of Australia aims to reduce overall savanna fire emissions by promoting early dry-season (pre mid-July) fires which consume less fuel and create a network of fire breaks that reduce the extent of late dry-season (post mid-July) fires, reducing overall fire frequency (Cook and Meyer, 2009; Russell-Smith et al., 2009). Given that the seasonality of the emission factor for methane in Australian tropical savannas has recently been shown to be different to that observed in African savannas (Meyer et al., 2012), it is evidently important to make regionally specific measurements of emission factors so that the most accurate assessments of the effectiveness of such policies, and the impact of fire management on greenhouse gas and reactive gas emissions, can be provided.

Despite the prevalence of savanna fires on the Australian continent, with approximately 400 000 km$^2$ of tropical savannas (~15% of global tropical savannas) and 150 000 km$^2$ of arid savannas burning annually (Russell-Smith et al., 2007; Meyer et al., 2012), the focus of most savanna emissions research has been on Africa (e.g. Bertschi et al., 2003a; Yokelson et al., 2003; Sinha et al., 2003; Keane et al., 2006). Only Hurst et al. (1994a, b), and Shirai et al. (2003) have published emission factors from Australian savannas, with most measurements made using airborne campaigns. Most recently, Meyer et al. (2012) have published emission factors for methane (CH$_4$) and nitrous oxide (N$_2$O) from Australian savannas using ground-based sampling methods.

Careful interpretation of data collected from these various methods is required to ensure the accurate characterisation of fire-averaged emissions (Akagi et al., 2011). Beyond these methodological complications are the factors that lead to the natural variability of emissions (e.g. interactions between meteorology, seasonality and combustion efficiency). As the recent study by van Leeuwen and van der Werf (2011) shows, within-biome temporal and spatial variations in emission factors are substantial, and despite their attempts to explain these using relationships between emission factors and a number of environmental variables (tree cover density, vegetation greenness, temperature, precipitation, length of dry season), correlations between CO, CH$_4$ and CO$_2$ EFs, and modified combustion efficiency (MCE), were rather low, with environmental variables explaining 33, 38, 19, and 34% of the variation in the reported emissions metrics, respectively. Van Leeuwen and van der Werf (2011) posit some potential reasons for the poor explanation of within-biome variability, including different measurement techniques for EFs, assumptions used to weight EFs by the amount of flaming and smouldering combustion and uncertainties in the environmental variables themselves. Explaining the large natural variability of EFs that is so pronounced in the findings of Akagi et al. (2011) remains an important challenge for biomass burning emissions science.

Through measurements of a large number of fires during multiple campaigns at different stages of tropical Australia’s dry season, this work aims to expand the current emission factors inventory for Australian savannas, whilst also exploring the nature and causes of the natural variability of fire emissions within this ecosystem. Long open-path Fourier transform infrared (OP-FTIR) spectroscopy was used for field measurements of trace gases emitted from 21 different experimental fires or wildfires in the tropical savanna of the Northern Territory, Australia. A thorough discussion of the methodology and emission factors for Australian temperate forest fires can be found in our partner paper, Paton-Walsh et al. (2014).

2 Methods

2.1 The open-path FTIR system and deployment

FTIR systems can simultaneously record infrared spectra at a high enough spectral resolution and across a wide wavenumber range, such that estimates of the path-averaged mole
fraction of many of the trace gases present in the path can be retrieved via spectral analysis. The OP-FTIR system operated by King’s College London consists of a MIDAC Corporation M2000 Series FTIR spectrometer (0.5 cm\(^{-1}\) resolution) equipped with a Stirling-cycle cooled mercury-cadmium-telluride (MCT) detector, and fitted with a MIDAC custom-built 76 mm Newtonian telescope. Unlike the monostatic pre-modulated source system used by Paton-Walsh et al. (2014), the MIDAC system employed here operates using a bi-static configuration, where the FTIR spectrometer is operated in “active” mode, viewing a remotely located infrared source. Here, the infrared source consists of a 12 V silicon carbide glowbar operating at 1500 K fitted in front of a 20 cm diameter gold-plated aluminium collimator. With a maximum optical path distance of 2 cm, each scan of the spectrometer mirror takes \(\sim 1.2\) s. To improve the signal-to-noise ratio, it is advantageous to average (co-add) a number of spectral measurements prior to analysis. The MIDAC system and its use to characterise biomass burning emissions via long OP-FTIR spectroscopy is detailed in Wooster et al. (2011), whilst an accuracy assessment with respect to the retrieval of greenhouse gas mole fractions from the collected spectra was undertaken by Smith et al. (2011).

A mixture of purposely lit “experimental” plots and wildfires were studied. At most plots, the OP-FTIR path was positioned on a road, track, or fire-break downwind of the fire plot with pathlengths ranging from 22 to 70 m (Table 1). There were, however, some exceptions. For some plots, where there was no safe location downwind of the fire plot, the OP-FTIR path was positioned over the plot, with smoke passing between the spectrometer and IR source before or after the fire front passed through this path. Spectra collected during the time in which the fire was burning within the open path were affected by infrared emission from the flames; this effect precludes the retrieval of gas mole fractions from spectra due to the superposition of emission and absorption lines in the spectral signatures (one disadvantage of the bi-static OP-FTIR configuration). Consequently, all spectra affected by thermal emission from hot flames were removed prior to the analysis stage. For other plots, and especially for the wildfire measurements, it was not possible to position a path beside or across the fire plot. In these situations, either the IR source, or both the spectrometer and the IR source, were positioned within the fire plot. To protect the equipment from the fire, the equipment was positioned within natural firebreaks (rocky outcrops, areas of low biomass) or artificial firebreaks (raked vegetation) with the OP-FTIR path penetrating into the fire plot, or being entirely enclosed by the fire plot (Fig. 1).

For the experimental plots, it was possible to assemble the OP-FTIR equipment prior to each fire. Burn-offs and wildfires, by their nature, are unpredictable; identification of a position for the OP-FTIR path and assembly of the equipment took place whilst the fire was burning upwind. In one case (at the Adelaide River Station), it was possible to follow the fire along a track, assembling and re-assemblying paths as the fire front passed each path location. Given that most plots were small (< 1 Ha) and in some cases burned quickly, spectra were collected using four co-added scans to increase signal-to-noise yet still allow for a high measurement frequency (generating approximately one measured spectrum every 5 s). Observations of atmospheric temperature and pressure (parameters required for the trace gas retrieval analysis) were made using a co-located WXT510 weather station (Vaisala Oyj, Helsinki, Finland) at the Arnhem Land campaigns. For the Darwin region campaign, temperature and pressure were taken from a Bureau of Meteorology (Australian Government) weather station in Batchelor (located within 60 km of all of the burn sites).

### 2.2 Quantitative analysis of infrared spectra

For measurements that use a bi-static approach, where the FTIR spectrometer is used to measure a remote collimated non-modulated IR source, as described here, the measured spectrum includes radiation that has originated from the IR source \(I_{\text{SOURCE}}\), but also from instrument self-emission and from other ambient temperature sources \(I_{\text{AMB}}\) within the field of view of the receiving telescope. At the higher wave-number (shorter wavelength) regions used to retrieve gases such as CO\(_2\), CO, CH\(_4\), H\(_2\)CO (2000–3100 cm\(^{-1}\)), see Fig. 2), the ambient signal contribution is almost non-existent. However, at wave-number regions below \(\sim 1500\) cm\(^{-1}\), where gases such as C\(_2\)H\(_4\), HCOOH, CH\(_3\)OH and NH\(_3\) are retrieved, the contribution from ambient sources becomes more important to account for. This ambient signal component introduces a zero-baseline offset to the measured spectra which needs to be removed prior to any trace gas retrieval as this component contains no information related to trace gas absorption (Müller et al., 1999). When \(I_{\text{AMB}}\) is small (at wave numbers > 1500 cm\(^{-1}\)) the effect on measured
transmittance is relatively minor. However, higher levels of $I_{\text{AMB}}$ at lower wave numbers can significantly alter the apparent transmittance ($T_{\text{app}}$):

$$T_{\text{app}} = \frac{I}{I_{\text{SOURCE}} + I_{\text{AMB}}}.$$  (1)

To correct for this zero-baseline offset, it is necessary to make a measurement of the ambient spectrum (e.g. a spectral measurement with the active IR source turned off) and subtract this spectrum from the active measurement of the active IR source, making slight adjustments for any MCT detector non-linearities in accordance with Müller et al. (1999) to ensure that the resultant “difference” spectra exhibit zero signal. Nevertheless, measurements of ambient spectra were made both before and after a fire where possible, to check for any large changes in the ambient background spectrum. Nevertheless, measurements of ambient spectra were made both before and after a fire where possible, to check for any large changes in the ambient spectrum during the course of a fire.

Trace gas mole fractions were retrieved from all open-path FTIR spectra (after subtracting the background spectrum) using the Multiple Atmospheric Layer Transmission (MALT) model, as described in Part 1 of this paper (Paton-Walsh et al., 2014). MALT includes a forward model and a parameter optimisation routine, and has been used previously for retrieving the mole fraction of biomass burning compounds from open-path measurements of smoke in Australia (Griffith et al., 1991; Meyer et al., 2012), and in the United States (Akagi et al., 2014). Working in African savannah, Wooster et al. (2011) used an alternative retrieval code that again used similar physical principles to MALT. A full description of MALT can be found in Griffith (1996).

There are a few differences between the retrieval parameters used here and those used in Part 1 of this paper. Due to the higher spectral resolution of the MIDAC spectrometer, a resolution of 0.5 cm$^{-1}$ (fixed; 0.96 cm$^{-1}$ in Part 1) was used here with a field of view of 20 milli-radians (fixed: 22 milli-radians in Part 1); here we apply a triangle apodisation function (Hamming in Part 1). Phase error and
wave-number shifts were fitted in accordance with Paton-Walsh et al. (2014).

2.3 Spectral regions and parameters for quantitative spectral analysis

A set of standardised spectral windows and retrieval parameters are outlined in Part 1 of this paper (Table 1 in Paton-Walsh et al., 2014). The higher spectral resolution and greater wave-number range (700–4000 cm\(^{-1}\)) of the MIDAC spectrometer and detector deployed for the measurements presented here allowed for the additional detection of HCN and \(\text{C}_2\text{H}_2\) in the 710–760 cm\(^{-1}\) region, beyond what was detectable using the spectrometer deployed by Paton-Walsh et al. (2014). A full set of example spectral fits to our 0.5 cm\(^{-1}\) spectra is presented in Fig. 2. Due to the underlying spectral response of the MIDAC spectrometer, a higher-order background polynomial was required for the HCOOH retrieval window (Fig. 2f); here we use a 3rd-order polynomial, rather than a 2nd-order polynomial as indicated in Table 1 in Paton-Walsh et al. (2014).

2.4 Fire-averaged emission factors and probing natural variability

Individual fire emission factors were calculated using the method outlined in Part 1 (Paton-Walsh et al., 2014; Sect. 4). Emission factors for \(\text{CO}_2\) and CO are calculated via the
summation method, using background mole fractions for these gases measured before ignition, or upwind of the fires. Emission factors for the other gases are calculated using their respective emission ratios to either CO$_2$ or CO as also described in Sect. 4 of Paton-Walsh et al. (2014). Given that the results presented here cover measurements at 21 different fires from distinct tropical savanna ecosystem vegetation assemblages, and measured at different stages of the dry season (Table 1), further analysis will include an assessment of the within-ecosystem variability in emission factors.

3 Site and fire descriptions

Savanna fire emissions were measured at experimental fires in July and September 2009 near Kulnguki on the Arnhem Land Plateau, Northern Territory. Further measurements of experimental fires, land management fires and wildfires were carried out during August to October 2010 in the cities of Darwin and Palmerston, Litchfield Municipality, and Coomalie Shire of the Northern Territory. The flora of these areas is typical of eucalyptus-dominated savannas that comprise much of northern Australia, from the Kimberley region of Western Australia, through the Top End of the Northern Territory to Cape York in Queensland (Fig. 3).

3.1 Arnhem Land

Measurements in Arnhem Land were conducted during two field campaigns described at length by Meyer et al. (2012). Six experimental fires were measured during the early-dry season (EDS) between 2–7 July 2009, and a further five experimental fires were measured during the late-dry season (LDS) between 30 September and 4 October 2009.

The climate of West Arnhem Land is tropical monsoonal, and 97% of the mean annual rainfall falls between November and April. The closest weather station, at Oenpelli (12°20' S 133°03' E), 100 km west-northwest of the experimental fire sites records a mean annual rainfall of 1417 mm, with a mean maximum temperature of 30°C in July and 38°C in October.

The experimental fire sites were located near Kulnguki, in West Arnhem Shire, Northern Territory (12°38' S 133°55' E, Fig. 4), 330 km east of Darwin. One set of fires were situated on the Queue land system (Lynch and Wilson, 1998; Meyer et al., 2012). Here, the dominant trees are similar to that of the TGOW, whereas the understorey is mainly dominated by hummock grass (Triodia) with a variety of shrubs (Fig. 5b). The vegetation of this
system will be referred to as “hummock grass open woodland” (HGOW). While tussock grasses can be found in savanna systems worldwide, the hummocky spinifex grasses are endemic to Australia (Meyer et al., 2012), adapted to the arid climate with a shrub-like growth form. In addition to the TGOW and HGOW vegetation classes, fires were measured in open woodland areas with very few trees, dominated by sorghum grasses (Fig. 5c); this vegetation class will be referred to as “sorghum open woodland” (SOW). Individual hummocks of spinifex grasses were also burned and measured for their emissions (Fig. 5d); this vegetation class will be referred to as “pure spinifex” (SPIN). An extended description of the particular species at these fire sites can be found in Meyer et al. (2012).

The tussock-grass open woodland and sorghum open woodland fires were conducted in a series of adjacent plots established prior to the first campaign along an access road. Each tussock-grass open woodland plot was approximately 500 m × 200 m in size, whilst each sorghum open woodland plot was approximately 100 m × 100 m in size. The dissected quartz sandstone landscape of the hummock-grass open woodland allowed for the identification of two isolated plots prior to the campaign, each with areas of 50–70 ha, surrounded by natural fire barriers. The plots were situated in isolated country accessible only by helicopter, approximately 6–10 km from the tussock-grass plots. These experimental plots were ignited by a line ignition pattern on the upwind side of the plot, with some plots having a back fire also ignited on the downwind side of the plot. The isolated hummocks of spinifex could be ignited individually without risk of establishing a fire front due to the surrounding rocky outcrop. For further descriptions of the Arnhem Land experimental fires, including fuel loads and fuel moisture content, refer to Meyer et al. (2012).

3.2 Darwin region

A third measurement campaign to the Northern Territory was conducted during August to October in the dry season of 2010. Measurements were made at a number of wildfires, as well as experimental fires, at various locations within a region 120 km to the south/southeast of Darwin (Fig. 4). The climate of this area is tropical monsoonal, with 95% of mean annual rainfall falling between April and November. According to the weather station at Batchelor (within 60 km of all fire measurement sites), mean annual rainfall is 1528 mm, with a mean maximum temperature of 32°C in July and 37°C in October.

The opportunistic nature of this campaign allowed for emissions to be measured from fires in a number of different savanna land systems within the vicinity of Darwin. These included the gently undulating plains of the Koolpinyah land system (Christian et al., 1953), dominated by open eucalyptus woodland with tussock grasses (Fig. 5e and f). This vegetation class is similar to the tussock grass open woodland of the Arnhem Land campaigns and will also be indicated using TGOW. Measurements were also made at fires on the flat perennially flooded plains of the Marrakai land system (Christan et al., 1953), mainly consisting of tussock grasslands (Fig. 5g). This vegetation class will be referred to as “tussock grass floodplain” (TGFP). Fires were also measured on the flat country of the Batchelor land system (Christian et al., 1953), dominated by tall closed woodland with tussock and gamba (Andropogon gayanus) grasses (Fig. 5h). This vegetation class will be referred to as “tussock and gamba grass woodland” (TGGW).

Unlike the purely experimental fires of the Arnhem Land campaign, most of the fires measured for the Darwin region campaign were either wildfires or burn-off, where there was no control over the timing and pattern of ignition, and with many fires burning across many hectares. Locations of burn-off and wildfires were identified by regular monitoring of the NAFI website (http://www.firenorth.org.au), which shows the locations of satellite-detected fires; and by regular
phone calls to Bushfires NT (Northern Territory Government), who are responsible for supporting landholders with fire mitigation.

Wildfires were measured at the Adelaide River station on Marrakai Road (TGFP) on 1 October 2010; near Lambells Lagoon off Arnhem Highway (TGOW) on 1 October 2010; and at Berry Springs (TGOW) on 4 October 2010. Burn-offs were measured at the Gould airstrip near Batchelor (TGGW) on 11 September 2010 and at Southport (TGOW) on 6 October 2010. In addition to the wildfires and burn-offs, measurements were made at a set of three experimental fires in the Territory Wildlife Park (TGOW) near Berry Springs on 5 October 2010. Each of the experimental plots was approximately 130 × 80 m in size, and the fires and fire conditions from this campaign are presented in Table 1.

4 Results

A time series of smoke-tracer mole fractions retrieved from FTIR spectra collected near to the downwind boundary of one of the experimental fires is shown in Fig. 6, with CO₂, CO and MCE shown in detail in Fig. 7. In most of the experimental burns (with the exception of the hummock grass plots), the fires were ignited with a back-burning fire line on the down-wind boundary of the fire plot; this fire moves slowly against the wind, forming a fire break. The up-wind boundary is then ignited, with the majority of the plot burning with a head fire driven by the prevailing wind. Generally, the smoke passing through the infrared beam will have first originated from the flaming combustion in the backburn fire line (notice the high MCE in the first minute of the time series in Fig. 7). Immediately following the initial burst of smoke from the flaming combustion, the smoke from the flaming back-burn will be mixing with smouldering smoke, also originating from the back-burn, with increasing CO mole fraction and a corresponding decrease in MCE (until ∼7 mins in Fig. 7). As the back-burn fire line burns away from the OP-FTIR path, most of the smoke crossing the infrared beam will have been emitted from the steadily increasing area of smouldering-phase combustion, with contributions from both the flaming combustion of the head fire and the smouldering area behind the head fire (towards the end of the time series in Fig. 7, as MCE decreases to ∼90% and below). This equates to the area of “flaming combustion” (i.e. the fire line) and the associated “smoking zone” of smouldering combustion located close behind the fire front, as detailed in the OP-FTIR African savannah measurements of Wooster et al. (2011).

4.1 Emission ratios

The emission ratios of each trace gas relative to CO₂ and CO are calculated using the generalised least squares linear regression slope (this is the best fit to the points that minimises
deviations from the line of fit in both the x and y axes, weighted by uncertainties in both x and y of the trace gas species mole fraction versus the CO$_2$ or CO mole fraction. As an example, Figs. 8 and 9 show a series of trace gas mole fractions plotted against CO$_2$ and CO, respectively. The generalised least squares linear best fits are shown in each case, with the gradient taken as the emission ratio [ER$_X$/CO$_2$ or ER$_X$/CO]. As is evident in both Figs. 8 and 9, it is clear that for all gases there is a stronger correlation to CO than to CO$_2$ (due to their preferential production during smouldering-phase combustion, as is also the case for CO). For most gases (CH$_4$, C$_2$H$_2$, C$_2$H$_4$, C$_2$H$_6$, CH$_3$OH, CH$_3$COOH, HCOOH, NH$_3$, HCN), their emission ratio to CO is used for the calculation of their respective emission factors. For H$_2$CO, the only gas to show a better mean correlation with CO$_2$ (across all 21 fires), the emission ratio to CO$_2$ is used for the calculation of its emission factor (Paton-Walsh et al. (2014) also found C$_2$H$_4$ to be better correlated to CO$_2$). Following Sinha et al. (2003) and Wooster et al. (2011), the strength of the coefficient of determination ($R^2$) between the mole fractions of each target gas and that of CO or CO$_2$ is used to confirm whether the emission ratio for the target gas is well determined. For gases with stronger IR absorption features (e.g. CO, CH$_4$, C$_2$H$_4$, NH$_3$), $R^2$ in the figures was generally higher than 0.8, whilst gases with weaker IR absorption (e.g. CH$_3$OH, HCN, HCOOH) showed poorer correlations to CO$_2$ and CO due to greater uncertainties in the retrieval process. Those species for which the coefficient of determination fell below 0.4 are assumed to have poorly determined emission ratios (Sinha et al., 2003; Wooster et al., 2011). Consequently, emission ratios where $R^2$ fell below 0.4 were not used in the calculation of emission factors.

Given that MCE is known to decrease as the fire progresses through the plot (Fig. 7), with a greater contribution from the increasing area of smouldering combustion behind the fire front, it is assumed that measurements with the highest MCE are of smoke that has predominantly originated from flaming combustion, whilst those measurements with lower MCE are of smoke that has a higher proportion of smouldering combustion (Yokelson et al., 1996). Whilst the within-fire variation in emission ratios is quite limited, as shown by the high value of $R^2$ on the linear least squares regression fits in the bivariate plots (Figs. 8 and 9), this ever-changing ratio of flaming-to-smouldering smoke does lead to some variation in the emission ratio of most of the gases to CO$_2$ (as evident in the deviation from the linear best-fit in most of the scatter plots shown in Fig. 8). To account for this, it is possible to calculate different emission ratios to CO$_2$ for the different combustion stages of each fire. Conventionally, laboratory studies use observations of fire activity to determine the combustion stage of emissions (e.g. Yokelson et al., 1996) or an approach where the maximum rate of increase of CO emission is taken as the beginning of the smouldering stage (e.g. Lobert et al., 1991). The OP-FTIR study of Wooster et al. (2011) had access to simultaneous airborne imaging, so was able to separate out times of different fire activity based on those thermal/optical airborne observations. Here, measured emissions cannot be ascribed to particular combustion stages using observations, and furthermore the relative contribution to measured emissions from flaming and smouldering combustion may vary depending on changes in the wind direction, ignition pattern of the fire, and heterogeneous vegetation cover and fuel moisture within the fire plot. Instead, the emissions measurements are split by MCE, so that measurements with high MCE (> 90 %) and those with low MCE (< 90 %) may be used separately to calculate emission ratios (Fig. 10). This approach has been used in the past for laboratory analysis, and the 90 % threshold was taken from studies that have published MCE for flaming and smouldering combustion stages (e.g. Lobert et al., 1991; Yokelson et al., 1996; Bertschi et al., 2003b). Emission ratios of each target species to both CO$_2$ and CO were assessed in this way and are presented in Tables 2 and 3, respectively, in addition to emission ratios derived from all measurements.

As discussed above, the within-fire variation in emission ratios is fairly limited, and can be understood by splitting measurements by MCE as already described. The between-fire variation, however, is larger, with the flaming emission ratio of CH$_4$, for example, varying from 0.0010 (for spinifex) to 0.0045 for one of the tussock grass open woodland plots. Most of this variation is caused by comparison between the “landscape” experimental fires (those that incorporate a number of fuel species; e.g. TGOW, HGOW, SOW) and
Figure 8. Example scatter plots of the measured trace gas path-length-averaged mole fractions used to calculate emission ratios. The data presented here are from LD-TGOW-A (30 September 2009). The slope of the least squares linear best fit to these data (thick red line) is used to derive the relevant emission ratio ($ER_{x/CO_2}$). For each of the 10 trace gases shown above, the emission ratio and the $R^2$ is given towards the top of each plot. The equivalent plots for emission ratios to CO are presented in Fig. 9.

the experimental fires that were conducted on the individual clumps of spinifex. Given that the spinifex measured here appeared to burn with high combustion efficiency (possibly due to measured high burn temperature, aerated structure, and leaf resin content), producing little observable "smoke" (see Fig. 5d), it is perhaps no surprise that its emissions characteristics are different from those from experimental fires containing multiple species, where vegetation that burn at different combustion efficiencies contribute to a plume-averaged efficiency (e.g. the tussock grass and hummock grass open
woodland landscapes consist of both fine and coarse fuels, each with different structure and degrees of compaction. As the purpose of this work is to characterise landscape-averaged emissions, the emission ratios and emission factors from the spinifex fires have been excluded from calculating average emissions parameters.

The emission ratios presented in Table 2 (ER_{X/CO_2}) and Table 3 (ER_{X/CO}) show the mean emission ratios for the landscape fire measurements (including HGOW, SOW, TGFP, TGGW, TGOW vegetation classes), excluding those derived from individual species fires (SPIN). Table 2 confirms that the emission ratios to CO_2 (i.e. ER_{X/CO_2}), for all of the gases investigated here, are higher in the measurements we class as “smouldering” (based on MCE < 90 %) than in those measurements classed as “flaming” (MCE > 90 %). The close correlation between the emissions of gases and those of CO (Fig. 9) also explains why there is no significant difference in the emission ratios to CO (ER_{X/CO}) recorded during these “flaming” and “smouldering” classes (the difference between “flaming” and “smouldering” ER_{X/CO} is smaller than the standard deviation of each ER). For formaldehyde (H_2CO), the higher emission ratio to CO during the “flaming” class might be due to the production of formaldehyde during the closely related (in space and time) pyrolysis combustion phase (Lobert and Warnatz, 1993); with some of the pyrolysed gas escaping combustion due to advective or convective winds. Wooster et al. (2011) suggested pyrolysis was the cause of the similarly higher OP-FTIR derived emission ratios to CO for formaldehyde in the flaming rather than the residual smoldering combustion stage of southern African savannah fires.

4.2 Drivers of variations in emission ratios and emission factors

Figures 8 and 9 depict trace gas retrievals and emission ratios derived from data of a typical experimental fire. Given
that a number of different vegetation types were sampled during the three Australian campaigns, with measurements made throughout the early and late dry season, and the likelihood that each experimental fire would have burned at different efficiencies depending on the fuel and meteorological conditions, it is important to assess the relative importance of each of these variables. Figure 11 includes all data from the 2009 campaigns, showing the scatter plot for CH$_4$ vs. CO$_2$ only, with each subplot (a–c) depicting the same data but with each data point (i.e. each individual retrieval from four co-added spectra collected over ~ 5 s) colour-coded depending on its MCE (Fig. 11a); the season of the experimental fire (Fig. 11b); and the predominant vegetation type for the burn (Fig. 11c). As discussed above, the between-fire variation in emission ratios can be primarily explained by a dependence on vegetation class (as is evident in Fig. 11c), with the spinifex (individual species) fires grouping towards the lower emission ratios and the sorghum dominated fires grouping towards the higher emission ratios. Figure 11a clearly demonstrates that this grouping is mainly a result of differences in combustion efficiencies of these different vegetation classes, with MCE appearing to be the major control on the emission ratio slope, despite differences in vegetation classes. There is no clear pattern to discern when the data are plotted by season (Fig. 11b).

By sub-sampling the data in Fig. 11a into 2 % bins of MCE (e.g. only measurements where 80 % < MCE ≤ 82 %), it is possible to calculate a range of emission ratios to CO$_2$ depending on MCE. Figure 12 depicts 11 different CH$_4$ emission ratios to CO$_2$ (ER$_{CH_4/CO_2}$) calculated in this way. This figure demonstrates the strong relationship between the emission ratio of methane and the modified combustion efficiency, which is unsurprising due to the well-defined CO to CO$_2$ and CH$_4$ to CO emission ratios already shown in Fig. 9. A clear relationship between the slope of the linear least squares regression lines and MCE is evident as the slope increases systematically with decreasing MCE. This relationship applies for the other trace gases that also correlate well with carbon monoxide. These MCE-dependent emission ratios can then be used to calculate MCE-dependent emission factors, discussed further below.

### 4.3 Final emission factors

Emission factors for CO$_2$ and CO for each fire were calculated by the summation method as described in Paton-Walsh et al. (2014). Emission factors for all other gases were calculated from the CO$_2$ and CO emission factors for the relevant fires and the emission ratio to CO$_2$ or CO (given in Tables 2 and 3). These emission factors are presented in Table 4. Mean emission factors calculated from the individual fire emission factors in Table 4 are listed in Table 5.

By combining measurements from all fires, it is possible to build a detailed understanding of the relationship between MCE and emission ratios (as depicted earlier in Figs. 11a...
Table 4. Emission factors (g kg\(^{-1}\) of dry fuel burned) for each individual fire, calculated using the standardised method outlined in Paton-Walsh et al. (2014). Uncertainties were calculated in quadrature from those associated with the trace gas emission ratios and a ±10 % uncertainty in the assumed fuel carbon.

<table>
<thead>
<tr>
<th>CO(_2) EF</th>
<th>CO EF</th>
<th>CH(_4) EF</th>
<th>C(_2)H(_5) EF</th>
<th>C(_2)H(_4) EF</th>
<th>CH(_3)OH EF</th>
<th>CH(_3)COOH EF</th>
<th>HCOOH EF</th>
<th>HCN EF</th>
<th>NH(_3) EF</th>
</tr>
</thead>
<tbody>
<tr>
<td>LD-TGOW-A</td>
<td>1638 ± 161</td>
<td>110 ± 18</td>
<td>2.56 ± 0.41</td>
<td>0.16 ± 0.03</td>
<td>0.82 ± 0.08</td>
<td>nr ± nr</td>
<td>1.76 ± 0.18</td>
<td>1.15 ± 0.18</td>
<td>1.97 ± 0.32</td>
</tr>
<tr>
<td>LD-TGOW-B</td>
<td>1603 ± 160</td>
<td>134 ± 21</td>
<td>3.22 ± 0.51</td>
<td>0.11 ± 0.02</td>
<td>0.64 ± 0.05</td>
<td>nr ± nr</td>
<td>2.08 ± 0.21</td>
<td>1.20 ± 0.19</td>
<td>1.96 ± 0.31</td>
</tr>
<tr>
<td>LD-SOW-A</td>
<td>1528 ± 153</td>
<td>169 ± 27</td>
<td>5.94 ± 0.95</td>
<td>0.18 ± 0.03</td>
<td>0.75 ± 0.08</td>
<td>nr ± nr</td>
<td>2.43 ± 0.24</td>
<td>1.76 ± 0.20</td>
<td>3.25 ± 0.52</td>
</tr>
<tr>
<td>LD-SPIN</td>
<td>1716 ± 172</td>
<td>75 ± 12</td>
<td>1.28 ± 0.20</td>
<td>0.11 ± 0.01</td>
<td>0.53 ± 0.05</td>
<td>nr ± nr</td>
<td>0.53 ± 0.05</td>
<td>0.67 ± 0.11</td>
<td>1.57 ± 0.25</td>
</tr>
<tr>
<td>LD-TGOW</td>
<td>1662 ± 166</td>
<td>100 ± 16</td>
<td>2.01 ± 0.32</td>
<td>0.09 ± 0.01</td>
<td>0.24 ± 0.02</td>
<td>nr ± nr</td>
<td>0.16 ± 0.02</td>
<td>0.04 ± 0.01</td>
<td>1.44 ± 0.14</td>
</tr>
<tr>
<td>ED-TGOW-A</td>
<td>1669 ± 167</td>
<td>96 ± 15</td>
<td>1.92 ± 0.31</td>
<td>nr ± nr</td>
<td>0.16 ± 0.02</td>
<td>nr ± nr</td>
<td>0.81 ± 0.13</td>
<td>1.64 ± 0.26</td>
<td>nr ± nr</td>
</tr>
<tr>
<td>ED-TGOW-B</td>
<td>1679 ± 168</td>
<td>87 ± 14</td>
<td>1.81 ± 0.29</td>
<td>nr ± nr</td>
<td>0.74 ± 0.07</td>
<td>nr ± nr</td>
<td>1.54 ± 0.19</td>
<td>0.99 ± 0.16</td>
<td>1.92 ± 0.31</td>
</tr>
<tr>
<td>ED-SOW-A</td>
<td>1670 ± 167</td>
<td>92 ± 15</td>
<td>2.16 ± 0.35</td>
<td>nr ± nr</td>
<td>0.52 ± 0.05</td>
<td>nr ± nr</td>
<td>1.39 ± 0.14</td>
<td>1.48 ± 0.24</td>
<td>2.64 ± 0.42</td>
</tr>
<tr>
<td>ED-SOW-B</td>
<td>1624 ± 162</td>
<td>118 ± 19</td>
<td>2.87 ± 0.46</td>
<td>0.07 ± 0.01</td>
<td>0.56 ± 0.08</td>
<td>nr ± nr</td>
<td>nr ± nr</td>
<td>nr ± nr</td>
<td>nr ± nr</td>
</tr>
<tr>
<td>ED-SPIN</td>
<td>1763 ± 176</td>
<td>39 ± 6</td>
<td>0.71 ± 0.11</td>
<td>nr ± nr</td>
<td>nr ± nr</td>
<td>nr ± nr</td>
<td>1.02 ± 0.16</td>
<td>0.96 ± 0.15</td>
<td>0.38 ± 0.06</td>
</tr>
<tr>
<td>BAT-TGOW</td>
<td>1713 ± 171</td>
<td>60 ± 10</td>
<td>1.65 ± 0.26</td>
<td>nr ± nr</td>
<td>nr ± nr</td>
<td>nr ± nr</td>
<td>nr ± nr</td>
<td>nr ± nr</td>
<td>nr ± nr</td>
</tr>
<tr>
<td>ARS-TGFP-A</td>
<td>1746 ± 175</td>
<td>45 ± 7</td>
<td>1.11 ± 0.18</td>
<td>nr ± nr</td>
<td>nr ± nr</td>
<td>nr ± nr</td>
<td>nr ± nr</td>
<td>nr ± nr</td>
<td>nr ± nr</td>
</tr>
<tr>
<td>ARS-TGFP-B</td>
<td>1676 ± 168</td>
<td>86 ± 14</td>
<td>1.61 ± 0.26</td>
<td>0.13 ± 0.02</td>
<td>nr ± nr</td>
<td>nr ± nr</td>
<td>nr ± nr</td>
<td>nr ± nr</td>
<td>nr ± nr</td>
</tr>
<tr>
<td>ARS-TGFP-C</td>
<td>1695 ± 169</td>
<td>76 ± 12</td>
<td>1.45 ± 0.23</td>
<td>nr ± nr</td>
<td>nr ± nr</td>
<td>nr ± nr</td>
<td>nr ± nr</td>
<td>nr ± nr</td>
<td>nr ± nr</td>
</tr>
<tr>
<td>ARS-TGFP-D</td>
<td>1746 ± 175</td>
<td>46 ± 7</td>
<td>0.84 ± 0.13</td>
<td>nr ± nr</td>
<td>nr ± nr</td>
<td>nr ± nr</td>
<td>nr ± nr</td>
<td>nr ± nr</td>
<td>nr ± nr</td>
</tr>
<tr>
<td>ARS-TGFP-E</td>
<td>1746 ± 175</td>
<td>46 ± 7</td>
<td>0.84 ± 0.13</td>
<td>nr ± nr</td>
<td>nr ± nr</td>
<td>nr ± nr</td>
<td>nr ± nr</td>
<td>nr ± nr</td>
<td>nr ± nr</td>
</tr>
<tr>
<td>BSP-TGOW</td>
<td>1695 ± 170</td>
<td>76 ± 12</td>
<td>1.79 ± 0.29</td>
<td>nr ± nr</td>
<td>nr ± nr</td>
<td>nr ± nr</td>
<td>nr ± nr</td>
<td>nr ± nr</td>
<td>nr ± nr</td>
</tr>
<tr>
<td>TWP-TGOW-A</td>
<td>1740 ± 174</td>
<td>49 ± 8</td>
<td>1.15 ± 0.18</td>
<td>nr ± nr</td>
<td>nr ± nr</td>
<td>nr ± nr</td>
<td>nr ± nr</td>
<td>nr ± nr</td>
<td>nr ± nr</td>
</tr>
<tr>
<td>TWP-TGOW-B</td>
<td>1655 ± 165</td>
<td>96 ± 15</td>
<td>2.60 ± 0.42</td>
<td>nr ± nr</td>
<td>nr ± nr</td>
<td>nr ± nr</td>
<td>nr ± nr</td>
<td>nr ± nr</td>
<td>nr ± nr</td>
</tr>
<tr>
<td>TWP-TGOW-C</td>
<td>1627 ± 163</td>
<td>114 ± 18</td>
<td>3.06 ± 0.49</td>
<td>0.07 ± 0.01</td>
<td>0.46 ± 0.05</td>
<td>nr ± nr</td>
<td>2.32 ± 0.23</td>
<td>0.87 ± 0.14</td>
<td>1.54 ± 0.25</td>
</tr>
<tr>
<td>SPT-TGOW</td>
<td>1721 ± 172</td>
<td>59 ± 9</td>
<td>1.23 ± 0.20</td>
<td>nr ± nr</td>
<td>0.31 ± 0.03</td>
<td>nr ± nr</td>
<td>0.99 ± 0.10</td>
<td>1.21 ± 0.19</td>
<td>nr ± nr</td>
</tr>
</tbody>
</table>

Figure 10. An example scatter plot of the measured path-length-averaged mole fractions of CO\(_2\) and CH\(_4\) (as in Fig. 8, top centre). Here, the data have been split using MCE, with measurements having an MCE greater than 90 % (red) being considered separately to those with a lower MCE (black). Different emission ratios and emission factors for flaming and smouldering emissions can then be determined (also shown on the plot in red and black text respectively, the green text indicates the emission ratio and factors determined using all data).

and 12). MCE-dependent emission ratios can be used to calculate MCE-dependent emission factors. Using the same methodology applied to calculate emission factors for individual fires, it is possible to calculate emission factors for each 2%-MCE bin. Figure 13 shows the relationship between MCE and the emission factor of CH\(_4\) (EF CH\(_4\)) calculated using the measurements from different vegetation classes and when using all measurements combined. This plot demonstrates that the majority of variation in the emission factor for CH\(_4\) is determined by MCE irrespective of vegetation type. The uncertainty bounds of the CH\(_4\) emission factor for any particular vegetation class lie within the bounds of the emission factor determined from all measurements for each MCE bin (except TGFP in the 86–88 % MCE bin).

5 Comparison to past EF and EF measures

The emission factors presented here are the first to be reported from OP-FTIR measurements of tropical savannas in Australia and follow only a few ground-based (Hurst et al., 1994a) and airborne studies (Hurst et al., 1994b; Shirai et al., 2003) in this ecosystem. The emission factors for CH\(_3\)OH, CH\(_3\)COOH, and HCOOH are the first to be reported by any methodology for Australian savanna fires (Table 5). The emission factors presented here have been estimated from measurements at up to 21 separate fires from a variety of vegetation types and from both controlled experimental burns and wildfires (Table 4). Table 5 compares the emission factors reported here with those reported by previous studies of tropical savannas in Australia (listed above) and with those reported by the emission factor databases for global savanna and grasslands in Andreae and Merlet (2001) and global savannas in Akagi et al. (2011).
Table 5. Emission factors (g kg\(^{-1}\) dry fuel burned) reported by this study and those for the same trace gases reported by previous Australian savanna studies (Hurst et al., 1994a, b; Shirai et al., 2003) and by Andreae and Merlet (2001) and Akagi et al. (2011) who estimate global savanna emission factors by amalgamating suite of emissions studies. For this study and the Australian savanna studies, the mean and standard deviation of the emission factor is calculated from individual fires.

<table>
<thead>
<tr>
<th></th>
<th>This study</th>
<th>Australian tropical savanna</th>
<th>All savanna</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>EF SD</td>
<td>Hu1994a EF SD</td>
<td>Hu1994b EF SD</td>
</tr>
<tr>
<td>CO</td>
<td>1674 ± 56</td>
<td>1595 ± 121</td>
<td>1646 ± 106</td>
</tr>
<tr>
<td>CO</td>
<td>87 ± 33</td>
<td>91.4 ± 27</td>
<td>58.9 ± 20</td>
</tr>
<tr>
<td>CH(_4)</td>
<td>2.10 ± 1.16</td>
<td>2.3 ± 1.29</td>
<td>2.3 ± 0.79</td>
</tr>
<tr>
<td>C(_2)H(_2)</td>
<td>0.11 ± 0.04</td>
<td>0.12 ± 0.08</td>
<td>0.10 ± 0.05</td>
</tr>
<tr>
<td>C(_2)H(_4)</td>
<td>0.49 ± 0.22</td>
<td>nr ± nr</td>
<td>0.53 ± 0.24</td>
</tr>
<tr>
<td>C(_2)H(_6)</td>
<td>0.08 ± 0.05</td>
<td>0.05 ± 0.03</td>
<td>0.07 ± 0.02</td>
</tr>
<tr>
<td>CH(_3)O</td>
<td>1.57 ± 0.44</td>
<td>0.26 ± 0.16</td>
<td>nr ± nr</td>
</tr>
<tr>
<td>CH(_3)OH</td>
<td>1.06 ± 0.87</td>
<td>nr ± nr</td>
<td>nr ± nr</td>
</tr>
<tr>
<td>CH(_3)COOH</td>
<td>1.54 ± 0.64</td>
<td>nr ± nr</td>
<td>nr ± nr</td>
</tr>
<tr>
<td>HCOOH</td>
<td>0.16 ± 0.07</td>
<td>nr ± nr</td>
<td>nr ± nr</td>
</tr>
<tr>
<td>HCN</td>
<td>0.53 ± 0.31</td>
<td>0.03 ± 0.02</td>
<td>0.03 ± 0.02</td>
</tr>
<tr>
<td>NH(_3)</td>
<td>0.70 ± 0.36</td>
<td>1.53 ± 0.86</td>
<td>0.61 ± 0.39</td>
</tr>
</tbody>
</table>

\(^{a}\) The emission factor reported by Andreae and Merlet (2001) is the mean emission factor of all of the contributing studies, and the standard deviation is that of the studies (unless only two studies were used for a particular emission factor, in which case, the uncertainty is represented by a range).

\(^{b}\) The emission factor reported by Akagi et al. (2011) is the mean of the contributing studies, whilst the uncertainty is an indication of the natural variability of the emission factor estimated from the reported variability in the contributing studies.

5.1 Carbonaceous species

A broad range of emission factors for carbon dioxide (EF CO\(_2\)) are reported in the literature from savanna ecosystems (Table 5). The emission factors reported here are higher than those reported by previous studies of northern Australian savanna fires, but fall within the standard deviation of measurements in northern Australia made by Hurst et al. (1994b) and Shirai et al. (2003). The EF CO\(_2\) reported here agrees well with those reported in Andreae and Merlet (2001) and subsequent updates in 2009), and Akagi et al. (2011), which draw on numerous studies in savanna ecosystems worldwide (the mean EF CO\(_2\) reported here is only 0.7 % smaller than that reported by Akagi et al. (2011) for savanna ecosystems).

The carbon monoxide emission factor (EF CO) reported here agrees well with previous studies of biomass burning in Australian tropical savannas, which appear to be slightly higher than the global average for savannas (Akagi et al., 2011) report an EF CO of 63 g kg\(^{-1}\), whilst the mean EF CO for the Australian studies, including this one, is 81 g kg\(^{-1}\)). This might indicate that the efficiency of burning in Australian tropical savannas might be lower than the global savanna average; indeed, the global ecosystem averages include emissions from grasslands, which are believed to burn with higher MCE due to the relatively higher abundance of finer fuels than is found in woodland savannas (van Leeuwen and van der Werf et al., 2011).

For the methane emission factor (EF CH\(_4\)), the EF reported here agrees with all of the previous Australian studies and the global savanna EF CH\(_4\). The mean emission factors of C\(_2\)H\(_2\) and C\(_2\)H\(_4\) are both lower than those reported by Andreae and Merlet (2001) and Akagi et al. (2011), and are closer to those measured in the northern Australian studies (the mean EF C\(_2\)H\(_2\) and EF C\(_2\)H\(_4\) for the northern Australian studies, including this one, at 0.11 g kg\(^{-1}\) and 0.44 g kg\(^{-1}\) respectively, are both about half of those reported for global savannas).

The emission factor for formaldehyde (EF H\(_2\)CO) is six-times higher than the only previous study to report EF H\(_2\)CO for Australian savanna fires (Hurst et al. 1994a), but is in closer agreement to the estimates for global savannas (Andreae and Merlet, 2001; Akagi et al., 2011). Paton-Walsh et al. (2010) (who report emission ratios to CO, but not emission factors, for northern Australian fires from solar occultation FTIR measurements) also find that their emission ratio for H\(_2\)CO is far higher than that reported in Hurst et al. (1994a) and suggest that one reason for this discrepancy might be that the H\(_2\)CO measured by Hurst et al. (1994a), whose study was based on cryogenically stored samples, may have been subject to losses during sampling or storage. Both the OP-FTIR methodology used here and the methodology of Paton-Walsh et al. (2010) use long open-path measurements and so do not suffer from such problems associated with losses to the walls of sampling or storage equipment. Wooster et al. (2011), who also use a long open-path FTIR methodology, also find their H\(_2\)CO emission factor for fires measured in Kruger National Park (South Africa) to
Figure 11. Scatter plots of the measured mole fractions of CO$_2$ and CH$_4$, including all measurements made during the two Arnhem Land campaigns in 2009. All three plots show the same data, but are colour-coded differently using (a) MCE; (b) season of measurement; and (c) vegetation class.

Figure 12. Same as Fig. 11a, but with all data from the 2009 and 2010 campaigns, including generalised least squares linear best-fit lines that have been calculated using measurements sub-sampled by narrow (2%) bins of MCE. Each line represents the line of best fit to the sub-sampled emission measurements (the MCE bin for each line is indicated on the plot). Similar plots for other gas species can be found in Supplement Figs. S1 and S2.

Figure 13. The emission factor for CH$_4$ (EF CH$_4$) grouped by MCE, calculated using the emission ratios determined from 2%-wide MCE bins for different vegetation classes, and for all measurements combined (error bars indicate the estimated uncertainty calculated in accordance with Paton-Walsh et al., 2014).

be significantly elevated compared with literature values for African savannah fires.

The emission factors for CH$_3$OH, CH$_3$COOH and HCOOH are the first to be reported for Australian savannas. Globally, only two savanna studies report EFs for these gases in the database of Andreae and Merlet (2001, with 2009 updates), whilst the standard deviation for the EFs of these gases is high in Akagi et al. (2011) (ranging from 35–48% of the mean EF). EF HCOOH and EF CH$_3$OH reported here lies closest to the latest estimates of Akagi et al. (2011) for
5.2 Nitrogenous species

There is a very clear discrepancy between the emission factors for HCN (EF HCN) reported here and those from Hurst et al. (1994a) and Hurst et al. (1994b). EF HCN reported here is over an order of magnitude higher than those from the previous Australian savanna studies. As was the case for H2CO, Paton-Walsh et al. (2010) also find their measured emission ratio of HCN to CO to be over 6 times higher and 3 times higher than that of Hurst et al. (1994a) and Hurst et al. (1994b) respectively. It is suggested that one reason for these discrepancies might be that the HCN measured by Hurst et al. (1994a) and Hurst et al. (1994b) may also have been subject to losses during sample handling and storage. The EF HCN reported here agrees with measurements from other savanna ecosystems (Table 5).

The emission factor for ammonia (EF NH3) lies closest to that reported by Hurst et al. (1994b) for Australian savannas (Table 5), and shows no significant difference from that reported for global savannas in Andreae and Merlet (2001) and Akagi et al. (2011).

6 Summary and conclusions

Griffith et al. (1991) first demonstrated the use of long open-path FTIR for vegetation fire emission ratio measurements, and this approach was then revisited by Wooster et al. (2011) where a methodology for deriving biomass burning trace gas emission factors from these long open-path FTIR measurements was developed and applied to fires in the savannas of Kruger. That work focussed on five key gases retrieved in different parts of the infrared spectral region (900–3100 cm⁻¹). Most recently, Akagi et al. (2014) also deploy a long open-path FTIR methodology at fires in the United States, greatly expanding the number of gases retrieved using this approach. The current work and the companion paper by Paton-Walsh et al. (2014) have retrieved emission factors for 15 different trace gases, has explored different ways of combining timeseries of emission measurements’ derived “whole-fire” emission factors, and has explored relationships between these emission factors and the measurements’ modified combustion efficiencies, seasonality and vegetation class. This work has applied these methods to a large number [21] of fires conducted in Australian savanna, including both prescribed/experimental burns and wildfires, and is one of only a few limited studies of biomass burning emissions in the Australian savannas, which is surprising given that Australian savanna burning is an important source of a number of climatically important atmospheric trace gases at both a national and global level.

Generally, the averaged emission factors presented here are in line with global savanna emission factors (e.g. Akagi et al., 2011; Table 5), with the exception of formaldehyde (H2CO) that has an emission factor of around double that reported by Andreae and Merlet (2001) and Akagi et al. (2011). This is in line with other studies that deploy open-path methodologies (Paton-Walsh et al., 2010; Wooster et al., 2011). The emission factors for H2CO and HCN are substantially larger than those measured in previous studies of Australian savannas (Hurst et al., 1994a, b). As suggested by Paton-Walsh et al. (2010), who also find discrepancies between their measurements of H2CO and HCN in northern Australia and those of Hurst et al. (1994a, b). One explanation for this underestimation by these earlier studies might be due to losses of H2CO and HCN during sample handling and storage. For CH3OH, CH3COOH, and HCOOH, the emission factors reported here represent one of only a few studies to report EFs of these gases in savanna burning globally and the first for Australian tropical savannas.

The reported variability in emission factors for individual fires (Table 4) and the standard deviation of the mean emission factor calculated from all of the measured fires is large, although it is in line with the variability reported by Akagi et al. (2011). Understanding the causes of this variability is one of the important scientific questions posed by those who wish to better model global biomass burning emissions (van Leeuwen and van der Werf, 2011). Here, and in Meyer et al. (2012), we find that the emission ratios and emission factors varied substantially for different vegetation classes (Fig. 11 and Table 4), but that the MCE of the burns can be used to explain the majority of this variation (Figs. 12 and 13). Data on MCE may soon be available from spaceborne total column measurements as Ross et al. (2013) demonstrate by reporting the first CH4 to CO2 emission ratios measured from space. Given that a significant majority of the variation in the emission factor for trace gases can be explained by MCE, irrespective of vegetation class, we conclude that the selection of emission factors for emissions modelling purposes need not necessarily require detailed fuel type information, if data on MCE (e.g. from such spaceborne measurements) or a correlated variable (e.g. CH4 emission ratios) were available.

The Supplement related to this article is available online at doi:10.5194/acp-14-11335-2014-supplement.
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References


