

Response to anonymous reviewer's #1 comments.

The authors thank the reviewer for their constructive comments. Our responses to anonymous reviewer's #1 comments are detailed below. Reviewer's comments are in italics and our responses in standard font.

General Comments:

Section 2.3 (Plume Dispersion Modelling) is not well described. 1) Was the same particle number used for the two fire types (boreal and temperate)? Is this reasonable? Are there references to support this approach? 2) What are the two different rates that are used for day and night emissions? Are there references to support different day/night rates? 3) When no fires were detected, why set the count to a minimum positive value instead of zero? Is this to account for undetected fires? What support is there for this approach? Please rewrite this section for clarity.

The authors apologise for the lack of clarity in the section. The section has been rewritten clarifying the steps undertaken to perform dispersion simulations, and addressing the specific questions raised in this comment. The answers (numbered in the reviewers comment) are listed below.

1) HYSPLIT particle number emitted per hour per fire detection was the same for both biomes. This was done for several reasons. Firstly, total particle number is not directly linked to particulate emission estimates. For example, if a grid cell has AOT value of 1, and 100 HYSPLIT particles are located within the cell during the satellite overpass, 80 of which were emitted two diurnal cycles ago, and 20 during the previous diurnal cycle, the grid cells AOT is split accordingly between the emission periods. Equivalently, if there are any particles emitted from different fire events, grid cells AOT is divided both between different emission periods and different fire events. The estimates for the two periods would not be different if there were more or less particles within the cell, its the relative ratios which matter. Secondly, the aim of the manuscript was to provide independent estimates therefore we tried to avoid using existing emission factors or emission coefficients (top-down emission estimates per unit of fire radiative energy). Notably, even if emission rates are indeed different for boreal and temperate events, the assumed identical particle numbers in the analysis would not have influenced the results significantly. This is due to the fact that mixing of the plumes from different biomes was minimal.

2) The emitted particle number per hour and fire detection were identical for daytime and night-time periods.

3) The count was set to minimum non-zero value to avoid total shutdown of emissions for a time period for which no MODIS fire detections were obtained, most likely due to cloud cover. This is an unlikely scenario for the long duration burning episodes presented in the manuscript.

The updated section:

Smoke transport for the selected fire events was simulated with the HYSPLIT model (Draxler et al., 2003). Plume dispersion from a source location was represented by the motion of a large number of discrete particles moved by the wind field with mean and random components. Global Data Assimilation System (GDAS) meteorological archive data was employed to drive the model.

For each day of burning, particles were continuously released into the model domain from the locations of the individual active fire detections within the fire event. In order to represent fire diurnal cycle, different MODIS active fire observations were used to release particles for two 12 hour intervals representing day and night emissions 09:00 to 21:00 and 21:00 to 09:00 local time respectively. Emission source number and locations for daytime periods were determined from the highest number of fire detections observed during a single either Terra or Aqua daytime

overpass with 10.30 and 13.30 equatorial crossing time. Similarly, emitted particle source numbers for the night periods were determined by the largest burning extent observed during one of the night-time overpasses with 22.30 and 1.30 equatorial crossing times. Notably, the Terra overpass at 22.30 in high latitudes makes observations of regions where local time is earlier than 21:00. In this study, however, all fires detected during this overpass were classed as night-time observations. If no valid observations were available for some of the time intervals, the count and fire pixel locations were set to a minimum non-zero value estimated for the burning episode from all daytime or night-time observations. This was done to avoid total temporary shut-down of the emissions, which is an unlikely scenario for a long duration burning episodes. Every hour, 20 particles were released for each fire pixel. As a result, emitted particle number for a burning episode was determined by the number of active fire pixels observed during a given time period.

Particles were uniformly distributed between the surface and the top altitude of the planetary boundary layer as given in GDAS archive. Satellite based plume height estimates (Val Martin et al., 2010, Peterson et al., 2014) indicate that in up to 80% of the events analysed, injection heights were limited to the planetary boundary layer. While confinement of the emissions to the mixing layer underestimates injection height for the most energetic burning episodes, such configuration should nonetheless represent the majority of burning episodes.

Throughout the simulations, modelled particle positions, their age and source burning event identifier were recorded each day at local solar noon. The generated point clouds were later used to compare against Terra and Aqua Aerosol Optical Thickness (AOT) observations. “

Water Content Retrieval. How do aerosol water fractions estimated in this work compare to aerosol water content that would be estimated using representative hygroscopic growth factors for representative relative humidities?

The inferred median water volume fractions would equate to geometric hygroscopic growth factors (gHGF) of 1.05 and 1.24 for boreal and temperate plumes respectively. Such factors suggest that boreal plumes belong to “nearly-hydrophobic” group (gHGF 1.0-1.11) while temperate plumes fall into “less-hygroscopic” category (gHGF range 1.11–1.33) as suggested in Swietlicky et al., 2008 review of measured growth factors. Note that the measured growth factors were recorded at 90% relative humidity, while water fractions inferred in our study were obtained at ambient relative humidities. In result, a direct comparison is not very meaningful, but our numbers do seem to fit reasonably well. Smoke aerosols are most often classified as less-hygroscopic with gHGFs of 1.11-1.33. We have added a short discussion:

“These estimates compare favourably to measured factors for biomass burning smoke (Swietlicky et al., 2008), indicating nearly-hydrophobic particles for boreal plumes, while temperate plumes could be classed as less-hygroscopic. Notably, measured geometric hygroscopic growth factors are reported at 90% relative humidity. In contrast, water volume fractions inferred in this study are representative of ambient humidity levels, and as a result direct comparison is not very meaningful.

The authors use “emission coefficients” in the text. Are these distinct from emission factors? If not, then rather use emission factors, as this is common terminology. If so, then please clarify the distinction in the text.

The term “emission coefficients” is used to contrast the top-down particulate emission estimates per unit of FRP with measured emission factors which are obtained for unit of fuel burned. Such nomenclature was used in other top-down studies (Kaiser et al., 2012; Ichoku and Ellison 2014). We have rephrased the paragraph introducing the concept:

“A~top-down global gridded Fire Energetics and Emissions Research (FEERv1) (Ichoku et al., 2014) product is based on collocated satellite FRP and AOT observations. Inferred total particular matter emissions rates are linked to observed FRP. The estimated TPM emission coefficients allow direct conversion from time integrated FRP to emitted particulate matter without invoking the emissions factors.”

Specific Comments:

Abstract: FRP is used in the Abstract (and Introduction), but the acronym is only defined on page 10.

We have replaced the acronym in abstract with “fire radiative power”, and added the definition to the first instance in the text.

Abstract, line 20: Is “low bias” meant to be “negative bias”?

Indeed, “negative bias” was the intended wording, changed accordingly.

Page 2, line 19: Is the 3.4 correction factor applied to address an underestimate or overestimate in emissions? Please be specific.

The correction factor was needed to address underestimation in emissions. The ambiguous statement has been changed to “enhancement factor”.

Page 2, lines 32-34: This sentence is confusing. Please reword for clarity. Currently it reads that average EFs conceal the lack of spatial and temporal representativeness. Is this what the authors mean to say?

The sentence has been rephrased to “Average EFs for different biomes are based on small sample numbers for some areas, and conceal large variability in individual measurements...”

Page 3, lines 2-3: Isn't the approach in this study also susceptible to AOT retrieval errors and uncertainties in smoke particle properties?

We perhaps didn't make this clear enough, but the statement was “top-down aerosol inversions are affected by AOT retrieval error and large uncertainties in assumed smoke particle properties”. The approach taken by this study is “top-down” by definition, therefore the statement was directed at our and other similar studies. To make it clear that it applies to any top-down method including this study, we have replaced “top-down aerosol inversions” with “top-down approaches”.

“Page 3, line 24: Please provide an appropriate reference or website for the MCD14ML data.

Reference (Giglio et al., 2006) has been added.

Page 4, lines 2-3: The approach used to identify fires is confusing. Are the “any pixels” pixels that include an active fire? Should there be consecutive active fire pixels within a 150 km radius? Please clarify in the text.

“any pixels” was used meaning any MODIS fire detections. This whole section has been rewritten for clarity as been requested by the other reviewer:

Large and long-lived fire events, likely strong emission sources, were identified and selected for the analysis. Burning episodes larger than 100km² are not numerous, but account for more than 80%

of total burned area in boreal North America (Stocks 2002, Kasischke 2002), and are a dominant mode of burning in parts of temperate regions as well (Strauss et al., 1989). In order to identify such events, individual MODIS active fire detections were agglomerated into large wildfire events by performing two step spatial-temporal clustering. First, any MODIS fire detections located closer than 10km in space and 24 hours in time were grouped together. Single detections not assigned to any of the formed clusters were removed from further analysis. The clusters were then filtered by selecting events with (i) spatial bounding box containing all fire detections belonging to the cluster larger than 100km² and (ii) duration longer than 7 days. The duration was determined by the time span between the first and the last MODIS active fire detection belonging to the cluster. The burning was considered uninterrupted if the largest temporal interval between subsequent MODIS fire observations was less than 24 hours. During the second step of clustering, any of the selected events active at the same time and located closer than 150km were grouped into large burning episodes, assigning a unique source label. These events were classified into boreal and temperate fires using the dominant emission source given in the GFEDv4 inventory for areas and periods when the events were active.

Page 5, lines 10-11: The change in resolution from nadir to the swath edges is true for the native resolution of the instrument, but the AOT product is at a nominal resolution of 10 km x 10 km.

It is our understanding that Collection 5 MODIS dark target algorithm uses blocks of 20 x 20 500m MODIS pixels while deep blue algorithm employs block 10 x 10 of 1 km MODIS pixels (Levy et al., 2013; Sayer et al., 2015). The block size is fixed across the swath, and hence the footprint increases proportionally with individual pixel size. The original statement in the manuscript “pixel size increases twice at the edge of the swath” was indeed incorrect as pointed out by the other reviewer. In fact, AOT pixel footprint increases approximately 9 times.

Page 5, line 12: More appropriate is the MODIS AOT uncertainty for scenes with aerosols from boreal and temperate fires. Please either estimate the error by comparing MODIS and AERONET AOT or provide a value reported in the literature.

We have added the regional analysis of Collection 5 MODIS AOT retrieval uncertainties (Hyer et al., 2011) to the paragraph:

“A regional MODIS M{^{*}}D04_L2 AOT product validation (Hyer et al., 2011) indicates that performance varies greatly within North America. The study found that for 0.2<AOT<1.4 conditions, root mean squared error varies from 0.01+0.51xAOT in arid Western America where retrieval is hindered by bright surfaces, to 0.01+0.31xAOT in boreal forests and 0.3+0.12xAOT in Eastern USA. The study reported positive bias in MODIS AOT for some locations, in particular for retrievals at extremely high aerosol loadings.”

Figure 3: What is the mass concentration of individual aerosol components (inorganic, organic, black carbon) from boreal and temperate fires estimated in this study?

While we did estimate volume fractions of black carbon and a mixture of organic and inorganic compounds represented by n values close to 1.53 when retrieving water content, a decision was taken not to present them in this manuscript. We believe that the discussion on different components should be left out of this manuscript which is focused on TPM estimates. In any case, the retrieved median volume fractions of black carbon were relatively low and hardly different when comparing the two categories, 0.01 and 0.009 for boreal and temperate plume observations respectively. Notably, we did not retrieve absorbing organic carbon or “brown carbon” fractions by utilising its wavelength dependent absorption. Organic carbon was represented as part of the third component encompassing organic and inorganic compounds.

Page 6, lines 6-7: This sentence is unclear. Is this merely a scaling to convert AOT to fire-emitted particle number?

The section has been updated clarifying the steps used when attributing AOT. Bellow is the relevant paragraph:

“If a mixture of particles was found within a cell indicating that multiple fires and multiple emission periods contributed towards the grid cell AOT, the attribution was performed by apportioning a grid cell's fire-emitted AOT in proportion to the numbers of modelled particles released during the emission periods and with origin found within the grid cell. For example, if a grid cell had AOT value of 1, and 100 HYSPLIT particles were located within the cell during the satellite overpass, 80 of which were emitted two diurnal cycles ago, and 20 during the previous diurnal cycle, the grid cell AOT was split accordingly between the emission periods. Panels K and L in figure 1 illustrate partitioning of total plume AOT to two different emission periods. Similarly, if there were any particles emitted from different fire events, grid cells AOT was divided both between different emission periods and different fire events.”

Table 1: Please change density to the Greek letter “rho” and enclose the units for density in square brackets for clarity.

Thank you for noting this, the symbol was changed accordingly.

Page 7, line 19: Why list both “organic carbon” and “organic matter”?

Thank you for pointing this out. The redundant phrase has been removed.

Page 8, lines 19-20: Incorrect in-text citation format for Ichoku and Ellison (2014). Please fix.

The citation has been fixed.

Page 10, lines 27-28: How different is median FRP for the two fire classifications if normalized to burned area or mass of biome burned?

We do agree that such a comparison would have been interesting. However, we did not employ area burned datasets in the study, and did not perform a comparison with biomass burned estimates. As a result, it is not feasible to include this information at this stage. In any case, the comparison of per-pixel FRP values is only a sideline to this study, and we did not place much weight on it, but felt the need to report it.

Page 10, line 33: Would smouldering fires be detected as part of the large wildfire events that are isolated in this work?

The authors are not aware of a method to directly detect smouldering combustion by remote sensing means. Median FRP values alone certainly do not provide enough information. However, smouldering combustion has been reported to be more important during night (Reid et al 2005); and that smoke from smouldering combustion can be lofted and entrained into main plumes by convection (Urbanski 2013). The statement is therefore merely an interpretation based on literature, not something which can be confirmed by the data employed. To avoid any possible confusion, We have removed the statement linking lower FRP and night time burning to smouldering combustion.

Page 11, line 15: What is “(4)” referring to? Is this Figure 4?

The reference to figure 4 has been corrected.

Page 11, line 17: Are the median water volume values fraction or percent? Fraction is stated, but the “%” symbol is used (page 12, line 1).

We apologise for this error. The median volume fractions were reported, not percentages. The symbol has been removed.

Figure 5: This figure is out of sequence in the text. It is referred to in the text before Figure 4. Please fix

Figure reference order has been corrected.

Page 14, line 17: Is 80% relative humidity?

Yes, relative humidity was meant. Added to the text.

Response to anonymous reviewer's #2 comments.

The authors thank the reviewer for their insightful and constructive comments. Our responses to anonymous reviewer's #2 comments are detailed below. Reviewer's comments are in italics and our responses in standard font.

General Comments:

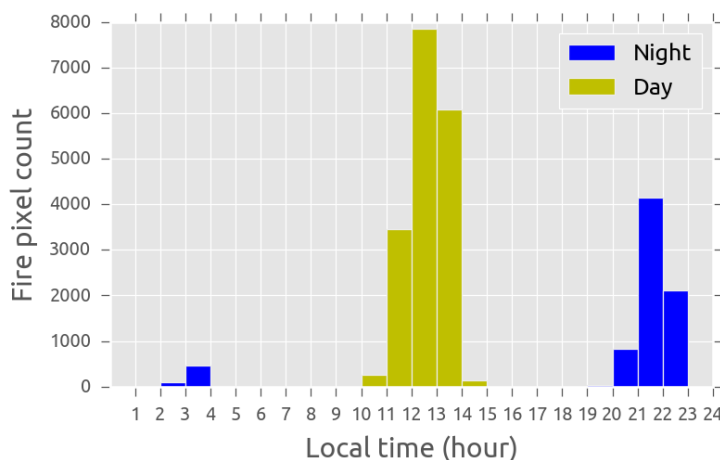
Your atmospheric simulation of smoke transport explicitly retains all smoke in the boundary layer. Wind shear in the vertical column, and other transport differences, will be a source of error in your estimates of smoke from many active fires that release smoke above the boundary layer (around 20% according to Val Martin ACP 2013 and Peterson 2014 JGR, but both of these estimates are based on satellite data with 1030am local overpass time, and thus likely conservative relative to overall fire behaviour). Is there any way these effects can be estimated with the data you have corralled for this study?

Indeed, restricting injection heights to the top of the planetary boundary layer is a limitation to the method. While the quantification of error and bias introduced by this limitation was not achieved, the authors expect that this effect is small when compared to other sources of uncertainty, both accounted and unaccounted for in the manuscript. Energetic burning episodes when smoke is injected directly into the free-troposphere can be expected to have significantly different transport pattern when compared to our within-PBL transport model output. And as a result, many of such cases should have been filtered out by the MODIS AOT and modelled plume extent matching step in the analysis. Consequently, this limitation should be primarily manifested as a selection bias in the results, excluding the most energetic events from the sample.

The difference between boreal and temperate fires' day-night behaviour is an interesting sidelight to this work. However, there is a good chance it is an artifact, and you must explore this before you finalize the paper. The basic idea is this: Terra and Aqua MODIS have nominal equatorial overpass times of 1030 and 1330 local solar time (LST, this can be calculated as $UTC + [longitude/15.]$, where longitude is from -180 to 180), with the opposite orbital nodes crossing at 2230 and 0130 respectively. At higher latitudes, the wide MODIS swath covers a larger range of LST. Thus, a portion of the 2230 Terra swath will have $LST < 2100$, and that portion will increase with latitude. So, if you define "daytime fires" as fires detected from 0900-2100LST, this will include all fires from the 1030 Terra overpass, all fires from the 1330 Aqua overpass, and depending on latitude, some fires from the 2230 Terra overpass. In order to avoid this, you should run the calculation using daytime=0600-1800LST, and see if the boreal-vs-temperate difference you observed holds up. I have attached a figure to illustrate this point, based on the MOD14 MODIS fire product.

The authors want thank the reviewer for this detailed comment. The problem here was that we did not state clearly how the daytime to night-time fire pixel counts were derived. The 0900-2100LST periods were used to emit particles, not to determine if an observation represents daytime or night-time burning. The authors were aware of the large spread of LST values for northern latitudes and therefore all fire detections from Terra overpasses with equatorial crossing time 2230 were considered to be night-time observations.

Please see the figure below, which shows counts of fire detections per local hour and day - night classification for all fire events analysed in the manuscript.



We have added a clarification on how fire detections were classified as daytime or night-time fires into section 2.33

“Emission source number and locations for daytime periods were determined from the highest number of fire detections observed during a single either Terra or Aqua daytime overpass with 10.30 and 13.30 equatorial crossing time. Similarly, emitted particle source number for the night periods were determined by the largest burning extent observed during one of the night-time overpasses with 22.30 and 1.30 equatorial crossing times. Notably, the Terra overpass at 22.30 in high latitudes makes observations of regions where local time is earlier than 21:00. In this study, however, all fires detected during this overpass were classed as night-time observations.”

page 4 Section 2.5 AOT attribution. This is the first of several very complicated steps, it is worth the effort to express very carefully how this was done. You have these ingredients: 1) modelled plume extent: this is a point cloud with the locations of all the smoke particle endpoints at solar noon on each day 2) MODIS AOT: you have the centroid location and retrieved AOT of each valid AOT retrieval on the day 3) background MODIS AOT: you have MODIS AOT and centroids for valid retrievals from two days prior to construct the background estimate As I understand it, you take these steps: 1) you interpolate modeled plume extent to 25km equal area grid, taking every grid cell that contains a portion of the plume and including it in the sample; 2) you interpolate MODIS AOD to the same 25km equal area grid; 3) you determine whether the number of valid same-day MODIS AOT data is at least $(\text{plume area} / 100\text{km}^2) \times 0.8$ (“80% coverage of plume area”)

If #3:

4) You calculate the background AOT using the 2-days-prior AOT 5) you calculate the smoke AOT increment for each grid cell by subtracting background AOT from same-day AOT If the median AOT increment is > 0: 6) you set negative AOT increments equal to zero 7) steps 1-6 are repeated for smoke transport times of up to 3 days, in increments of 12 hours. 8) If multiple days / multiple fires contribute to a grid cell AOT increment, you apportion the grid cell AOT increment to fire events and emission periods according to the number of smoke particles from the HYSPLIT simulation in each grid cell Note that the cutoff in Step 3 will systematically eliminate coverage from scenes covered by the MODIS swath edge, because the smoke retrievals will be too few to cover the area based on the assumed 100km² retrieval footprint.

Thank you for the detailed suggestions and advice on how to describe the method. This section is conveying very complex processing steps and we perhaps didn’t achieve sufficient clarity. The section has been rewritten following the above advice and clarifying what was not stated properly. It

is our belief that the method description now reads better and it is much more clear what was done in order to obtain our estimates:

“Elevated MODIS AOT observations were attributed to a specific fire event and emission period by comparing above background MODIS AOT retrievals to plume extent modelled by HYSPLIT (Fig. 1). The attribution required to determine three pieces of information; (i) event-specific background AOT value, (ii) modelled plume extent at local solar noon for each day of burning and (iii) coinciding MODIS AOT observations. First of all, background AOT value was estimated for each of the selected burning events. It was determined by the median value of the AOT retrievals within 150 km radii from the fire event centroid observed two days prior to ignition. For each day of fire activity, modelled plume extent (Fig. 1 (D–F)) was determined from the locations of all HYSPLIT particle endpoints at solar noon, and AOT observations (Fig. 1 (A–C)) from either Terra or Aqua platform with the highest spacial coverage for the day and plume area were selected.

After the required information was obtained, the following steps were performed for each day of burning attempting to estimate fire-emitted AOT. First, plume regions bounding the particles released during the previous three daytime and night-time emission periods were identified. Estimation of emission was attempted individually for each of the regions representing plume areas emitted during a specific time interval. This allowed the estimation of emitted AOT for up to three previous days from a single day of MODIS imagery. Importantly, such approach allows the estimation for some emission periods even if full MODIS plume overview is not available. Emitted AOT attribution was performed for the plume regions which satisfied two conditions. The region had (i) at least 80 % of MODIS AOT areal coverage assuming that a single AOT pixel represents 100 km² area, and (ii) with-region AOT median value was higher than the estimated background value for the fire event.

MODIS AOTs for the selected plume regions were interpolated to a 25 km resolution equal area grid (Fig. 1 (G–I)) by employing radial basis function interpolation with a ~linear kernel. Fire-emitted AOT were estimated by subtracting the background value from the within-plume AOT. The estimated fire-emitted AOT in every within-plume grid cell was apportioned to different emission periods and different sources based on information on release time and source of the HYSPLIT particles contained within the cell. If all particles found within a grid cell were released during the same emission period and originated from a single source, the cell's AOT was simply attributed to that emission period and source. If a mixture of particles were found within a cell, indicating that multiple fires and multiple emission periods contributed towards the grid cell AOT, the attribution was performed by apportioning a ~grid cell's fire-emitted AOT in proportion to the numbers of modelled particles released during the emission periods and with origin found within the grid cell. For example, if a grid cell had AOT value of 1, and 100 HYSPLIT particles were located within the cell during the satellite overpass, 80 of which were emitted two diurnal cycles ago, and 20 during the previous diurnal cycle, the grid cell AOT was split accordingly between the emission periods. Panels K and L in figure 1 illustrate partitioning of total plume AOT to two different emission periods. Similarly, if there were any particles emitted from different fire events, grid cell AOT was divided both between different emission periods and different fire events.”

page 2 Line 55: “Consumed biomass estimates inherit errors of fire location”. The papers cited here cover a lot of ground, but I don’t think they really cover errors associated with fire location. That source of uncertainty is described by Hyer and Reid (GRL, 2007).

Thank you for this suggestion, the reference has been added to support the relevant statement.

page 3 line 11 “larger than 100km² and with duration longer than 7 days” Please elaborate slightly on the data and calculations used for these determinations, especially the 7 days.

The paragraph has been updated detailing what was meant by “larger than 100km²” and “duration longer than 7 days”. Fire event size was determined by the size of the bounding box containing all fire detections for the event. Event duration was determined by the time span between the first and the last MODIS fire detection for the event. Burning episode was considered continuous if there were no 24h or longer gaps between the consecutive observations.

page 3 line 15 the Stocks and Kasischke papers relate to fire size distribution in the boreal forest. While the dominance of large fires has been documented for certain parts of temperate north America (see Strauss, Bednar, and Mees, Forest Science, 1989), it does not hold for all areas and in any event is not covered by those citations.

Thank you for noting this. The paragraph was changed clarifying: “Burning episodes larger than 100km² are not numerous, but account for more than 80% of total burned area in boreal North America (Stocks 2002, Kasischke 2002)” and “are a dominant mode of burning in parts of temperate regions as well (Strauss et al., 1989)

Page 3 line 25 “particles were continuously injected” HYSPLIT in your configuration simulates transport of discrete particles, please specify the interval at which particles were released in HYSPLIT Page 3 line 25 “vertically distributed” please specify the discrete intervals at which particles were released in HYSPLIT page 3 line 25 “within the planetary boundary layer” as diagnosed by GDAS? Please specify.

This section has been rewritten as requested by other reviewer, addressing all issues raised here as well. The specific points are clarified below.

“particles were continuously injected” has been changed to “20 particles released per hour per each active fire pixel within the fire event”.

“vertically distributed” was changed to “uniformly distributed between the surface and the top of the boundary layer as given in the GDAS archive”

page 4 line 6 “is about twice the size at swath edges.” Actually, the single MODIS pixels increase roughly 8x in size from nadir to swath edge, and the 20x20 pixel footprints used by MxD04_L2 increase proportionally. However, there is significant overlap between MxD04_L2 footprints at swath edge, see Sayer et al. (<http://www.atmos-meastech.net/8/5277/2015/>). You may not need to quantitatively account for this for this study, but you should be aware of this.

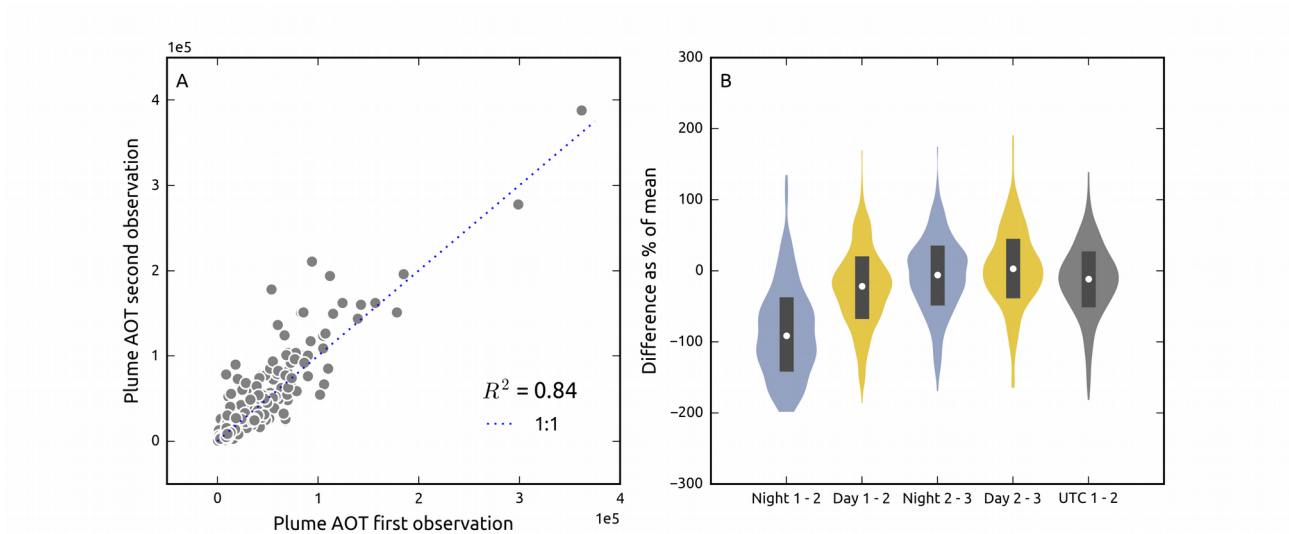
Thank you for noting this discrepancy. MODIS pixel size is indeed ~9 times larger at the edge of the swath as demonstrated in the suggested study (Sayer et al., 2015). The relevant paragraph has been updated stating “...10 x 10 spatial resolution at nadir. MODIS pixel size increases with view angle, and pixels at the edge of the swath are approximately 9 times larger.

Page 4 Section 2.4 how were AOT data selected from MxD04_L2 (quality flags, cloud fraction, etc.)?

The selection criteria were added to the text “all retrievals with quality assurance confidence > 0 were selected. To maximise coverage, no cloud fraction filtering was applied.”

Page 7 line 21: “the agreement between two or more estimates for the same emission period is reasonably static across the plume age categories.” I do not see where this is shown in figures or tables. If it is there somewhere, please direct the reader to it when you make this statement. One simple change would be to add a second panel to Figure 5 showing the agreement between Day 1 and Day 3 AOT for the same event/time pairs.

Indeed this statement was not supported by any of the figures. As suggested, we have now included panel B to Figure 5 showing difference between two estimates obtained at different stages of plume development. The statement now refers to the new figure.



Minor corrections and typos

Thank you for taking time to find these mistakes. The authors apologise for leaving them in. All of these now have been corrected.

Particulate emissions from large North American wildfires estimated using a new top-down method

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Abstract. Particulate matter emissions from wildfires affect climate, weather and air quality. However, existing global and regional aerosol emission estimates differ by a factor of up to 4 between different methods. Using a novel approach, we estimate daily total particular matter (TPM) emissions from large wildfires in North American boreal and temperate regions. Moderate Resolution Imaging Spectroradiometer (MODIS) fire location and aerosol optical thickness (AOT) datasets are coupled with HYSPLIT atmospheric dispersion simulations, attributing identified smoke plumes to sources. Unlike previous approaches, the method (i) combines information from both satellite and AERONET observations to take into account aerosol water uptake and plume specific mass extinction efficiency ~~in~~when converting smoke AOT to TPM, and (ii) does not depend on instantaneous emission rates observed during individual satellite overpasses, which do not sample night-time emissions. The method also allows multiple independent estimates for the same emission period from imagery taken on consecutive days.

Repeated fire-emitted AOT estimates for the same emission period over two to three days of plume evolution show ~~increases~~increase in plume optical thickness by approximately 10 % for boreal events, and by 40 % for temperate emissions. Inferred median water volume fractions for aged boreal and temperate smoke observations are 0.15 and 0.47 respectively, indicating that the increased AOT is partly explained by aerosol water uptake. TPM emission estimates for boreal events, which predominantly burn during daytime, agree closely with bottom-up Global Fire Emission Database (GFEDv4) and Global Fire Assimilation System (GFASv1.0) inventories, but are lower by approximately 30 % compared to Quick Fire Emission Dataset (QFEDv2) PM_{2.5}, and are higher by approximately a factor of 2 compared to Fire Energetics and Emissions Research (FEERv1) TPM estimates. The discrepancies are larger for temperate fires, which are characterised by lower median ~~FRP~~fire radiative power values and more significant night-time combustion. The TPM estimates for this study for the biome are lower than QFED PM_{2.5} by 35 %, and are larger by factors of 2.4, 3.2 and 4 compared with FEER, GFED and GFAS inventories respectively. Large underestimation of TPM emission by bottom-up GFED and GFAS indicates low biases in emission factors or consumed biomass estimates for temperate fires.

1 Introduction

Large and often severe fires in boreal and temperate forest regions alter atmospheric composition, considerably affecting the Earth's radiative budget (Langmann et al., 2009; Bond et al., 2013) and degrading air quality (Johnston et al., 2012).

Burning regime in these regions is dominated by episodic extreme events (Stocks et al., 2002) emitting continental scale

plumes (Colarco et al., 2004) with inter-hemispheric transport potential (Damoah et al., 2004; Dahlkötter et al., 2013). Future climate predictions indicate both dryer conditions and greater than average warming for northern latitudes, projecting a likely increase in area burned (Liu et al., 2010) and soil carbon consumption (Turetsky et al., 2015). For the quantification of smoke radiative forcing and impacts on human health, a realistic representation of biomass burning emissions in climate and air quality models is needed. Disagreement between bottom-up and top-down emission estimates of particulate matter, however, remains large (Kaiser et al., 2012; Ichoku and Ellison, 2014).

Bottom-up emission inventories use emission factors (EF) (~~Andreae and Merlet, 2001; Janhäll et al., 2010; Akagi et al., 2011; ?~~)(~~Andreae and Merlet, 2001; Janhäll et al., 2010; Akagi et al., 2011; ?~~) ratios of gases and particulate matter emitted per unit of dry fuel burned, compiled for different biomes from a range of burning experiment measurements across the globe. Emission factors are applied to biomass burned estimates which are typically based on satellite observations of ubiquitous but highly variable fire activity. The Global Fire Emission Database (GFED) (Van der Werf et al., 2010) makes use of satellite burned area products (Randerson et al., 2012; Giglio et al., 2013) and active fire pixel counts, while the Global Fire Assimilation System (GFAS) (Kaiser et al., 2012) employs fire radiative power (FRP) measurements (Giglio et al., 2006). Burned area estimates are converted to biomass burned using modelled carbon pools and soil moisture dependent combustion completeness characteristic to the fuel types. FRP based methods rely on observed relationships between observed FRP and biomass combustion rates (Kaufman and Tanre, 1998; Wooster et al., 2003, 2005).

The more top-down methods utilize satellite aerosol optical thickness (AOT) observations. The Quick Fire Emission Database (QFED) uses regional AOT measurements to scale emissions based on EFs (Darmenov and da Silva, 2015). Similarly, atmospheric model assimilation of GFAS emissions (Kaiser et al., 2012) suggested a 3.4 global ~~correction~~ enhancement factor was needed to reconcile TPM estimates with observed AOTs. Purely top-down methods estimate emissions through inverse modelling of satellite AOT retrievals (Ichoku and Kaufman, 2005; Dubovik et al., 2008). A top-down global gridded Fire Energetics and Emissions Research (FEERv1) (Ichoku and Ellison, 2014) ~~emission-coefficients~~ product is based on collocated satellite FRP and AOT observations. ~~The product allows~~ Inferred total particular matter emissions rates are linked to observed FRP. The estimated TPM emission coefficients allow direct conversion from time integrated FRP to emitted particulate matter without invoking the emissions factors.

Global and regional particulate matter estimates from the bottom-up burned area and fire pixel count based GFED agree well with the FRP based GFAS estimates. Model assimilation of these bottom-up emissions, however, suggest TPM underestimation by a factor of 2 to 4, compared to satellite AOT observations (Kaiser et al., 2012). Enhanced GFAS TPM estimates and scaled QFED agree better with top-down FEER emission coefficients on global scales. Notable discrepancies, however, are present for individual regions. North American emissions are larger for enhanced GFAS TPM and QFED when compared to top-down FEER, while FEER agrees closely with the bottom-up GFED inventory.

A number of uncertainties in both bottom-up and top-down estimates can contribute towards the apparent TPM discrepancies. Average EFs for different biomes conceal ~~the lack of spatial and temporal representativeness~~ small sample numbers for some areas, and large variability in individual measurements ~~introduced by results from~~ within-biome inconsistencies in vegetation density, climatic and burning conditions (Van Leeuwen and Van der Werf, 2011; Van Leeuwen et al., 2014). Consumed biomass estimates inherit errors of satellite burned area (Randerson et al., 2012), fire location (Hyer et al., 2009) or FRP

retrieval (Giglio et al., 2006), and depend on a range of assumptions on availability and consumption of carbon in aboveground and soil pools (French et al., 2004). Top-down ~~aerosol-inversions~~ approaches are affected by AOT retrieval error and large uncertainties in assumed smoke particle properties, which are required to relate aerosol extinction to particulate mass (Reid et al., 2005b). Moreover, estimates of emission rates based on near source retrievals are representative of burning conditions at the time of satellite overpass. A recent study indicated that night-time TPM emissions might be underestimated by a factor of 20 - 30 for a large temperate forest fire in Western USA (Saide et al., 2015), stressing the need for better representation of night-time emissions in the inventories. Methods based on regional AOT observations, on the other hand, must take into account poorly constrained ageing effects (Reid and Hobbs, 1998; O'Neill et al., 2002).

This study presents estimates of particulate matter emissions from large wildfires with identifiable plumes in North American boreal and temperate regions. A newly developed top-down method is applied which attributes satellite aerosol observations to a specific fire event and emission period. Quantified daily fire-emitted AOT takes into account aerosols injected throughout the diurnal cycle and does not rely on instantaneous emission rates observed during a satellite overpass. In some cases, AOT attribution for the same emission period is achieved from satellite images taken on successive days, allowing assessment of uncertainty and investigation of systematic changes in plume optical thickness over time. Total particulate matter is quantified by applying mass extinction efficiency which is simulated using AERONET particle properties, and accounting for inferred water uptake by aerosols. The results are compared with existing estimates in order to investigate systematic differences between the approaches.

2 Data and methods

Daily total particulate matter emissions for large and persistent fire events were estimated by combining Moderate Resolution Imaging Spectroradiometer (MODIS) active fire observations and aerosol optical thickness retrievals with plume dispersion simulated using the Hybrid Single-Particle Lagrangian Integrated (HYSPLIT) model.

2.1 Active fires

To represent fire activity we used the active fire location dataset MCD14ML produced by the University of Maryland and provided by NASA Fire Information for Resource Management System (Giglio et al., 2006). The data product is based on MODIS mid-range and thermal infrared observations. MODIS sensors are flown on board the sun-synchronous polar-orbiting Terra and Aqua satellites passing the equator at 10:30 and 13.30 local time during the daytime hours, and 22.30 and 1.30 at night respectively. The instruments have a wide swath of approximately 2330 km, each providing nearly global coverage daily. For high latitudes the coverage is better due to increasing overlap between consecutive overpasses. Each detection in the dataset represents an active fire in a 1 km² pixel at the time of satellite overpass, and contains information on the retrieved fire radiative power.

2.2 Fire event selection

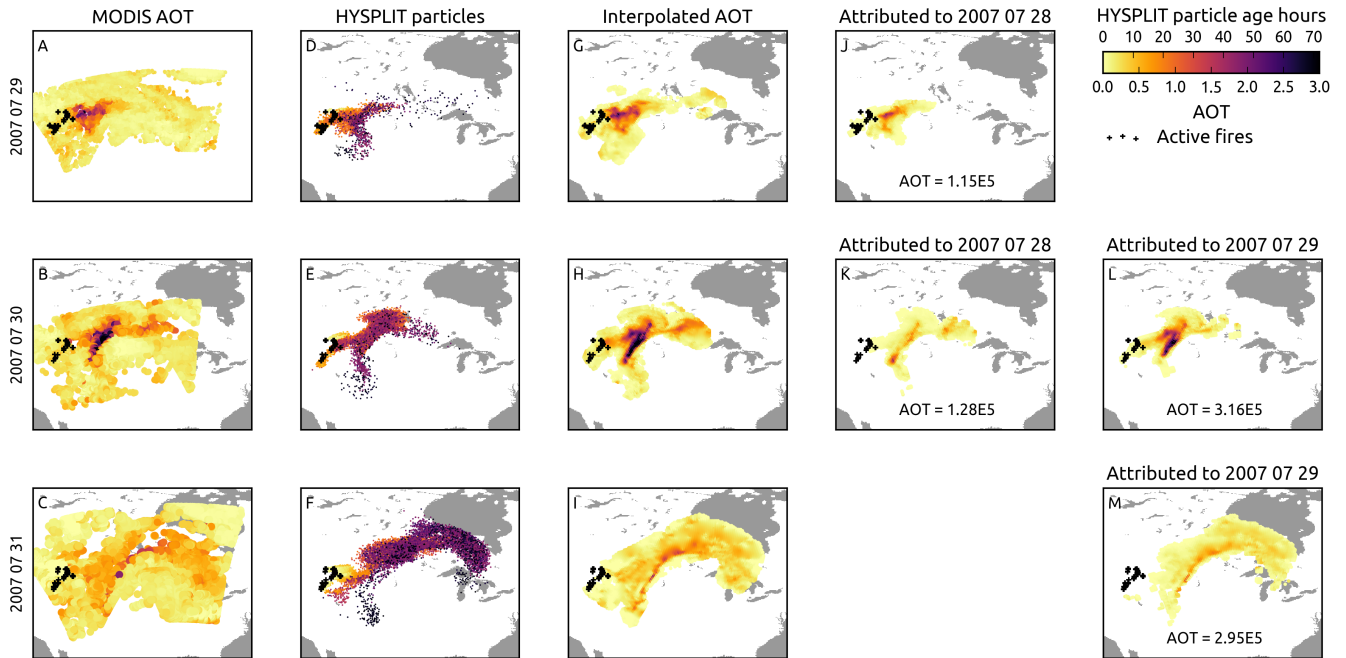


Figure 1. An illustration of the method showing an example of fire-emitted AOT attribution for two diurnal cycles of a temperate fire. Rows in the figure represent three successive days of satellite imagery from which the attribution was achieved. Columns from left to right show MODIS AOT retrievals for the day from a single platform with the highest coverage (A–C), snapshots of HYSPLIT particle positions and age taken at local noon (D–F), and AOT interpolated to 25 km equal area grid (G–I). The two right columns show fire-emitted AOT attributed to 28th (J) and (K) and 29th (L) and (M) of July 2007 determined from images taken on different days. Total attributed AOT is shown within the plots.

~~The active fire identifications~~ Large and long-lived fire events, likely strong emission sources, were identified and selected for the analysis. Burning episodes larger than 100 km² are not numerous, but account for more than 80% of total burned area in boreal North America (Stocks et al., 2002; Kasischke et al., 2002), and are a dominant mode of burning in parts of temperate regions as well (Strauss et al., 1989). In order to identify such events, individual MODIS active fire detections were

5 agglomerated into large wildfire events by ~~grouping any pixels performing two step spatial-temporal clustering~~. First, any MODIS fire detections located closer than 15010 km in space and 24 hours in time ~~. For the analysis we selected events were grouped together. Single detections not assigned to any of the formed clusters were removed from further analysis. The clusters were then filtered by selecting events with (i) spatial bounding box containing all fire detections belonging to the cluster larger than 100 km² and with (ii) duration longer than 7 days, as they were likely to be strong emission sources. Fires~~

10 of such size or larger are a dominant mode of burning in North American boreal and temperate forests contributing more than 80to total burned area in these regions (Stocks et al., 2002; Kasischke et al., 2002). Burning. The duration was determined by

the time span between the first and the last MODIS active fire detection belonging to the cluster. The burning was considered uninterrupted if the largest temporal interval between subsequent MODIS fire observations was less than 24 hours. During the second step of clustering, any of the selected events active at the same time and located closer than 150 km were grouped into large burning episodes, assigning a unique source label. These events were classified into boreal and temperate fires using the dominant emission source given in the GFEDv4 inventory for areas and periods when the events were active.

2.3 Plume dispersion modelling

Smoke ~~dispersion-transport~~ for the selected fire events was simulated with the HYSPLIT model (Draxler and Rolph, 2003). ~~The model was run using~~ Plume dispersion from a source location was represented by the motion of a large number of discrete particles moved by the wind field with mean and random components. Global Data Assimilation System (GDAS) meteorological archive data ~~was employed to drive the model.~~

For each day of burning, particles were continuously ~~injected and vertically distributed within the planetary boundary layer. Particle number was scaled proportional to active fire pixel counts observed during the satellite overpasses. For each diurnal cycle when the fire event was active, particles were injected at two different rates released into the model domain from the~~ locations of the individual active fire detections within the fire event. In order to represent fire diurnal cycle, different MODIS active fire observations were used to release particles for two 12 hour intervals representing day and night emissions 09:00 to 21:00 and 21:00 to 09:00 local time respectively. ~~The injection rates were scaled in proportion to the highest detected fire pixel count from either Aqua or Terra overpasses during the timeperiods. When no fires were detected the count was~~ Emission source number and locations for daytime periods were determined from the highest number of fire detections observed during a single either Terra or Aqua daytime overpass with 10.30 and 13.30 equatorial crossing time. Similarly, emitted particle source numbers for the night periods were determined by the largest burning extent observed during one of the night-time overpasses with 22.30 and 1.30 equatorial crossing times. Notably, the Terra overpass at 22.30 in high latitudes makes observations of regions where local time is earlier than 21:00. In this study, however, all fires detected during this overpass were classed as night-time observations. If no valid observations were available for some of the time intervals, the count and fire pixel locations were set to a minimum ~~positive non-zero~~ value estimated for the ~~fire event burning episode~~ from all daytime or night-time observations. ~~Modelled particle positions~~ This was done to avoid total temporary shut-down of the emissions, which is an unlikely scenario for a long duration burning episodes. Every hour, 20 particles were released for each fire pixel. As a result, emitted particle number for a burning episode was determined by the number of active fire pixels observed during a given time period.

Particles were uniformly distributed between the surface and the top altitude of the planetary boundary layer as given in GDAS archive. Satellite based plume height estimates (Val Martin et al., 2010; Peterson et al., 2014) indicate that in up to 80 % of the events analysed, injection heights were limited to the planetary boundary layer. While confinement of the emissions to the mixing layer underestimates injection height for the most energetic burning episodes, such configuration should nonetheless represent the majority of burning episodes.

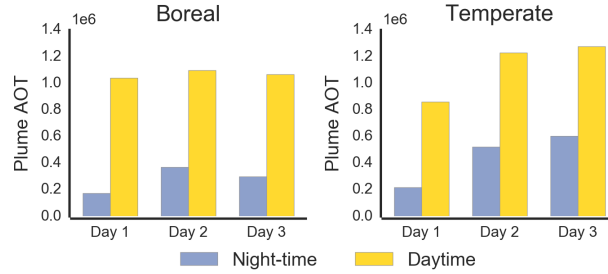


Figure 2. Changes in attributed AOT over time. Shown are 39 boreal and 37 temperate diurnal emission cycles for which estimates were obtained on three consecutive days, for both daytime and night-time periods.

Throughout the simulations, modelled particle positions, their age and source burning event identifier were recorded each day at local solar noon were-. The generated point clouds were later used to compare against Terra and Aqua Aerosol Optical Thickness (AOT) observations.

2.4 Satellite aerosol data

- 5 MODIS AOT collection 5.1 data products M*D04_L2 were used in this study. The dark target algorithm (Kaufman and Tanre, 1998; Levy et al., 2009) retrieves AOT at 550nm and 10km × 10km spatial resolution at nadir. Pixel-MODIS pixel size increases with view angle and is about twice the size at swath edges-, and pixels at the edge of the swath are approximately 9 times larger. For this study, all M*D04_L2 AOT retrievals with quality assurance confidence > 0 were selected. To maximise coverage, no cloud fraction filtering was applied. The AOT product global validation against ground-based AERONET AOT
- 10 observations suggest a one sigma error which increases linearly with aerosol loading $\pm(0.05 + 0.15\%)$ (Levy et al., 2010)-. The AOT $\pm(0.05 + 0.20\%)$ (Levy et al., 2010) for overland cases. A regional MODIS M*D04_L2 AOT product validation (Hyer et al., 2011) indicates that performance varies greatly within North America. The study found that for $0.2 < \text{AOT} < 1.4$ conditions, root mean squared error varies from $-0.01 + 0.51 \times \text{AOT}$ in arid Western America where retrieval is hindered by bright surfaces, to $0.01 + 0.31 \times \text{AOT}$ in boreal forest and $0.3 + 0.12 \times \text{AOT}$ in Eastern USA. The study reported positive
- 15 bias in MODIS AOT for some locations, in particular for retrievals at extremely high aerosol loadings. The AOT retrieval values have an upper limit of 5.0, and in addition, opaque smoke is often rejected as bright surface or cloud by the algorithm (Livingston et al., 2014), preventing retrievals over extremely optically dense plumes. Consequently, AOT near the emission source is often not retrieved and the algorithm performs better when plumes are dispersed into regional haze.

2.5 AOT attribution

- 20 Elevated MODIS AOT observations were attributed to a specific fire event and emission period by comparing regional retrievals to particle positions above background MODIS AOT retrievals to plume extent modelled by HYSPLIT (Fig. 1). Attribution required three pieces of information: (i) event-specific background AOT value, (ii) modelled plume extent at local solar noon

for each day of burning and (iii) coinciding MODIS AOT observations. First of all, background AOT value was estimated for each of the selected burning events. It was determined by the median value of the AOT retrievals within 150 km radius from the fire event centroid observed two days prior to ignition. For each day of fire activity, ~~MODIS AOT observations from either Aqua and Terra platforms~~ modelled plume extent (Fig. 1 (D–F)) was determined from the locations of all HYSPLIT particle endpoints at solar noon, and AOT observations (Fig. 1 (A–C)) from either Terra or Aqua platform with the highest spacial coverage for the day ~~were matched with modelled plume extent. The matching was performed iteratively for the modelled plume regions dominated by particles emitted and plume area were selected.~~

After the required information was obtained, the following steps were performed for each day of burning attempting to estimate fire-emitted AOT. First, plume regions bounding the particles released during the previous ~~1 to 6 twelve hour emission periods~~, representing three full diurnal cycles, three daytime and night-time emission periods were identified. Estimation of emission was attempted individually for each of the regions representing plume areas emitted during a specific time interval. This allowed the estimation of emitted AOT for up to three previous days from a single day of MODIS imagery. Importantly, such an approach allows the estimation for some emission periods even if full MODIS plume overview is not available. Emitted AOT attribution was performed for the plume regions ~~with which~~ satisfied two conditions; (i) ~~the region had~~ at least 80 % of ~~area with available AOT retrievals from either Aqua or Terra platforms; MODIS AOT areal coverage~~ assuming that a ~~single MODIS single~~ AOT pixel represents 100 km² area, and (ii) ~~with with-region~~ AOT median value ~~was~~ higher than the estimated background value for the fire event. ~~The background value for a fire event was determined from the median value of the AOT retrievals in the fire region observed two days prior to ignition.~~

MODIS AOTs for the selected ~~cases~~ ~~plume regions~~ were interpolated to a 25 km resolution equal area grid (Fig. 1 (G–I)) ~~by~~ employing radial basis function interpolation with a linear kernel. ~~The background value was subtracted from Fire-emitted AOT were estimated by subtracting the background value from the~~ within-plume AOTs ~~and negative values set to zero. AOT retrievals above the background value were attributed AOT.~~ The estimated fire-emitted AOT in every within-plume grid cell was apportioned to different emission periods and different sources. ~~The based on information on release time and source of the HYSPLIT particles contained within the cell. If all particles found within a grid cell were released during the same emission period and originated from a single source, the cell's AOT was simply attributed to that emission period and source. If a mixture of particles were found within a cell, indicating that multiple fires and multiple emission periods contributed towards the grid cell AOT, the~~ attribution was performed by ~~partitioning~~ apportioning a grid cell's fire-emitted AOT in proportion to the numbers of modelled particles ~~emitted-released~~ during the emission periods and with origin found within the ~~cell~~ grid cell. For example, if a grid cell had AOT value of 1, and 100 HYSPLIT particles were located within the cell during the satellite overpass, 80 of which were emitted two diurnal cycles ago, and 20 during the previous diurnal cycle, the grid cell AOT was split accordingly between the emission periods. Panels K and L in figure 1 illustrate partitioning of total plume AOT to two different emission periods. Similarly, if there were any particles emitted from different fire events, grid cell AOT was divided both between different emission periods and different fire events.

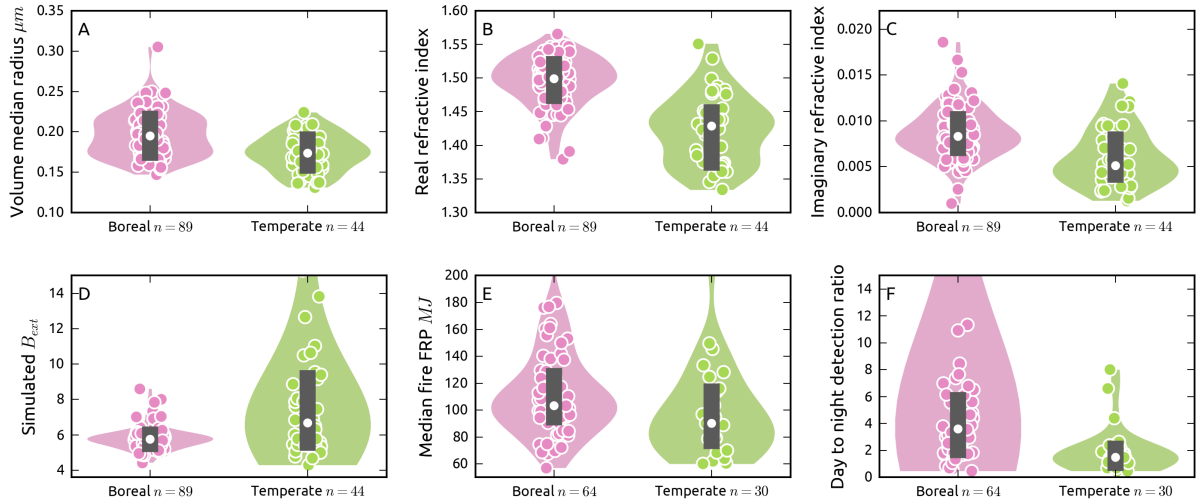


Figure 3. Distributions of AERONET-retrieved particle properties (A–C) attributed to boreal and temperate fires, simulated mass extinction efficiencies (D), median MODIS FRP values (E) and ratios of daytime to night-time active fire detection counts observed during a single overpass (F). Shown are kernel density estimates and individual observations; boxes indicate median values and interquartile range.

2.6 Smoke aerosol properties

The Aerosol Robotic Network (AERONET) (Holben et al., 1998) level 2 retrievals (Dubovik and King, 2000) of aerosol microphysical and optical properties were used to characterise particles in plumes under investigation. AERONET consists of ground-based globally distributed sun-sky scanning photometers with a narrow field of view. The instruments are continuously monitored and calibrated, and the retrieved properties have estimated accuracy ranges. The direct sun beam extinction measurements provide spectral AOT at several wavelengths ranging from 0.34 to 1.02 μm with uncertainty of 0.01 - 0.02 (Dubovik et al., 2000). Measured AOT and angular distribution of sky radiances are used to retrieve column integrated aerosol volume size distribution at 22 size bins from 0.05 to 15 μm and spectral refractive index at 0.44, 0.67, 0.87 and 1.02 μm . Size retrieval is expected to be accurate within 25 % for particles with radii between 0.1 and 7 μm and 25–100 % for size bins outside this range. Scans at high aerosol loadings (AOT 0.44 $\mu m \geq 0.4$) allow retrieval of refractive index with estimated uncertainties of 0.04 and 30 % for real and imaginary parts respectively (Dubovik et al., 2000).

Available observations within areas identified by the dispersion analysis as biomass burning plumes were attributed to a specific emission event and land cover type. Only retrievals containing refractive index (AOT 0.44 $\mu m \geq 0.4$) were selected. In order to minimize the presence of dust and urban aerosol dominated retrievals, cases with volume concentration of fine mode (particle diameter < 1 μm) fraction less than 0.8, sphericity parameter lower than 0.98 and absorption Ångström exponent lower than 1 were filtered out. To make the samples more representative of plumes for which particulate matter was estimated, we selected AERONET observations within-plume areas dominated by particles aged for 1 to 3 days.

Table 1. Real (n) and imaginary (k) parts of refractive index, and density (p) of the components used in the Maxwell-Garnett effective medium approximation calculations. All components were assumed to have spectrally flat refractive index. Uncertainty in p for the species represented by the second inclusion was propagated into combined errors of retrieved water volume fraction and particle density.

Species	n	k	$p \text{ g/cm}^3$	Source
Black carbon	1.95	0.79	1.8	Bond and Bergstrom (2006)
Organic and inorganic compounds	1.53	0.00	1.2–1.4	Kirchstetter et al., (2004), Turpin and Lim (2001), Toon et al., (1976)
Water	1.33	0.00	1.0	

2.7 Water content retrieval

The available AERONET spectral refractive indices were used to infer smoke aerosol water uptake. We employed the Maxwell-Garnett effective medium approximation (Bohren and Huffman, 1983) which provides a method to derive volume fractions of the components in the mixture if their refractive indices are known. The approach is described in detail and demonstrated by Schuster et al. (2005) for retrieving black carbon concentrations from AERONET climatologies. It was further developed to infer brown carbon content (Arola et al., 2011), aerosol water uptake (Schuster et al., 2009), and to simultaneously retrieve fractions of carbonaceous absorbers and dust (Schuster et al., 2016).

To infer water content we employed a three component mixture of black carbon and organic-inorganic matter inclusions in water host (table 1). For black carbon we assumed the refractive index and density suggested in Bond and Bergstrom (2006). The second inclusion was used to represent a broad range of chemical species observed in biomass burning plumes (Brock et al., 2011), including organic carbon, ~~organic matter~~, ammonium sulphate and ammonium nitrate. These species were represented by a single component because they have n values close to 1.53. This value is characteristic to dry ammonium sulphate (Toon et al., 1976), was measured for organic carbon (Kirchstetter et al., 2004) and lies within the range of values measured for dry organic compounds (Dick et al., 2007). Volume fractions of the inclusions and water host were retrieved in two steps. First, we deduced black carbon utilising spectral imaginary refractive index of the component. The Maxwell-Garnett mixing rule was applied for a range of different fractions of black carbon in water host with negligible imaginary index. Volume fraction of the inclusion was estimated determining the configuration which provided minimum χ^2 :

$$\chi^2 = \sum_{i=1}^N \frac{(k_i^{ret} - k_i^{mg})^2}{(k_i^{ret})^2}, \quad (1)$$

where k_i^{ret} is AERONET-retrieved imaginary index, k_i^{mg} is the value calculated by the Maxwell-Garnett mixing rule, i is summation over the selected AERONET wavelengths. We used AERONET k at 0.87 and 1.02 μm to retrieve black carbon fraction, assuming that it is the only absorber at this part of the spectrum. k at shorter wavelengths can be enhanced by absorption by organic carbon (Kirchstetter et al., 2004), which is retrieved as a part of the second inclusion. After volume

fraction of black carbon was established, we kept it fixed and varied the fraction of the second inclusion in the mixture, minimizing the equation (1) for real part of the refractive index at all four AERONET wavelengths.

2.8 Conversion of aerosol optical thickness to mass

Particle mass within the atmospheric column can be inferred from smoke AOT observations if mass extinction efficiency (B_{ext}) is known:

$$M_{plume} = \frac{\tau_{plume}}{B_{ext}}, \quad (2)$$

where M_{plume} is mass of plume aerosols, and τ_{plume} is a product of mean fire-emitted AOT and plume area. B_{ext} represents extinction in area units per unit of aerosol mass, usually expressed as $[m^2/g]$. It can be measured or calculated invoking Mie theory. In-situ measurements of fresh North American smoke suggest B_{ext} values ranging from 3.9 to 4.6 m^2/g (Hobbs et al., 1996). Equivalent measurements for aged plumes are not available for the region, but smoke samples collected in other forest ecosystems indicate slightly larger B_{ext} values ranging 4.0 to 5.3 m^2/g (Reid et al., 2005b; Chand et al., 2006) for older emissions. Similar B_{ext} at 550nm ranging from 4.5 to 5.2 m^2/g were inferred by Reid et al. (2005b) from AERONET retrievals (Dubovik et al., 2002) of dominant particle size distributions and index of refraction for North American boreal regions. ~~(Ichoku and Ellison, 2014)~~ Ichoku and Ellison (2014) applied a uniform 4.6 m^2/g value (Reid et al., 2005b) in deriving FEER TPM emission coefficients. Notably, plumes in their analysis were relatively young, up to a few hours old at most. In contrast, smoke discussed in this study is aged for few days.

To avoid making assumptions on smoke optical properties, B_{ext} was inferred utilizing available AERONET-retrieved refractive indices and particle size distributions. We used Mie code (Bohren and Huffman, 1983) to calculate B_{ext} assuming spherical internally mixed particles:

$$B_{ext} = \frac{\int_{r_{min}}^{r_{max}} \sigma_{ext}(n, k, \lambda, r) \frac{dN(r)}{d \ln r} d \ln r}{\int_{r_{min}}^{r_{max}} \sigma_{ext}(n, k, \lambda, r) \frac{dN(r)}{d \ln r} d \ln r} \frac{\int_{r_{min}}^{r_{max}} V_{dry} \rho_{dry} \frac{3}{4} \pi r^3 \frac{dN(r)}{d \ln r} d \ln r}{\int_{r_{min}}^{r_{max}} V_{dry} \rho_{dry} \frac{3}{4} \pi r^3 \frac{dN(r)}{d \ln r} d \ln r}, \quad (3)$$

where σ_{ext} is the extinction ~~cross-section~~ cross-section of a single particle which depends on refractive indices (n, k), wavelength and particle radius (r). V_{dry} is particle dry volume fraction, ~~ρ_{dry}~~ ρ_{dry} is particle dry fraction density, both determined from aerosol water uptake analysis (section 2.7). σ_{ext} was calculated at 0.55 μm using Mie code for every radius in the AERONET size distribution and averaged n and k retrievals at 0.44 and 0.67 μm . The numerator in the equation 3 is single particle extinction ~~cross-sections~~ cross-sections integrated over number distribution, while denominator is aerosol dry fraction mass within the column given by the product of particle density and integrated particle volume.

2.9 Uncertainty in derived quantities

Uncertainties in AERONET smoke aerosol properties, particle density and daily fire-emitted AOT attribution were propagated using a Monte Carlo method retrieving water volume fraction, mass extinction efficiency and deriving total TPM estimates for the biomes. Throughout the study we report median values and interquartile range for the distributions, unless otherwise stated.

5 3 Results and interpretation

Attribution of fire-emitted AOT for at least two diurnal cycles of emission was achieved for 94 large fire events. Boreal sources constitute 64 of the events, with the remaining identified as temperate forest fires. In total, fire-emitted AOT estimates were obtained for 620 days of burning. The daily attributed AOT include particulate matter emitted during the full diurnal cycle of emission accounting for both daytime and night-time emissions. These estimates are representative of large and likely intense burning events and clear sky conditions for which sufficient satellite observations were available. Particulate matter emitted by the events on the days for which our estimates were obtained account for approximately 3 to 20 % of total GFED and GFAS emissions for the North America region depending on the year. The representativeness, however, is probably better than suggested by this figure, assuming that emissions from the sampled events were similar on the days for which estimation was not achieved.

15 3.1 Systematic changes in plume attributed AOT

An important advantage of the AOT attribution method presented in this study is that it allows us to gauge combined error originating from uncertainties in plume injection height, dispersion modelling, MODIS AOT retrievals and applied interpolation. Critically, any systematic changes in fire-emitted smoke optical thickness in evolving plumes can be inferred as well. This was facilitated by a number of cases when two or more AOT attributions based on imagery taken on consecutive days were performed for the same emission period. Figure 2 shows daily AOT estimates for days of emission for which the attribution was achieved from imagery taken on three consecutive days, for both night-time and daytime emission periods.

Overall, determined smoke AOT based on retrievals at later stages of plume development tend to have a positive bias compared to estimates for the same period of emission obtained on previous days. Notably, the largest increase in estimated AOT is observed when comparing estimates for the previous night-time emission cycle (smoke aged for 3 to 15 hours) to AOT attributed to the same period determined from the following day's imagery, after the plume has aged for an additional 24 hours. Inferred changes in daytime fire-emitted AOT over the first two days of ageing are smaller. Optical thickness for temperate smoke increases by approximately 30 % from the first observation of daytime emissions which are already aged for 15 to 27 hours, compared to estimates for the same emission period determined from the imagery collected the following day. Changes in estimated daytime fire-emitted AOT for boreal plumes appears to be negligible. Notably, consecutive 24 hours of ageing does not change estimated plume AOT significantly for both biomes and both daytime and night-time emissions. A slight decrease in optical thickness is observed for boreal smoke, but this should be treated with caution given the level of

uncertainties involved. For the limited number of emission cycles presented in figure 2, contributions of day and night emissions appear to differ between the biomes. Night-time emissions constitute 30–40 % of total fire-emitted AOT for temperate events. Boreal plumes are dominated by daytime emissions with night-time emissions comprising under 20 % of total daily AOT. The difference is influenced by generally larger number of night-time active fire pixels observed for temperate fires (Fig. 3 (F)) and, consequently, more particles released during night-time emission period in the dispersion simulations.

The effect of increasing AOT over time could be in part explained by uncertainty in plume dispersion modelling. However, the modelling error is expected to increase with time and hence should be manifested by progressively larger disagreement and biases for older estimates. In contrast, the results suggest that the agreement between ~~two~~the estimates for the same emission period is ~~smaller at later stages of plume development~~reasonably static across the plume age categories (Fig. 4 (B)). The bias, on the other hand, is clearly largest for the first and the second plume observations within the first two diurnal cycles. It is possible that the model-emitted night-time particles get mixed with subsequent daytime emissions during the transport, effectively scavenging part of AOT from the other emission periods during the attribution. However, the observed daytime AOT tends to increase as well. Additionally, there are significant differences in inferred AOT changes between boreal and temperate plumes, indicating that some physical processes might be driving the change.

Particulate matter estimation and comparison with other methods are based on fire-emitted AOT during emission cycles starting and ending at 00.00 UTC. For 159 and 125 emission periods for boreal and temperate events respectively, AOT was determined from imagery taken on consecutive days allowing us to estimate the attribution error. These estimates do not include the problematic previous night emissions. Figure 4 (A) shows the differences in fire-emitted AOT estimates for these cases. Given that the differences are approximately normally distributed, we propagated 50 % one sigma uncertainty in attributed daily fire-emitted AOT to derive confidence intervals for TPM emission estimates.

3.2 Fire FRP and daytime - night-time pixel counts

Large and persistent fire events discussed in this study exhibit distinctiveness in ~~fire radiative power (FRP)~~FRP values and diurnal burning cycle. Median MODIS FRP retrieved for the boreal fires is 103 (94–117) MW, while median FRP for temperate events is 90 (78–103) MW. This suggests higher burning intensity and combustion rates for boreal fires. A more striking difference, however, emerges when comparing ratios of maximum active fire pixel counts detected during individual daytime and night-time satellite overpasses. The proportion of active fires at night are typically much higher for temperate fires. The average daytime to night-time pixel count ratio is 1.4 (1.1–1.9) for the fires in this biome compared to median value of 3.6 (1.8–4.8) for boreal fires. Such a pattern indicates a higher contribution of night burning for temperate events~~and potentially more important smouldering combustion~~.

3.3 Variability in particle properties

The identified AERONET observations of boreal and temperate smoke suggest distinctiveness in retrieved size distributions and refractive index (Fig. 3 (A–C)). The selected observations indicate that boreal emissions tend to have larger particles with median volume median radius value of 0.19 (0.17–0.21) compared to 0.17 (0.16–0.19) μm obtained for temperate smoke.

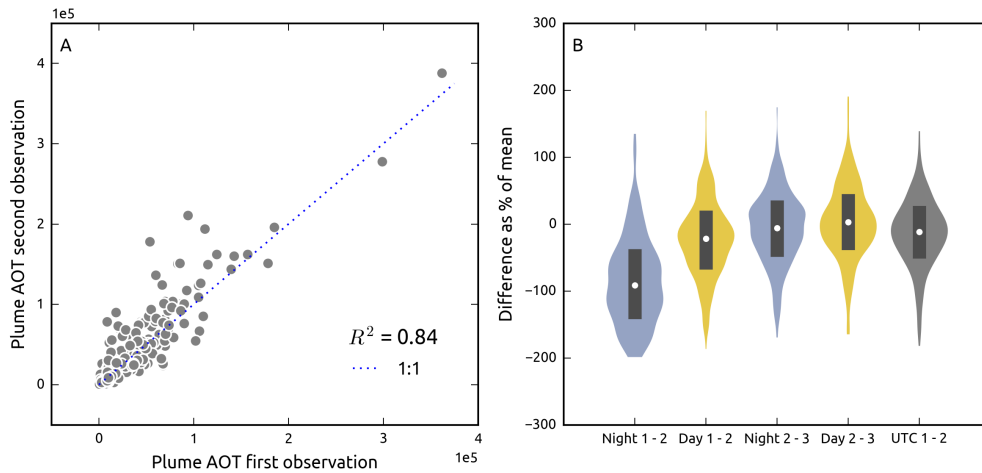


Figure 4. ~~Inferred volume fraction~~ Fire-emitted AOT for 284 cases with two estimates for the same diurnal emission period starting and ending at 00:00 UTC, obtained at different stages of ~~water~~plume development (A). ~~Error bars show interquartile range~~ Plot (B) shows differences between two fire-emitted AOT estimates expressed as a percentage of ~~inferred values resulting from uncertainties~~ their mean value for the three night-time and daytime emission periods and two UTC periods shown in AERONET particle properties panel A.

These differences may be influenced by differences in combustion phase between the biomes. Very intense and predominantly flaming fires emit larger particles than events with more important smouldering combustion (Reid et al., 2005a). Substantial differences exist comparing the complex index of refraction. Boreal plumes exhibit higher median n value of 1.49 (1.47–1.52) in contrast to 1.43 (1.37–1.45) observed for plumes attributed to temperate forest fires. Although boreal smoke generally is more absorbing with median k value 0.008 (0.007–0.01) compared to the 0.005 (0.004–0.008) value obtained for temperate emissions, plumes from both biomes are only weakly absorbing and characteristic k values have a negligible influence on calculated B_{ext} . Variability in the real part of the refractive index between the plume categories, on the other hand, is larger and indicates differences in particle chemistry.

3.4 Inferred volume water fractions

10 3.5 ~~Inferred volume water fractions~~

Maxwell-Garnett medium approximation calculations using the discussed optical constants result in substantially different inferred water content for the two sources (Fig. 5). The variability is mainly driven by the real part of the refractive index. Inferred median black carbon fractions are less than 1 % for both classes and thus have minimal impact on water content retrieval. Median water volume fraction for boreal fires is 0.15 (0.1–0.31), whereas temperate plumes have median value of 0.47 (0.29–0.67). The derived values agree with water volume fractions inferred by Schuster et al. (2009) using a similar approach, although dust was not included as one of the components in our retrieval. Converting the inferred median water

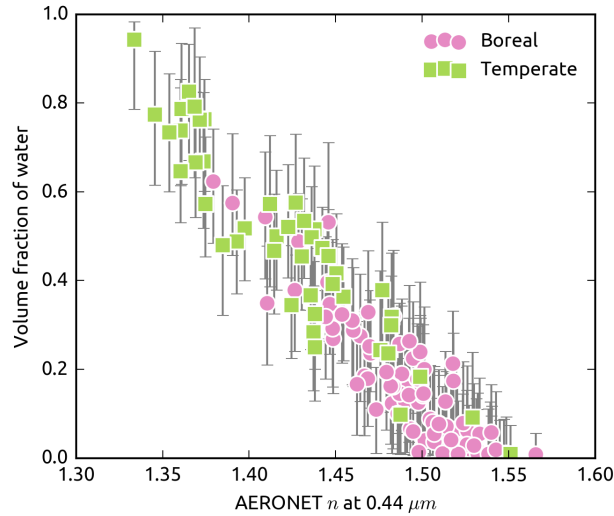


Figure 5. Fire-emitted AOT for 284 cases with two estimates for the same diurnal emission period starting and ending at 00:00 UTC, obtained at different stages. Inferred volume fraction of plume development water. Inset shows distribution. Error bars show interquartile range of the differences between the estimates as a percentage of their mean value. Inferred values resulting from uncertainties in AERONET particle properties.

volume fractions to geometric hygroscopic growth factors results in values of 1.05 and 1.24 for boreal and temperate plumes respectively. These estimates compare favourably to measured factors for biomass burning smoke (Swietlicki et al., 2008), indicating nearly-hydrophobic particles for boreal plumes, while temperate smoke could be classed as less-hygroscopic. Notably, measured geometric hygroscopic growth factors are reported at 90 % relative humidity. In contrast, water volume fractions inferred in this study are representative of ambient humidity levels, and as a result direct comparison is not very meaningful.

The main limitations of the presented method are (i) the assumption that aerosols with $n \geq 1.53$ are dry and (ii) large uncertainties in the chosen n values and different components used in the retrieval. In addition to increasing water content, formation of organic compounds may alter aerosol optical properties. Measured n for dry ambient organic aerosol are typically lower than the 1.53 value used in this study, ranging from 1.47 to 1.53 (Dick et al., 2007) and appear to change with age (Rudich et al., 2007). Although the uncertainties in AERONET properties and particle density were propagated in the retrieval, water fractions inferred in this study critically depend on n of the dry major component being close to 1.53. Any departures from this value result in inaccurate water uptake retrieval.

3.5 Simulated mass extinction efficiencies

The differences in plume particle properties, primarily n and particle size, coupled with distinctiveness in inferred volume water fractions, drive differences in simulated B_{ext} for the dry volume content of the plumes. Boreal plumes have larger

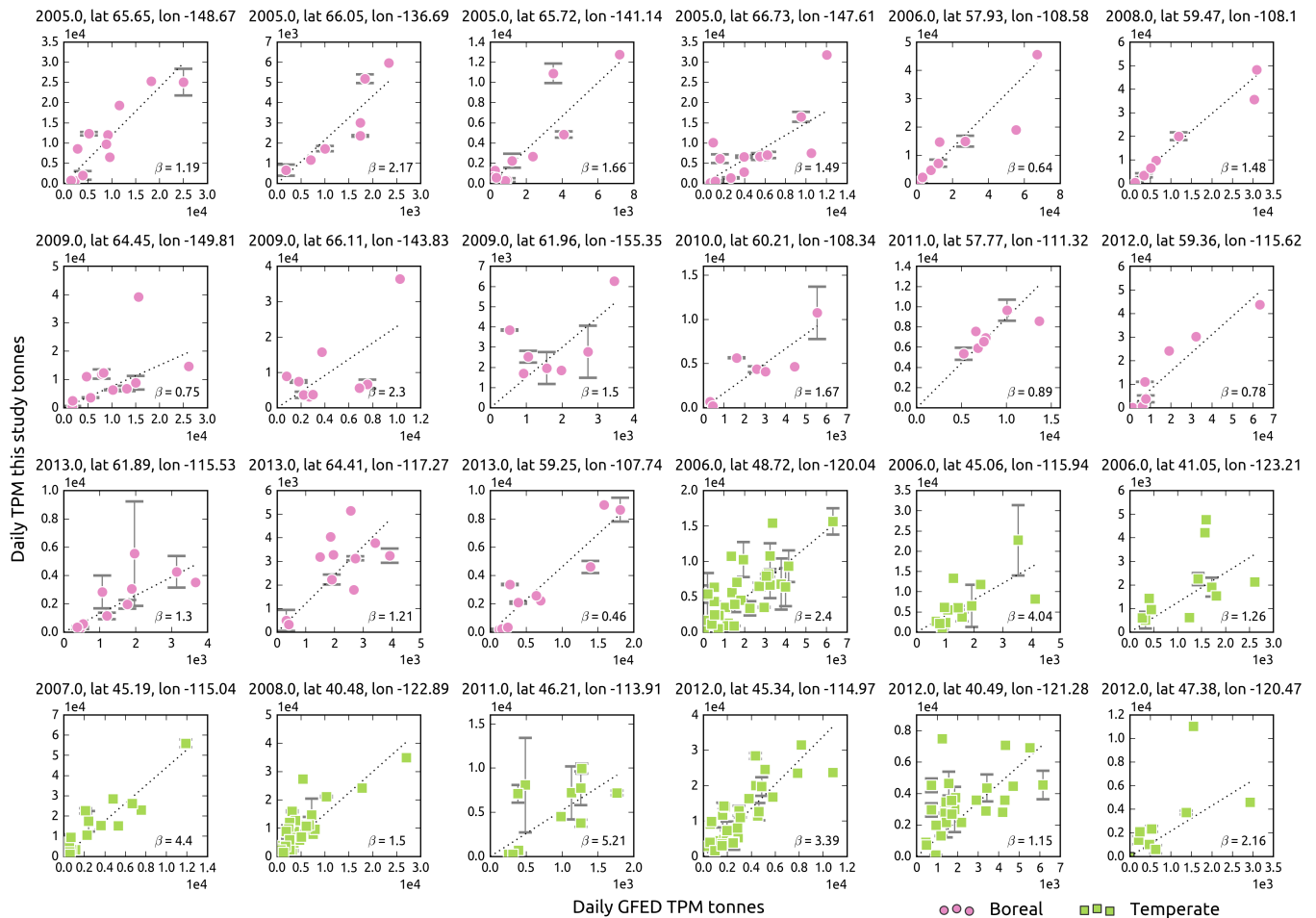


Figure 6. Daily estimated TPM from this study and GFED for individual fire events. Error bars represent difference between two TPM values for the days of emission for which two estimates were obtained. Shown are robust linear fits, β parameter indicates the slope.

particles, higher values of refractive index, but smaller water fractions and hence have lower median B_{ext} value of 5.7 (5.1–6.5), while emissions originating from temperate forests have a median B_{ext} value of 6.7 (5.4–9.2) m^2/g due to inferred greater water content. The identified AERONET observations are for ambient plumes which are aged for at least 1 to 3 days, and consequently, computed B_{ext} values for dry volume fractions are larger than the $4.7 \pm 0.7 \text{ m}^2/\text{g}$ value suggested for dry aged

5 boreal and temperate emissions (Reid et al., 2005b). Somewhat higher values ranging from 4.7 to 5.5 m^2/g were calculated (Reid et al., 2005b) for a set of AERONET retrievals from North American boreal forest (Dubovik et al., 2002). The main difference between that aerosol climatology and the retrievals used in this study are in the real part of the refractive index. Dubovik et al. (2002) climatology for boreal smoke represents generally dryer plumes with an average n value of 1.5, compared to 1.49 and 1.43 median n values attributed to boreal and temperate emission in this study.

3.6 Interpretation of changes in smoke optical thickness

The increase in attributed AOT in aged plumes determined in this study is consistent with well documented smoke particle evolution. Aerosols grow considerably in size as plumes age. Particles undergo rapid changes during the first few hours after emission due to combined effects of condensation and coagulation (Reid and Hobbs, 1998), with reported growth rates in volume median radius as high as $0.04\text{ }\mu\text{m}$ per hour (Hobbs et al., 1996). On the time scales of days plume particles continue to grow in dense plumes but at substantially lower rates, primarily due to coagulation and hygroscopic growth. Reported increases in volume median radius at these time scales are in the order of $0.02\text{--}0.03\text{ }\mu\text{m}$ (Reid et al., 2005a; Nikonovas et al., 2015). Condensation of organic and inorganic species and secondary particle production increase particle plume mass, while coagulation only transforms particle distribution. Both processes alter smoke optical thickness mainly by enlarging scattering cross-section and scattering efficiency, which is a strong function of particle size. Condensation has been reported to increase particle mass by up to 30 - 40 % in Amazonian plumes, but is thought to be important only during the first 24 hours at most (Reid and Hobbs, 1998). The inferred increase in fire-emitted AOT over the first two days of ageing reported in this current study only partially overlaps with this period. The first few hours of plume development when condensation is thought to be the most active are not represented, therefore condensation is unlikely to contribute significantly towards the inferred AOT growth. A growth in volume median radius of $0.02\text{ }\mu\text{m}$ due to coagulation theoretically could increase scattering efficiency by up to 30 % without changes in plume mass, but this process can not explain differences in the magnitude of AOT change observed between the biomes.

An additional factor driving changes in AOT is water uptake by smoke particles. Absorption of water depends on air relative humidity and aerosol solubility which in turn tends to increase with atmospheric processing. It increases particle size further, enhancing scattering cross-section. Hygroscopic growth factors measured and inferred by optical methods for biomass burning smoke at 80 % [relative humidity](#) range from 1.1 to more than 2 (Kotchenmther and Hobbs, 1998; Kreidenweis et al., 2001; Magi and Hobbs, 2003). Reid et al. (2005b) suggested an average enhancement factor of 1.35 ± 0.2 . B_{ext} values derived for dry volume fraction in this study suggest median scattering cross-section enhancement factors of 1.2 and 2 for boreal and temperate plumes, assuming the 4.7 B_{ext} value for dry smoke (Reid et al., 2005b).

Notably, the magnitude of AOT increase over time shown in figure 2 corresponds to inferred median water fractions for the two biomes. Temperate emissions exhibit generally hydrophilic particles with much greater water content, while boreal plumes seem to contain much less aerosol water. This distinctiveness could be due to different ratios of smouldering and flaming combustion. Field measurements indicate that prescribed burns and in particular wildfires in temperate regions have lower combustion efficiencies ([\(Urbanski, 2014\)](#)). Temperate fires discussed in this study have lower mean FRP values and a less pronounced diurnal burning cycle, and the emitted plumes have higher ratios of night-time emissions. Smouldering night-time smoke has been reported to contain more soluble organic compounds (Hoffer et al., 2006), which could explain the presence of more hydrophilic aerosols in temperate plumes. In addition, factors not accounted for in this study, such as significant differences in relative humidity and atmospheric processing between the biomes, may be partly responsible for the inferred variability in water uptake.

3.7 Daily TPM estimates for individual fires

On an individual event basis the relationships between daily particulate emissions given by the global inventories and this study exhibit varying degrees of agreement. Figure 6 shows this study's and GFED TPM for the events for which estimation was performed for at least seven diurnal cycles. Although some fires exhibit only fair or weak agreement, the result is nonetheless encouraging considering error in AOT attribution and conversion to TPM method in this study, and large uncertainty associated with the [date of burn in](#) daily burned area product (Giglio et al., 2013) on which GFED depends. Robust linear fits between GFED TPM and daily estimated TPM shown in figure 6 indicate considerable variability in slopes, even comparing the events with generally good agreement. This suggests distinctive combustion and emission characteristics for individual events. As well as variability on a per burning event basis, large differences exist when comparing relationships for fires in boreal and temperate forests. Notably, for every tonne of GFED TPM, this study shows TPM ranges from 0.46 to over 2 tonnes for boreal burning events, while for temperate fires the conversion factors range from approximately 1 to more than 5. The relationships are similar in terms of agreement comparing daily TPM estimates with other inventories (not shown), but scaling factors which are needed to reconcile the estimates differ.

3.8 Comparison of total emissions and emission coefficients

Total TPM emission estimates obtained in this study for the wildfires examined are large in comparison to FEER, and to a lesser degree GFED and GFAS inventories, but are smaller than QFED estimates (Fig. 7). QFED emissions are reported for PM_{2.5} aerosol fraction only, which typically constitutes 70 to 85 % of TPM for the biomes discussed (Akagi et al., 2011). As a result, QFED TPM estimates should be approximately 20–40% higher than indicated in figure 7.

Substantial differences exist comparing the estimates for boreal and temperate fires. For boreal forest events, total TPM emissions for this study are in close agreement with the bottom-up GFED and GFAS TPM estimates. The agreement indicates that application of the proposed 2.2 enhancement factor (Kaiser et al., 2012) to GFAS TPM would overestimate boreal emissions for the events discussed. [In fact, assuming increase in aerosol mass and AOT in ageing plumes, boreal TPM emissions for this study are low when compared the near-source GFED and GFAS estimates.](#) Regional AOT based QFED inventory suggests PM_{2.5} emissions higher by 40%, while near-source FEER TPM estimates are smaller by a factor of 2.8 when compared to TPM for this study.

For temperate forests, a striking contrast exists between GFED and GFAS inventories and methods based on regional AOTs. The largest estimates are given by the QFED inventory, which suggests PM_{2.5} emissions ~~which~~ are higher by 50% than the TPM estimates for this study. If bottom-up estimates of the boreal emissions agree well with this study's TPM, for temperate events the discrepancies are much larger. Scaling factors of 3.2 and 4 are needed to reconcile GFED and GFAS emissions with the estimates obtained in this study. FEER emissions are closer to bottom-up approaches suggesting much lower emitted TPM compared to the other top-down methods. This appears to be characteristic to North America as has been reported in Ichoku and Ellison (2014), indicating potential underestimation of the emissions in the region. For other continents, FEER generally predict higher TPM emissions than the bottom-up inventories and agree closely or even exceed QFED PM_{2.5} estimates.

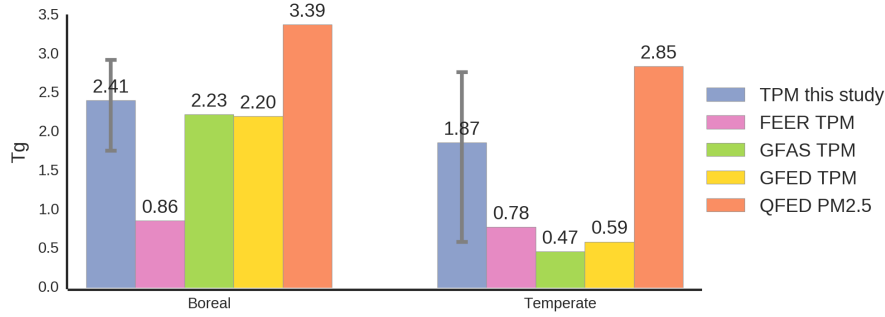


Figure 7. Total TPM emissions derived in this study and estimates for the same events and days of emission given by other methods. Error bars represent 95 % confidence interval determined taking into account uncertainties in (i) AERONET retrievals, (ii) inferred water fraction, (iii) particle density, (iv) modelled B_{ext} , and (v) estimated error in attributed daily AOT.

Table 2. Total particulate matter emission coefficients derived using GFASv1.0 FRP product and particulate matter emission estimates for the burning events discussed.

	Emission coefficients (gMJ^{-1})	
	Boreal	Temperate
TPM this study	27 (25–30)	31 (24–37)
FEER TPM	10	12
GFAS TPM	25	8
GFED TPM	25	10
QFED PM _{2.5}	38	47

The above emission budgets suggest particulate matter emission coefficients of 27 (23–30) and 31 (24–37) g per MJ^{-1} of time integrated GFASv1.0 FRP (table 2). They comprise approximately 70 % of coefficients derived for QFED PM_{2.5} emissions, and are 2.5 times larger than equivalent values derived using FEER emission coefficients. Notably, although differing in magnitude, all three top-down methods indicate slightly larger emission coefficients for temperate events. In contrast, more bottom-up approaches suggest 2.5 to 3 times larger emission coefficients for boreal forest. The TPM emissions factors employed in GFAS and GFED inventories are identical for both forest types, but large differences exist in consumed biomass estimates. The GFAS inventory employs a three-fold larger FRP to dry matter combustion rate conversion factor for boreal fires, attributable to high organic soil content in the biome. [In contrast, emission coefficients for boreal and temperate forests derived in this study are statistically indistinguishable.](#)

A number of factors may contribute towards the discrepancies between TPM estimates for this study and other methods. Relatively large estimates compared to near source GFED, GFAS and FEER inventories may be influenced by unaccounted processes in ageing plumes. A several-fold growth in plume mass due to condensation and secondary particle production, however, seems implausible given that the reported magnitude of increase in particle mass driven by these processes is within 50 % (Reid and Hobbs, 1998). The difference may be partly due to large sizes of the events sampled in this study. Field measurements for large wildfires are scarce (~~Akagi et al., 2011; ?~~)(Akagi et al., 2011; Urbanski, 2014), and such events are under-represented in compiled EFs. The agreement between top-down and bottom-up methods is better for boreal fires than it is for temperate events. Fire events sizes are similar for both biomes, at least for the fires sampled. Therefore, it seems that fire size considerations alone fail to explain the varying degree of agreement between this study's TPM, and GFED and GFAS estimates when comparing boreal and temperate cases. Comparably low FEER estimates, ~~on the other hand~~ apart from ageing effects, might be partly determined by sampled event size. Infrequent and large fires prevailing in North American forests make it difficult to reliably derive combustion coefficients from near the source imagery (Ichoku and Ellison, 2014).

Considering the above factors it seems that for the large fire events discussed, boreal emissions are underestimated by a factor close to 2 by FEER inventory. Temperate TPM appears to be underestimated by factors of 2 to 4 by FEER, GFED and GFAS. QFED on the other hand, ~~seem~~ seems to overestimate particulate emissions by 40 to 50 %. The previously suggested GFAS TPM 2.2 enhancement factor seems to represent an average value for the region. It is not required for boreal fires, and is close to 4 for temperate plumes. ~~It is not clear if~~ While the underestimation by bottom-up GFED and GFAS ~~is~~ may be driven by low emissions factors ~~or, the magnitude of the difference indicates that~~ biomass consumed estimates ~~More important smouldering combustion in temperate forests, however, suggests larger emission factors for this biome. Current measurements imply large underestimation for night-time emissions (Saide et al., 2015), suggesting the need for further investigation~~ are the most likely source for discrepancies.

4 Conclusions

Refined particulate matter emission estimates are needed to improve future climate simulations and predict regional air quality at shorter time scales. Existing global estimates differ by a factor of 2–4. The method presented in this study enables the estimation of daily TPM emissions from large wildfires with identifiable plumes and sufficient satellite AOT observations. Daily estimates take into account particulate matter emitted throughout a full diurnal cycle including both daytime and night-time emissions. Importantly, repetitive estimates are obtained for the same period of emission during up to three consecutive days of plume evolution allowing assessment of the AOT attribution error and systematic changes in smoke optical thickness over time.

Important insights are gained by partitioning plume AOT to daytime and night-time emissions. Night-time plume AOT seems to double when comparing observations of relatively young emissions of up to 18 h age to AOT attributed to the same period of emission from the following day's imagery. Only small changes are observed after the subsequent 24 hours of ageing. Daytime emitted AOT increases by approximately 30 % for temperate fires, but does not change over time in boreal smoke.

These changes have to be accounted for when reconciling emission estimates obtained near the source and from regionally dispersed aged plumes.

We utilized available coinciding AERONET observations to infer characteristic aerosol water content in discussed plumes and parametrize Mie calculations of smoke mass extinction efficiency. Coinciding AERONET retrievals indicate median water volume fractions of 0.15 (0.1–0.31) and 0.47 (0.29–0.67) for boreal and temperate plumes respectively. Calculated B_{ext} of the dry particle fraction suggest median values of 5.7 (5.1–6.5) and 6.5 (5.5–9.2) ~~m^2/g~~ m^2/g for the two plume categories. The inferred water fractions indicate that hygroscopic growth accounts for the majority of the observed increase in plume optical thickness.

Daily total particulate matter emissions determined using simulated B_{ext} indicate differences in agreement with other inventories for the two forest type fires. For boreal fires which have higher median FRP values and burn predominantly during the daytime, TPM estimates agree closely with GFED and GFAS inventories, are higher by a factor of 2 compared to FEER, and are lower by 30 % than QFED $PM_{2.5}$ estimates. For temperate events, which are characterised by small changes in active fire pixel count throughout the diurnal cycle and generally lower median FRP values, the discrepancies are larger. Our TPM estimates are lower than QFED $PM_{2.5}$ by 35 %, and higher by a factor of 4, 3.2 and 2.4 compared to GFAS, GFED and FEER TPM estimates for the same emission events. The previously suggested scaling factor of 2.2 for GFAS particulate emissions is not required for boreal fires, but is too small for temperate events.

The large fire event bias ~~in this study~~ and rapid ageing effects unaccounted for in this study could drive part of the difference, but are unlikely to explain all of it. Low FEER TPM for the discussed events could be attributed to these factors to a larger extent. The comparison of TPM obtained in this study to GFAS and GFED, however, suggest that TPM emission factors and consumed biomass estimates are underestimated for temperate fires within the bottom-up datasets.

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