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Daily burned area and carbon emissions from boreal fires in Alaska

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Abstract

Boreal fires burn carbon-rich organic soils, thereby releasing large quantities of trace gases and aerosols that influence atmospheric composition and climate. To better understand the factors regulating boreal fire emissions, we developed a statistical ⁵ model of carbon consumption by fire for Alaska with a spatial resolution of 500 m and a temporal resolution of one day. We used the model to estimate variability in carbon emissions between 2001 and 2012. Daily burned area was mapped using imagery from the Moderate Resolution Imaging Spectroradiometer combined with perimeters from the Alaska Large Fire Database. Carbon consumption was calibrated ¹⁰ using available field measurements from black spruce forests in Alaska. We built two nonlinear multiplicative models to separately predict above- and belowground carbon consumption by fire in response to environmental variables including elevation, day of burning within the fire season, pre-fire tree cover and the differenced normalized burn ratio (dNBR). Higher belowground consumption occurred later in the season and for

- ¹⁵ mid-elevation regions. Aboveground and belowground consumption also increased as a function of tree cover and the dNBR, suggesting a causal link between the processes regulating these two components of consumption. Between 2001 and 2012, the median fuel consumption was 2.48 kgCm⁻² and the median pixel-based uncertainty (SD of prediction error) was 0.38 kgCm⁻². There were considerable amounts of burning in other cover types than black spruce and consumption in pure black spruce stands was generally higher. Fuel consumption originated primarily from the belowground
- fraction (median = 2.30 kgCm^{-2} for all cover types and 2.63 kgCm^{-2} for pure black spruce stands). Total carbon emissions varied considerably from year to year, with the highest emissions occurring during 2004 (67 TgC), 2005 (44 TgC), 2009 (25 TgC),
- ²⁵ and 2002 (16 TgC) and a mean of 14 TgC per year between 2001 and 2012. Our analysis highlights the importance of accounting for the spatial heterogeneity within fuels and consumption when extrapolating emissions in space and time. This data on



daily burned area and emissions may be useful for in understanding controls and limits on fire growth, and predicting potential feedbacks of changing fire regimes.

1 Introduction

- Fire is the most important landscape disturbance in the boreal forest (Chapin et al., 2000; Krawchuk et al., 2006). Increases in the extent and severity of burning in the last several decades have been reported for Alaska and Canada (Gillett et al., 2004; Kasischke and Turetsky, 2006; Kasischke et al., 2010; Turetsky et al., 2011). With accelerated warming predicted for the boreal region during the remainder of the 21st century (Collins et al., 2013), intensification of the fire regime is expected (Amiro et al., 2009; Balshi et al., 2009; de Groot et al., 2013). The interactions between the boreal ecosystem, fire and climate are complex. Boreal fires have both positive and negative climate feedbacks (Randerson et al., 2006; Bowman et al., 2009; Oris et al., 2014). Cooling is primarily caused by the increased surface albedo due to more persistent snow cover and early successional vegetation in burned areas mainly occurring during the spring and winter months (Randerson et al., 2006; Rogers et al., 2013), and organic performance of the spring and winter months (Randerson et al., 2014). The aminipine of graenbauge general and the spring and winter months (Randerson et al., 2014).
- carbon aerosol effects (Tosca et al., 2013). The emission of greenhouse gases and black carbon aerosols (Bowman et al., 2009), and the deposition of black carbon on snow and ice (Flanner et al., 2007) are the dominant warming feedbacks.

The magnitudes of these feedbacks are tightly linked with the severity of the disturbance (Beck et al., 2011a; Turetsky et al., 2011; Jin et al., 2012). Severity is often referred to in a general way describing the amount of environmental damage that fire causes to an ecosystem (Key and Benson, 2006). In the context of mostly stand-replacing boreal fires, severity is expressed as the degree of consumption of belowground organic matter. Differences in ground layer burn depths control the amount of carbon combusted, and impact post-fire succession trajectories and consequent albedo feedbacks (Johnstone and Kasischke, 2005; Johnstone et al., 2010; Jin et al., 2012). Heterogeneity in fuels, fuel conditions, topography and fire



weather can result in different post-fire effects over the landscape. Resolving the spatial heterogeneity in severity using post-fire remote sensing observations can improve emissions estimates (Veraverbeke and Hook, 2013; Rogers et al., 2014) and more accurate carbon emissions estimates could lower the uncertainties in estimating the net climate feedback from boreal fires under the current and future climate (Oris et al., 2014).

Fire emissions are generally calculated as the product of burned area, fuel consumption and emission factors. (Seiler and Crutzen, 1980; van der Werf et al., 2010). Fuel consumption represents the amount of biomass consumed by the fire, and gas-specific emission factors describe the amount of gas released per unit of biomass

- gas-specific emission factors describe the amount of gas released per unit of biomass consumed by the fire. Examples of this approach include the Wildland Fire Emissions Information System (WFEIS, French et al., 2011) and the Global Fire Emissions Database version 3 (GFED3, van der Werf et al., 2010), updated with contributions of small fires (GFED3s, Randerson et al., 2012). Several similar approaches have been
- ¹⁵ developed specifically for boreal forests (Kasischke et al., 1995; Amiro et al., 2001; Kajii et al., 2002; Kasischke and Bruhwiler, 2002; French et al., 2003; Soja et al., 2004; de Groot et al., 2007; Tan et al., 2007; Kasischke and Hoy, 2012). The quantification of fuel consumption in boreal emission models is often driven by empirical relationships between fire weather variables and combustion completeness that vary by fuel type
- (Amiro et al., 2001; de Groot et al., 2007). However, de Groot et al. (2009) found that although these relationships may be relatively strong for experimental fires, they are clearly weaker for wildfires. A defining characteristic of fire emissions in the boreal forest is that the consumption of the ground layer (comprised of moss, lichens, litter, and organic soils) is larger than the consumption of aboveground biomass (McGuire
- et al., 2009; Boby et al., 2010; Kasischke and Hoy, 2012). Because of the seasonal thawing of the permafrost, the active layer becomes deeper and drier throughout the fire season and thus more prone to deeper burning (Lapina et al., 2008; Turetsky et al., 2011; Kasischke and Hoy, 2012). Based on this rationale, several authors have developed scenarios in which they assign ground fuel consumption values based on



the seasonality of the burn (Kajii et al., 2002; Kasischke and Bruhwiler, 2002; Soja et al., 2004). The dryness of the forest floor depends both on the time within the season and local drainage conditions (Kane et al., 2007; Turetsky et al., 2011). Kasischke and Hoy (2012) incorporated this expert knowledge to derive emissions from a set
of Alaskan fires by accounting for differential impacts of fire seasonality on several topographic classes.

A complementary approach to derive different levels of fuel consumption is to use direct post-fire remote sensing observations of severity (Kolden and Abatzoglou, 2012; Veraverbeke and Hook, 2013; Rogers et al., 2014). Severity is often referred to as fire or burn severity (Lentile et al., 2006; Keeley, 2009) and the difference between both terms lays within the temporal dimension of the post-fire environment. By definition fire severity measures the immediate impact of the fire, whereas burn severity incorporates both the immediate fire impact and subsequent recovery effects (Lentile et al., 2006; Veraverbeke et al., 2010). Spectral changes after a fire have shown to

- ¹⁵ be strongly related to field measurements of severity in a wide range of ecosystems (e.g. van Wagtendonk et al., 2004; Cocke et al., 2005; De Santis and Chuvieco, 2007; Veraverbeke and Hook, 2013), including the boreal forest (Epting et al., 2005; Allen and Sorbel, 2008; Hall et al., 2008; Soverel et al., 2010). In particular the differenced normalized burn ratio (dNBR) has become accepted as the standard spectral index to
- assess severity (López García and Caselles, 1991; Key and Benson, 2006). The dNBR is an index that combines near and short-wave infrared reflectance values obtained before and after a fire (Eidenshink et al., 2007). The spectral regions in dNBR are especially sensitive to the decrease of vegetation productivity and moisture content after the fire. Because of this, dNBR is a good indicator of aboveground biomass
 consumption, but may be less effective in estimating belowground consumption of boreal fires (French et al., 2008; Hoy et al., 2008; Kasischke et al., 2008). Other studies, however, have reported significant relationships between spectral indices, including dNBR, and belowground consumption measurements from field sites in boreal ecosystems (Hudak et al., 2007; Verbyla and Lord, 2008; Rogers)



et al., 2014). Rogers et al. (2014) also found a strong correlation between field measurements of aboveground and belowground consumption, which partly explained the observed relationship between dNBR and belowground consumption. Extrapolation of relationships like this needs further calibration with field data that span a wide range of fire seasonality and topographic conditions.

The day of burning within the fire season covaries with the depth of burning in the ground layer and thus temporal information on the time of burning within the season may aid prediction of belowground consumption (Turetsky et al., 2011; Kasischke and Hoy, 2012). Convolving burned area detection algorithms with active fire hotspots from the multiple overpasses per day from the Moderate Resolution Imaging Spectroradiometer (MODIS) allows for the development of daily burned area estimates (Parks, 2014; Veraverbeke et al., 2014). Daily burned area products using this approach have already been developed for regional studies in interior Alaska to develop daily fire emission estimates for a limited number of years (Kasischke and Hoy, 2012), and to infer burned area–climate relationships (Sedano and Randerson, 2014).

Daily burned area and emissions estimates may allow several improvements in studies focused on the composition and transport of fire aerosols, fire behavior, and fire modeling. Hyer et al. (2007) found that emissions from boreal fires averaged over

- ²⁰ 30 day intervals resulted in a reduction of 80 % of the variance compared to daily and weekly data in a fire aerosol transport simulation. The temporal resolution of emission data is especially important for boreal fires since they often reach most of their burned area in only a couple of days when the spatiotemporal patterns of ignitions and fire weather optimally coincide (Abatzoglou and Kolden, 2011; Sedano and Randerson,
- 25 2014). The representation of extreme fire weather periods and their influence and their influence on burned area in models may allow for more accurate predictions of interannual and decadal changes in the fire regime caused by climate warming (Jin et al., 2014). High resolution emission time series may also improve knowledge about differences in composition of aerosols and trace gases originating from flaming



and smoldering fires (Yokelson et al., 2013) as well as allowing for better prediction of human health impacts in downwind areas. More fundamentally, daily burned area estimates are critical for quantitatively examining landscape and weather controls on fire spread rates and severity.

In this paper we aim to develop a statistical model of carbon consumption by fire (kg C m⁻² burned area) in Alaska based on relationships between field measurements and environmental variables, including post-fire remote sensing observations of severity. Subsequently, we apply the derived model over the state of Alaska for the years 2001–2012 in synergy with a daily burned area product to derive state-wide daily carbon emissions from fire. The derived daily burned area and carbon emissions product is referred to as the Alaskan Fire Emissions Database (AKFED).

2 Spatiotemporal domain and data

2.1 Spatiotemporal domain

- The spatial domain covers the area between 58 and 71.5° N, and 141 and 168° W. This represents almost the entire mainland of Alaska with exclusion of the southern part of the Alaska Peninsula and Southeast Alaska, west of British Columbia (Fig. 1). The temporal domain of the study includes the years 2001–2012. Most Alaskan fires occur in the interior of the state, which consists of a mosaic of vegetation types (Fig. S1 in the Supplement). Black spruce forest dominates on cold, poorly drained, northoriented or lowland sites, whereas white spruce and deciduous species (mainly aspen and birch) prevail on warmer, better drained, south-oriented sites without permafrost (Viereck, 1973; Bonan, 1989). Grass- and shrubland ecosystems occur in early successional stands, poorly drained sites, steep slopes and at and above the treeline. The vegetation mosaic in interior Alaska is constantly reshaped by the occurrence of
- ²⁵ fire and subsequent post-fire succession. Fewer fires occur in the tundra regions in the north and at the western coastal areas of the state, however, the 2007 Anaktuvuk River



fire on the North Slope is the largest tundra fire on record (Jones et al., 2009; Mack et al., 2011; Kolden and Rogan, 2013), and the occurrence of tundra fires may increase with global warming (Hu et al., 2010; Rocha et al., 2012).

2.2 Field data

- ⁵ We assembled field data from three different publications (Boby et al., 2010; Turetsky et al., 2011; Rogers et al., 2014). Due to limited data availability for other land cover types than black spruce, we focused on black spruce plots and we retained all plots burned since 2000 for which cloud-free one year-post fire dNBR observations were available in the Monitoring Trends in Burn Severity (MTBS, Eidenshink et al., 2007)
 ¹⁰ database, resulting in a total of 126 plots (Table S2 in the Supplement). The location of the field plots is given in Fig. 1. Boby et al. (2010) and Rogers et al. (2014) provided a direct estimate of the belowground carbon consumption representing 39 plots. These plots, except for one of the Boby et al. (2010) plots, also included an estimate of aboveground consumption. Detailed description of the field sites and data acquisition
 ¹⁵ can be found in the respective publications. For the Turetsky et al. (2011) plots, belowground carbon consumption was calculated from depth of burn measurements
- using separate, region-wide mean soil-carbon accumulation curves for lowland, upland, and slopes with South (S), North (N), and East or West (EW) aspect (Fig. S2). We used a digital elevation model (Sect. 2.3.3) resampled to 500 m resolution for assigning the topographic classes to the field plots. Concave flat (slope ≤ 2%) areas were classified
- as lowland (L), convex flat areas were as upland (U). Sloped terrain was categorized as N aspect (aspect \geq 315 or < 45°), S aspect (aspect \geq 135 and < 225°), and E or W aspect (aspect \geq 45 and < 135°, or \geq 225 and < 315°).



2.3 Geospatial data

2.3.1 Alaska Large Fire Database (ALFD)

The Alaska Large Fire Database (ALFD) currently contains fire perimeters for the state of Alaska from 1940 till now (downloaded from http://afsmaps.blm.gov/imf/imf.jsp?site= firehistory, last accessed 25 November 2014). The database receives yearly updates. The reliability of the database increased over time as mapping technologies advanced, and since the 1980s consistent mapping with high quality and few omissions has been achieved (Kasischke et al., 2002, 2011). For our study, we extracted the fire perimeters of the years 2001–2012 from the ALFD (Fig. 1).

10 2.3.2 Active fire data

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The Global Monthly Fire Location Product (MCD14ML) contains geographic location and time for each fire pixel detected by MODIS on Terra (launched in December 1999) and Aqua (launched in May 2002). Additional information on brightness temperature, fire radiative power, scan angle and detection confidence is also provided. The product ¹⁵ is based on a contextual active fire algorithm that exploits the strong emission in the mid infrared region from fires (Giglio et al., 2003, 2006). We extracted the fire detections from all confidence levels for our domain for the months May–September, the months of the fire season, for all years. MODIS on Terra experienced an extended outage during our study period from 16 June through 2 July 2001 (Giglio et al., 2013).

20 2.3.3 Environmental variables

Ground layer consumption by fire depends on the amount of available dry fuels, which is determined by thickness, density and moisture content of the organic layer. After an extensive literature review, we selected a set of environmental variables to predict ground layer consumption over the landscape (Table 1). The selected environmental



variables were elevation, slope, northness (defined as the cosine of the aspect), tree cover, time of burning, and dNBR.

Topography is a good proxy of site conditions for several reasons. Elevation regulates organic layer thickness, carbon density, drainage and permafrost thaw by means of
its control on climate. At higher elevations the seasonal permafrost thaw starts later (Kasischke and Johnstone, 2005; Kasischke and Hoy, 2012). Uplands generally have shallower organic layers with a slightly higher carbon density than lowlands. (Kane et al., 2005, 2007; Turetsky et al., 2011). Uplands are generally also better drained than lowlands (Barrett et al., 2010; Kasischke and Hoy, 2012). Steep terrain is better drained
than flat land, but above a certain threshold steepness limits the establishment of trees resulting in shallower organic layers at steeper sites. (Hollingsworth et al., 2006). In

crown fire ecosystems, fire severity tends to increase with steepness when the wind direction aligns upslope (Rothermel, 1972; Pimont et al., 2012; Lecina-Diaz et al., 2014), and this may also affect ground layer consumption. N-oriented slopes are wetter

and colder than S-faced slopes and have thicker, less dense organic layers (Kane et al., 2007; Turetsky et al., 2011). Here we derived elevation, slope and northness from the Advanced Spaceborne Thermal Emission and Reflection Radiometer Global Digital Elevation Model Version 2 (ASTER GDEM 2, Tachikawa et al., 2011). The ASTER GDEM 2 is a 30 m elevation model extracted from ASTER stereo-pair images.

Pre-fire tree cover is closely related to to site productivity and stand age, and thus influences organic layer thickness, density and moisture content. (Kasischke and Johnstone, 2005; Beck et al., 2011b; Rogers et al., 2013). Tree cover generally increases with stand age and better drainage conditions (Beck et al., 2011b; Rogers et al., 2013). Tree cover also is directly related to the amount of biomass available for aboveground consumption, which has been shown to correlate reasonably well with belowground consumption within a single fire (Rogers et al., 2014). For the comparison with the field plots, we used 30 m tree cover data from the Landsat-based tree cover continuous field product for the year 2000 (Sexton et al., 2013). For the



statewide extrapolation, tree cover was downloaded from the annual Terra MODIS

Vegetation Continuous Fields Collection 5 product at 250 m resolution for the years 2000–2010 (MOD44B, Hansen et al., 2003). The generation of the MOD44B product was discontinued after 2010, and we therefore used the tree cover layer of the year 2010 as pre-fire tree cover for the year 2012.

The time of burning in the season covaries with the mean depth of the active layer, and thus may be related to the amount of dry ground fuels available for burning (Turetsky et al., 2011; Kasischke and Hoy, 2012). We assigned the time of burning for each pixel based on the MODIS active fire observations. We found that the nearest neighbor variant of the inverse distance weighting technique, in which the pixel is assigned the value of the closest active fire detection excluding scan angles larger than 40°, performed best and with a within-one-day accuracy for most pixels (Fig. S3). The resulting progression maps were binned with a daily time step, at local solar time. We investigated the dNBR as an explanatory variable in our fuel consumption model

because extensive literature suggests that it may have some predictive power in
 boreal forest ecosystems. For the comparison with the field plots, 30 m dNBR data
 was retrieved from the Landsat-based Monitoring Trends in Burn Severity database
 (MTBS, Eidenshink et al., 2007). For the statewide extrapolation, we calculated the

NBR from MODIS surface reflectance data in the near infrared (NIR, centered at

- 858 nm) and short-wave infrared (SWIR, centered at 2130 nm) bands: NBR = (NIR SWIR)/(NIR + SWIR). We used the surface reflectance data contained in the 16 day
- Terra MODIS Vegetation Indices 16 day Collection 5 product at 500 m resolution for the years 2000–2013 (MOD13A1, Huete et al., 2002). To account for cloudy observations in single MODIS composites, we created summer NBR composites using the five 16 day composites between days of the year 177 and 256. We only used
- good data as indicated by the MOD13A1 quality flags. NBR values were calculated as the mean of all available good observations within the five composites. The dNBR was calculated using the one-year pre-and post-fire NBR layers. Within the MTBS database, we also only considered one-year post-fire dNBR information. This



minimized potential differences in the interpretation of dNBR values from different postfire years (Veraverbeke et al., 2010).

The above six variables (elevation, slope, northness, tree cover, time of burning and dNBR) were targeted to develop the belowground consumption model. dNBR and tree cover were used as predictors of aboveground consumption.

2.3.4 Land cover data

We used the Fuel Characteristic Classification System (FCCS, Ottmar et al., 2007; Riccardi et al., 2007) layer of the year 2001 at 30 m to represent land cover in the study area (downloaded from http://landfire.cr.usgs.gov/viewer/, last accessed
¹⁰ 25 November 2014). The FCCS is a national effort by the U.S. Forest Service to provide a fuel type classification that is compiled from literature, inventories, photo series and expert opinion (Ottmar et al., 2007; Riccardi et al., 2007). For Alaska, the layer is available for the years 2001 and 2008. Since less than one percent of reburning occurred during the period of our study (2001–2012), we decided to only use the 2001
¹⁵ layer. We aggregated the fuel types into five land cover classes: black spruce, white spruce, deciduous, tundra-grass-shrub and non-vegetated (Table S1, Fig. S1).

3 Methods

AKFED provides daily burned area and carbon emissions for the state of Alaska between 2001 and 2012 at 500 m resolution. Since unburned islands are not fully accounted for within perimeters (Kasischke and Hoy, 2012; Kolden et al., 2012; Rogers et al., 2014; Sedano and Randerson, 2014), and some small fires are not accounted for outside the perimeters (Randerson et al., 2012), we developed a burned area mapping approach that screened dNBR values within the ALFD perimeters and in the vicinity of active fire pixels outside the perimeters.



The carbon consumption model was formulated for black spruce based on the relationship between the observed carbon consumption at the field locations and the environmental variables. We extracted the pixel values of elevation, slope, northness, pre-fire tree cover and dNBR at 30 m at the location of the field plots. The time of ⁵ burning was assigned from the nearest active fire observation.

To extrapolate the model in space and time, we used a spatial resolution of approximately 500 m, the native resolution of the MOD13A1 product used to derive the dNBR layers. The exact spatial resolution of AKFED is 450 m, which is the multiple of 30 m closest to the exact 463 m native resolution of the MOD13A1 product. This spatial resolution facilitated spatial averaging of 30 m DEM, tree cover and land cover data, and is referred to in this manuscript as 500 m. The decision to extrapolate the model at this resolution was driven by data availability. We aimed at complete spatial

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coverage. Even with current efforts such as MTBS and the Web-Enabled Landsat Data (WELD, Roy et al., 2010), initial exploration of these datasets indicated that complete
 Landsat dNBR coverage for every burned pixel was still partly constrained by clouds, smoke, snow and gaps due to the Landsat 7 scan line corrector failure.

Elevation, aspect and northness were spatially averaged from the native 30 m resolution of the ASTER GDEM 2 to the 500 m resolution. Similarly, the MOD44B tree cover product was spatially averaged from its native 250 m resolution to the 500 m

- resolution. The time of burning was also obtained at this grid resolution. To account for other land cover types than black spruce, the aggregated FCCS product was rescaled to 500 m in a way that every pixel at 500 m contained the percentage of black spruce, white spruce, deciduous, tundra-grass-shrub and non-vegetated land (Fig. S1). All analyses were performed within the Albers equal area projection for Alaska (central
- ²⁵ meridian: 154° W, standard parallel 1: 55° N, standard parallel 2: 65° N, latitude of origin: 50° N) with North American Datum 1983 (NAD83). An overview of the workflow is given in Fig. 2.

We compared the yearly burned area and carbon emissions estimates derived from AKFED with those from GFED3s (Randerson et al., 2012), and therefore spatially



averaged the AKFED estimates over the 0.25 $^{\circ}$ grid from GFED3s for the period 2001–2010.

3.1 Daily burned area mapping (500 m)

Annual burned area maps at 500 m were derived by applying a threshold on the dNBR values of pixels within the perimeters of the ALFD and outside the perimeters but within 1 km of an active fire pixel. If a fraction of the 500 m pixel was covered by a fire perimeter polygon, then this pixel was considered in the perimeter. Pixels with a dNBR value larger than 0.15 were classified as burned. The dNBR threshold was determined based on the dNBR variability that exists within unburned pseudo-invariant pixels (100% barren pixels at 500m, FCCS code 931). We found that, as expected, 10 that the mean dNBR was close to zero (0.01). The SD of the distribution equaled 0.15. By selecting this value as our threshold for the burned area mapping, we aimed to minimize commission errors, however, we recognize that this may have incurred a small omission error for pixels that were only partially burned and/or burned with low severity. The threshold value we used was the same as the one applied by Sedano 15 and Randerson (2014) who used a similar burned area mapping approach within ALFD perimeters. All burned pixels were assigned a time of burning from the detection time of the nearest active fire observation excluding observations with view zenith angles

higher than 40° (Fig. S3).

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20 3.2 Carbon consumption model (30 m)

We aimed to separately predict below- and aboveground carbon consumption based on the relationships between field plot data and environmental variables at 30 m (elevation, slope, northness, pre-fire tree cover, time of burning and dNBR). We investigated two modeling techniques to formulate a carbon prediction model. The first technique was multiplicative nonlinear regression. This technique is based directly on the interpretation of empirical relationships that may exist between the environmental



variables and the carbon consumption field data. This technique has demonstrated to be effective in a similar application to predict fire occurrence and size in Southern California (Jin et al., 2014). The second technique was gradient boosting of regression trees. We applied this technique because previous work indicated that it may have ⁵ value in predicting depth of burning in boreal black spruce forest (Barrett et al., 2010, 2011). Gradient boosting is a machine learning technique, which produces a prediction model in the form of an ensemble of, in our case, multiple regression trees. We parameterized the gradient boosting of regression trees with the requirement that

- each leaf of at least 10% of the data per leaf and using 50 weak learner trees. Both multiplicative nonlinear regression and gradient boosting of regression trees allow 10 for nonlinear relationships between the dependent and independent variables, and interactions between the independent variables. We opted to use the multiplicative nonlinear regression model for our study because we found it to outperform the gradient boosting model in predicting observations that were not used to train the models (Fig. S4).
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3.3 Daily carbon emissions, 2001–2012 (500 m)

Once the carbon consumption prediction model was optimized, we extrapolated this model over the spatiotemporal domain of the study at 500 m resolution. To do so, we first guantified the linear relationship between Landsat and MODIS dNBR and tree cover (Fig. S5). The resulting regression equations were applied to the MODIS-derived dNBR and tree cover layers to allow direct application in the carbon consumption model that was optimized using Landsat data.

Due to data paucity of carbon consumption in other land cover types than black spruce, we were only able to formulate a prediction model for black spruce forest.

Deciduous and white spruce stands generally have higher aboveground and lower 25 belowground fuel loads (Kasischke and Hoy, 2012; Rogers et al., 2014). We assumed that fuel consumption in these land cover types was controlled by the same variables as from the black spruce consumption model. However, we multiplied the estimates



derived from our black spruce-based equations by consumption ratios for aboveand belowground deciduous and white spruce stands that were developed using the Consume 3.0 fuel consumption model (Ottmar et al., 2006) (Fig. S6). Consumption estimates derived from the black spruce model were multiplied with the consumption

- ⁵ ratios proportional to the land cover fractions within each pixel. For the state wide extrapolation over pixels classified as tundra-grass-shrub and non-vegetated, we used the model derived for black spruce. The mean tree cover value of all 30 m tundra-grassshrub and non-vegetated pixels within the ALFD perimeters between 2001 and 2012 was 11 % (SD = 14 %).
- ¹⁰ A pixel-based uncertainty was included in the extrapolation, here reported as the SD of the prediction error. The uncertainty in the total consumption was calculated from the uncertainties in the below- and aboveground consumption predictions:

uncertainty_{total} =
$$\sqrt{\text{uncertainty}_{\text{belowground}}^2 + \text{uncertainty}_{\text{aboveground}}^2}$$
.

4 Results

4.1 Carbon consumption model

The relationship between the depth of burn and the individual environmental variables was strongest for dNBR and tree cover ($R_{adjusted}^2 = 0.25$, p < 0.001), with both relationships modeled using exponential response functions (Fig. S7). Depth of burn responded with a relatively strong Gaussian relationship to elevation ($R_{adjusted}^2 = 0.24$, p < 0.001), with the deepest burning occurring in the elevation range between 300 and 600 m. A weak relationship was found for slope ($R_{adjusted}^2 = 0.05$, p < 0.05), and no relationship was observed for time of burning or northness. The strongest individual predictors for belowground consumption were time of burning ($R_{adjusted}^2 = 0.09$, p < 0.001), slope ($R_{adjusted}^2 = 0.08$, p < 0.01) and dNBR ($R_{adjusted}^2 = 0.06$, p < 0.01). ²⁵ Northness ($R_{adjusted}^2 = 0.04$, p < 0.05), elevation ($R_{adjusted}^2 = 0.04$, p < 0.05), and tree 17594



(1)

cover ($R_{adjusted}^2 = 0.03$, p < 0.05) had a weaker influence (Fig. S8). The aboveground carbon consumption demonstrated stronger, also exponential, relationships with prefire tree cover ($R_{adjusted}^2 = 0.53$, p < 0.001) and dNBR ($R_{adjusted}^2 = 0.23$, p < 0.001) variables (Fig. S9).

The optimized multiplicative nonlinear model for the depth of burn and belowground carbon consumption in black spruce forest based on the field and 30 m data was formulated as:

$$depth_{30 m} or C_{belowground, 30 m} = c_1 + c_2 \cdot e^{c_3 \cdot dNBR} \cdot e^{c_4 \cdot tc} e^{c_5 \cdot DOY} \cdot e^{\frac{-(elev - c_6)^2}{2 \cdot c_7}}$$

where $c_{1,...,7}$ are the optimized coefficients, dNBR is the differenced normalized burn ratio, tc is the pre-fire tree cover, DOY is the day of the year and elev is the elevation. This model yielded a $R_{adjusted}^2$ of 0.40 and a root mean square error (RMSE) of 5.44 cm (p < 0.001) for depth of burn (Fig. 3a) and a $R_{adjusted}^2$ of 0.29 and a RMSE of $1.99 \text{ kg} \text{ Cm}^{-2}$ (p < 0.001) for belowground carbon consumption (Fig. 3c). Inclusion of slope and northness did not improve the performance of the models.

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The aboveground carbon consumption in black spruce forest was modeled as a multiplicative exponential model of the 30 m dNBR and pre-fire tree cover variables:

 $C_{\text{aboveground,30 m}} = c_1 + c_2 \cdot e^{c_3 \cdot \text{dNBR}} \cdot e^{C_4 \cdot \text{tc}}$

where $c_{1,...,4}$ are the optimized coefficients. This model had a $R_{adjusted}^2$ of 0.53 and a RMSE of $0.26 \text{ kg} \text{ Cm}^{-2}$ (p < 0.001) (Fig. 3e), the same as the individual relationship between the aboveground carbon consumption and pre-fire tree cover. In the mostly 20 stand-replacing fires that occur in black spruce forest, dNBR did not contribute to the aboveground model. However, we decided to include dNBR to allow for variations in the severity of burning in other land cover types than black spruce, for example in deciduous stands which often burn less severe and frequent (Viereck, 1973; Cumming, 2001). We found no trends in the residuals of any of the models (Fig. 3b, d and f). The 25

Discussion Pape **Daily burned area** and carbon emissions from Discussion Paper boreal fires in Alaska (2)S. Veraverbeke et al. **Title Page** Introduction Abstract **Discussion** Paper Conclusions References Tables **Figures** (3)Close Back **Discussion** Paper Full Screen / Esc **Printer-friendly Version** Interactive Discussion

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environmental variables used in the carbon consumption models are plotted for the spatiotemporal domain in Fig. S10.

Using the consumption ratios derived for white spruce and deciduous cover (Fig. S6), below- and aboveground carbon consumption at 500 m resolution were calculated as:

$$C_{\text{belowground,500 m}} = (\text{fr}_{\text{bs}} + 0.66 \cdot \text{fr}_{\text{ws}} + 0.31 \cdot \text{fr}_{\text{dec}}) \cdot C_{\text{belowground,30 m}}$$

$$C_{\text{aboveground,500 m}} = (\text{fr}_{\text{bs}} + 1.56 \cdot \text{fr}_{\text{ws}} + 1.75 \cdot \text{fr}_{\text{dec}}) \cdot C_{\text{aboveground,30 m}}$$

$$(4)$$

$$(5)$$

where fr_{bs} is the fraction of black spruce within the 500 m pixel, fr_{ws} is the fraction of white spruce and fr_{dec} is the fraction of deciduous. The black spruce model was also applied to residual tundra-grass-shrub and non-vegetated parts of each pixel. Before application in Eqs. (4) and (5), the dNBR and tree cover from MODIS were first converted into their Landsat equivalent values using the equations from Fig. S5. The total carbon consumption was calculated as the sum of the below- and aboveground consumption.

To assess the importance of the individual variables in the model we compared ¹⁵ all models inputting two or more variables for the depth of burn and belowground consumption models (Fig. 4). For the depth of burn model, elevation was the most important explanatory variable. 2-variables models combining elevation with time of burning and dNBR for example performed better than the 3-variables models that excluded elevation. For the belowground consumption model, the time of burning variable was crucial. All 2-variables models combining time of burning with any of the other variables performed better than the 3-variables models that excluded time of burning.

4.2 Daily burned area and carbon emissions, 2001–2012

The total carbon emissions and uncertainty for the spatiotemporal domain are shown

²⁵ in Figs. 5 and 6. Daily carbon emissions over the entire spatial domain were primarily driven by daily burned area (Fig. 6), however, there was considerably spatial variability



in carbon consumption (Fig. 5a). Annual burned area in Alaska from AKFED ranged between 35 and 2268 kha per year between the years 2001 and 2012, resulting in a carbon emission range between 1 and 67 Tg year⁻¹ (Fig. 7a). 1% of the burned area was mapped from active fire detections outside the perimeters, whereas 18% of the pixels within the ALFD perimeters were mapped as unburned after dNBR screening (to remove edge effects due to low resolution bias burned pixels within 900 m of the perimeter edge were considered in the perimeter, and unburned pixels in the perimeter within 900 m of the perimeter edge were considered outside the perimeter in these statistics). 2004 (2268 kha), 2005 (1642 kha), 2009 (1030 kha) and 2002 (723 kha) were the largest fire years. More than 50% of the burned area between 2001 and 10 2012 burned in 53 single days (Fig. 6). Most of the burning occurred in July and August (Fig. 7b). The seasonal pattern of carbon emissions generally followed the same seasonal pattern as the burned area. However, carbon consumption increased by a small amount later in the season. Mean annual carbon consumption increased slightly with total annual burned area (Fig. 8) and mean carbon consumption per fire 15 and fire size were positively correlated (r = 0.22, p < 0.001). Most of the burned area occurred in black spruce and white spruce ecosystems (61%), followed by tundragrass-shrub ecosystems (23%), deciduous forests (14%), and non-vegetated areas

The median total carbon consumption in AKFED for all burned pixels between 2001 and 2012 was 2.48 kg m⁻², with the majority originating from belowground losses. The median aboveground consumption was 0.14 kg Cm^{-2} (median = 2.3 kg Cm^{-2}) (Fig. 9a). The median pixel-based uncertainty of the total consumption was 0.38 kg Cm^{-2} and originated largely from the uncertainty in belowground consumption (median = 0.35 kg Cm^{-2}) (Fig. 9b). The median uncertainty in aboveground carbon consumption was 0.11 kg Cm^{-2} . The median belowground consumption was higher in burned pixels that had a vegetation cover of 100% black spruce (median = 2.63 kg Cm^{-2}) (Fig. 10). However, belowground consumption estimates for black spruce were lower than those observed in the field data median = 3.1 kg Cm^{-2}). The

(2%), with an overall mean tree cover of 32%.



field data also had a higher median elevation, tree cover and dNBR compared to the distribution of the burned area between 2001 and 2012 (Fig. S11).

Yearly burned area and carbon emissions from AKFED and GFED3s were similar in all years except 2001 (Table S3) when MODIS experienced an outage (Giglio et al.,

- ⁵ 2013). A strong linear relationship existed between the annual spatial distribution of AKFED and GFED3s 0.25° burned area between 2001 and 2010 (r = 0.95, intercept = 0.04 and slope = 0.98). The similarity between the burned area from AKFED and GFED3s is not surprising since they operate with similar algorithms. Both algorithms look at changes in a spectral index derived from MODIS surface reflectance imagery
- within a spatial context that is known to have high burned area probability. GFED3s therefore exclusively relies on the MODIS active fire data, while AKFED made combined use of the perimeters of the ALFD and MODIS active fire data outside the perimeters. Carbon emissions were also closely related between the two products (r = 0.87, intercept = 0 and slope = 0.79), as well as mean consumption (2.48 kg C m⁻²)
- ¹⁵ for AKFED vs. 2.57 kg Cm⁻² for GFED3s). Statistics of individual years generally followed these trends, except for the year 2001. AKFED and GFED3s exhibited no significant spatial correlation for carbon consumption (r = -0.04, p = 0.13). The difference between the consumption estimates from GFED3s and AKFED is likely a consequence of the different parametrization of belowground consumption in the two products. GFED3s estimates organic layer consumption in boreal soils directly as
- 20 two products. GFED3s estimates organic layer consumption in boreal soils directly as a function of soil moisture (van der Werf et al., 2010). AKFED uses elevation and time of burning to estimate soil moisture, and is complemented with pre-fire tree cover and post-fire dNBR data. The difference in spatial resolution between AKFED (500 m) and GFED3s (0.25°) may also explain part of the discrepancy.



5 Discussion

5.1 Burned area

Our results corroborate previous work highlighting the importance of unburned islands, which amounted to 18% of the fire perimeters. This value is close to the estimates of

14, 15 and 17% reported by Kolden et al. (2012), Sedano and Randerson (2014) and 5 Rogers et al. (2014) for fires in interior Alaska. Burned area and emissions peaked in July and August and extended later in the fire season than previously reported for burning before the 2000s (Kasischke et al., 2002) (Fig. 7b). This tendency towards lateseason burning was found for the four largest fire years that occurred during the study period (2002, 2004, 2005 and 2009) (Fig. 6) and can be attributed to the late-season occurrence of weather conditions favorable to fire spread during these large fire years (Hu et al., 2010; Sedano and Randerson, 2014). Some of these late-season fires have occurred outside the fire-sensitive region of interior Alaska which is dominated by black spruce forest. The large Anaktuvuk River fire of 2007 on the North Slope is a wellknown example of this (Hu et al., 2010; Mack et al., 2011; Rocha et al., 2012). This 15 suggests that late-season fires may increasingly occur in sparsely vegetated areas like tundra that, despite low aboveground biomass, contain large belowground carbon stocks.

5.2 Environmental variables controlling carbon consumption

20 5.2.1 Elevation

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Elevation was the most important variable in the depth of burn model and contributed to the skill of the belowground carbon consumption model (Fig. 4). Barrett et al. (2010, 2011) and Turetsky et al. (2011) demonstrated the explanatory power of topographic variables for belowground consumption in black spruce forests. Kasischke and Hoy (2012) elaborated on their findings and assigned curves of carbon consumption that



linearly increased with time of burning throughout the fire season to three different topographic classes. The predictive power of elevation for belowground consumption is likely explained by two effects. Elevation captures the spatial distribution of cold temperatures that limit the development of soils and black spruce establishment at the higher elevations (Fig. S7a). Elevation also captures the fuel moisture controls resulting in generally wetter fuels at lower elevations. This moisture control is dynamic through the fire season and this is captured by interactions between elevation and time of burning (Fig. 4). Inclusion of the northness and slope variables did not improve our model prediction. This contrasts with the findings of Barrett et al. (2010, 2011)
who ranked slope and aspect, and derived drainage indicators, in the top predictors for depth of burn. It contrasts with Turetsky et al. (2011) who found differences in average consumption among different aspect classes. As an individual variable, slope did display some explanatory power (Figs. S7b and S8b), but did not contribute to the final model. The contrasting findings of this study compared to Barrett et al. (2010, 2010)

- ¹⁵ 2011) and Turetsky et al. (2011) can partly be explained by the scale-dependency of controls on carbon consumption. The model in this study was developed for regional state-wide emission predictions. At this scale, the topographic variable explaining most of the variability in belowground fuel consumption (as a proxy of drainage condition and soil organic layer thickness) was elevation. At a more local scale, for example within one
- fire, differences in elevation may be smaller, and the variability in drainage conditions and hence belowground fuel consumption may be better captured by including slope and aspect variables. Hollingsworth et al. (2006) found a similar scale-dependency explaining the occurrence and abundance of black spruce types from local to regional scales. Further improvements of the model could include fine scale drainage effects driven by slope and aspect superimposed on the elevation control on consumption.

5.2.2 Time of burning

Time of burning within the fire season was the most important variable in the belowground carbon consumption model (Fig. 4). Time of burning is used as a proxy



for seasonal drying of the soil organic layer. This, however, also depends on elevation as lower elevations will dry earlier than higher elevations. Because of typically drier conditions of the belowground fuel later in the season, late-season fires tend to have higher consumption rates (Figs. 6 and 7b, S8d). We found the magnitude of this
seasonal change to be smaller than previously reported by Turetsky et al. (2011) and implemented by Kasischke and Hoy (2012). A potential improvement we tested is the replacement of the time of burning with actual fire weather indices that quantify the drying of the ground layer like the drought and duff moisture codes (Van Wagner, 1987). We calculated these indices from daily air temperature, relative humidity, precipitation and wind speeds at 32 km resolution from the North American Regional Reanalyis (NARR, Mesinger et al. (2006), downloaded from http://www.esrl.noaa.gov/psd/data/gridded/data.narr.html, last access: 25 November 2014) and assigned the daily value, as determined by time of burning estimates, of the weather indices from the nearest grid cell to the field plot locations. At the spatial resolution used in our

- ¹⁵ analysis the weather indices did not improve the carbon consumption model. Although strongly related during experimental fires, the relationship between fire weather and belowground fuel consumption has generally been weak for wildfires (de Groot et al., 2009). This may potentially be the result of the common approach of assigning fire weather index values from the nearest ground weather station, which may not fully
- 20 represent the actual weather conditions at the place of burning, or the use of gridded meteorological data at a coarser spatial resolution. Weather data with higher spatial resolution, for example derived from downscaling coarse resolution data like NARR, may provide a valuable pathway to re-assess the relationships between field-observed consumption of the organic layer in wildfires and fire weather conditions (Abatzoglou
- and Kolden, 2011). Weather indices may also capture the climatological and fuel moisture controls that are currently imposed in the model by elevation (Sect. 5.2.1). Replacement of time of burning and elevation with weather indices in future work may enable broadening the geographical scope of the current regional carbon consumption model to other regions of boreal North America.



5.2.3 Burn severity (dNBR)

The utility of dNBR in the boreal region has been subject to much debate (French et al., 2008). We found a relatively strong relationship between dNBR and aboveground consumption (Fig. S9a). The criticism on the use of dNBR, however, has focused on its ability to predict belowground consumption (French et al., 2008; Hoy et al., 2008). Some authors have found relatively strong correlations between field measures of belowground consumption and dNBR (Hudak et al., 2007; Verbyla and Lord, 2008; Rogers et al., 2014). Barrett et al. (2011) also ranked dNBR in the top third predictors of a set of 35 spectral and non-spectral environmental variables. French et al. (2008)

- ¹⁰ concluded that the use satellite-based assessments of burn severity, including dNBR, in the boreal region "need to be used judiciously and assessed for appropriateness based on the users' needs". We found here that, as an individual variable, dNBR was the top predictor of depth of burn in black spruce forests together with pre-fire tree cover (Fig. S7). We also found that including dNBR in the model resulted in
- additional explained variance compared to models that excluded dNBR (Fig. 4). In addition, dNBR and tree cover were found to vary at a finer spatial scale than elevation or time of burning (Fig. S12), and their inclusion in the model as such likely improved the representation of the spatial heterogeneity in fuel consumption.

The relatively high correlations between depth of burn, and dNBR and tree cover suggest that crown fires in high density black spruce plots also cause deeper burning into the ground layer. Burning into the ground layer is primarily controlled by fuel moisture in the ground layer, which was modeled here as a function of elevation and time of burning. For a certain moisture condition of the ground layer, determined by elevation and fire seasonality, dNBR thus adds complementary power for the prediction

of the consumption of the ground layer. This may explain why the dNBR has performed well in studies that focused on one single fire in which the elevation and time of burning were relatively constant (Hudak et al., 2007; Verbyla and Lord, 2008; Rogers et al., 2014). We conclude that the dNBR can well be used as a stand-alone indicator of



belowground consumption in boreal ecosystems when drainage and fire seasonality remain relatively constant. When used over large areas and over a range of burn conditions, the synergistic use of the dNBR with other environmental variables is essential for improving model performance. This finding agrees with Barrett et al. (2010,

⁵ 2011), who found that a combination of spectral and non-spectral data optimized depth of burn prediction in black spruce forest.

5.2.4 Tree cover

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This is the first study to evaluate the potential of tree cover as a predictor of fuel consumption in black spruce forests. The relatively strong relationship with aboveground consumption is intuitive as black spruce forest mostly experience stand-replacing crown fires and tree cover is directly related to the amount of available biomass and the probability that the crown fire can spread from tree to tree. Its utility for predicting belowground consumption is less obvious and more indirect. We included tree cover in our analysis since we hypothesized that it would be a good proxy of stand

¹⁵ age and site productivity which directly relates to drainage conditions, and thickness and density of the organic layer (Kasischke and Johnstone, 2005; Beck et al., 2011b; Rogers et al., 2013). High intensity crown fires in dense black spruce plots also may provide more radiant heating (and drying) of the ground layer, enabling deeper burns. Both dNBR and tree cover relationships with depth of burn observations were consistent with this effect.

5.3 Representativeness of the field data and uncertainties

The state wide median belowground consumption estimate of 2.3 kg Cm^{-2} of this study is lower than the median values of the field data used in this study (Fig. 10). Most available field data focus on black spruce ecosystems. While this is the ecosystem most affected by fire in Alaska, other land cover types like white spruce, deciduous, shrub or grass cover also burn (Kolden and Abatzoglou, 2012). We estimated a conifer



fraction burned of 61 % (39 % black spruce and 22 % white spruce), a tundra-grassshrub fraction of 23%, and a deciduous fraction of 14%. Although deciduous and white spruce trees may have higher amounts of available aboveground fuel (Kasischke and Hoy, 2012; Rogers et al., 2014), it is well known that the belowground consumption of fuels dominates in boreal ecosystems and that the largest fire-affected belowground carbon stocks are stored in black spruce ecosystems (McGuire et al., 2009; Kasischke and Hoy, 2012). The burning in other land cover types than black spruce thus partly explains the relatively low state-wide median belowground consumption estimate of 2.3 kg Cm⁻². We here used the FCCS classification to characterize land cover. We found this layer indicative for the vegetation patterns in the region. However, we 10 found that 60 out of the 126 black spruce plots from the field dataset (Sect. 2.2) were misclassified as white spruce (29), tundra-grass-shrub (14), deciduous (12) and non-vegetated (5). We also found that of eight white spruce-aspen plots from Rogers et al. (2014), seven were classified as black spruce and one as shrubgrass-tundra. The sample size of the land cover ground truth data available was too 15 small to robustly quantify potential over- or underrepresentation of land cover types in the FCCS layer. An improved land cover characterization, including quantitative uncertainty estimates, is thus essential for reducing region-wide uncertainties. An underrepresentation of the black spruce cover in the FCCS would result in lower state-

- ²⁰ wide average consumption estimates and vice versa. While some land cover types, such as spruce and deciduous cover, are likely well separable from remote sensing based on spectral and phenological characteristics, more detailed distinctions, for example between black and white spruce, may be more challenging but not impossible with the combined use of structural and optical attributes (e.g. Goetz et al., 2010).
- ²⁵ The median belowground consumption estimate increased to 2.63 kg C m⁻² in pure black spruce stands (Fig. 10). This value was still clearly lower than the median of 3.1 kg C m⁻² from the field data used in this study. The field data were, relative to the state wide distribution, disproportionally sampled in mid-elevation areas with high tree cover and high dNBR (Fig. 11s). This suggests that currently available field



measurements of carbon consumption in black spruce forest are biased towards areas that tend to have higher levels of fuel consumption. Part of this bias can also be explained by the fact that most of the field plots used in this study were sampled in fires from the large and severe fire year 2004 (85 out of the 126 plots). We found that large fires years generally have higher consumption estimates (Fig. 8). This corroborates findings of Turetsky et al. (2011) and Kasischke and Hoy (2012), although the increase of consumption with higher annual burned area was less than reported in these studies. We also found support for the finding of Duffy et al. (2007) and Beck et al. (2011a) that fire size and consumption level are positively correlated. In our study the belowground

fraction of the total carbon consumption was higher than those from the field plots used.

Our model provided a per-pixel uncertainty estimate, expressed as the SD from the prediction error (Fig. 9b). Median uncertainty was 0.38 kgCm⁻², mostly originating from the belowground fraction (median 0.35 kgCm⁻²) and equaling approximately 15% of the modeled consumption values on a per-pixel basis. These values are slightly lower than the per-pixel uncertainties from model prediction estimated by Rogers et al. (2014), who demonstrated that these per-pixel uncertainties largely average out when scaled over larger areas. Most other studies that have modeled landscape level carbon emissions from Alaskan fires have relied on scenarios in which uncertainties of

different variables were assigned based on expert knowledge (French, 2004; Kasischke and Hoy, 2012). These studies found uncertainties in the range of 5 to 30 %, expressed as the coefficient of variation (SD/mean).

5.4 Future directions

Databases of burned area, severity and emissions with high spatial and temporal resolution are a necessity to advance several related fields. They allow for a more thorough evaluation of models used to relate weather, fuels and topography to fire spread rates that were originally derived using laboratory measurements (Rothermel, 1972; Beer, 1991). Spatially explicit burned area data with high temporal resolution will also allow for quantitative assessments of constraints on fire progression due to fuel



discontinuities and fire weather conditions (Krawchuk et al., 2006). Knowledge of these constraints on fire spread may prove valuable when predicting future fire regimes that will result from changes in fire weather, vegetation dynamics and inherent landscape heterogeneity.

- At least two aspects of the boreal fire regime in Alaska have already changed over the last half century (Kasischke et al., 2010). There has been an increase in the burned area, driven by a few excessively large fire years, and a shift towards more severe late-season burning. Both changes, however, also provoke additional increases in the surface albedo through deciduous dominance in the early stages of succession and this offect laste for many decedes (Panderon et al. 2006; Johnstone et al. 2010;
- this effect lasts for many decades (Randerson et al., 2006; Johnstone et al., 2010; Barrett et al., 2011; Beck et al., 2011a). Jin et al. (2012) showed that the observed albedo increase scaled with the severity of burning, while Randerson et al. (2006) demonstrated that boreal forest fires can have a slight cooling effect on the climate. The net effect of boreal fires on radiative forcing will have to come from an analysis that the observed is the severe of the severe of
- integrates the warming, mainly from greenhouse gas emissions, and cooling, mainly from albedo increases, effects over multiple decades with consideration for the spatial heterogeneity in fuels and their consumption.

Fire emission databases with high temporal and spatial resolution also may enable improvements in our understanding of fire aerosol composition and decrease the large uncertainties that currently exist in fire emission factors of different gas species by

- comparing them to in situ measurements. Fire emission actions of different gas species by transport models therefore need to be convolved at high spatial and temporal resolution. In addition, fires in the boreal region are episodic and most of the burned area and emissions occur in relatively short amount of time. High resolution time series
- ²⁵ are therefore of paramount importance as input to transport models and to infer and forecast possible health effects in populated areas exposed to smoke plumes (Hyer et al., 2007).



Conclusions 6

By integrating field and remote sensing variables at multiple scales, we developed a database of burned area and carbon emission by fire at 500 m spatial resolution and daily temporal resolution for the state of Alaska between 2001 and 2012. We found that, although most of the fires burned in black spruce forest, considerable area burned white spruce forest, deciduous forest, and grass- and shrubland that displayed lower consumption. This partly explained why the median consumption of 2.48 kg Cm⁻² over the spatiotemporal domain was lower than that typically observed in black spruce plots. However, even when only pure black spruce pixels were considered the median consumption was still lower than most field observations. This suggests

- 10 that the currently available field data are biased toward high consumption sites and we recommend caution in extrapolating these values over large areas without taking into account spatial heterogeneity in fuels and consumption over the landscape. Further improvements in land cover characterization are required to remove potential biases
- that may originate from uncertainties in this layer.

Our carbon consumption model was driven by four environmental variables: elevation, time of burning, differenced normalized burn ratio (dNBR) and pre-fire tree cover. At the regional level of our study, elevation controlled fuel moisture conditions and soil organic layer thickness. Time of burning within the fire season, in combination with elevation, was used to estimate the seasonal thawing of the permafrost and

- subsequent availability of dry organic soil. dNBR and tree cover explained additional model variance, and superimposed spatial detail on the consumption estimates that was unavailable from the other layers. While the use of dNBR as stand-alone indicator of belowground consumption by fire in boreal ecosystems may have limitations, our
- model predictions benefited from its synergy with other environmental variables. Higher 25 dNBR values showed a clear trend towards deeper burning. Tree cover was found to be an excellent proxy of stand age and site productivity and related soil organic layer thickness and density. The observed relationships between belowground consumption,



and dNBR and tree cover, suggest a mechanism where high density black spruce plots provide an easier heat transfer of fire from the canopy to the ground layer, and vice versa, resulting in higher consumptions.

The Alaskan Fire Emissions Database (AKFED) is available from sites.google.com/ ⁵ a/uci.edu/sander-veraverbeke/akfed and will be updated regularly. This data could further contribute to the knowledge on spatiotemporal patterns, controls and limits on fire progression, air pollution and exposure, and aerosol composition of boreal fires.

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BGD 11, 17579–17629, 2014 **Daily burned area** and carbon emissions from boreal fires in Alaska S. Veraverbeke et al. Title Page Introduction Abstract Conclusions References Tables **Figures** Close Back Full Screen / Esc **Printer-friendly Version** Interactive Discussion

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Table 1. Environmental variables selected to predict ground layer consumption by fire in black spruce forest. The pre-fire tree cover and dNBR layers were also used to predict the aboveground consumption.

Variable	Rationale	Data
elevation (m)	Elevation regulates soil organic layer thickness, carbon density, drainage conditions and seasonal timing of permafrost thaw (Kane et al., 2005, 2007; Kasischke and Johnstone, 2005; Barrett et al., 2010; Kasischke and Hoy, 2012)	ASTER GDEM 2 (Tachikawa et al., 2011).
slope (°)	Slope regulates drainage conditions, organic layer thickness and fire behavior. Sloped terrain is generally better drained than flat terrain (Barrett et al., 2011). Steepness of the terrain regulates tree establishment (Hollingsworth et al., 2006) and fire spread rates and severity (Rothermel, 1972).	derived from ASTER GDEM 2
northness – cosine of aspect	Northness regulates drainage conditions, organic layer thickness and carbon density. Wetness increases with northness, solar insolation decreases with northness. North-oriented slopes are wetter and colder than south-faced slopes and have thicker, (Kane et al., 2007; Turetsky et al., 2011).	derived from ASTER GDEM 2
pre-fire tree cover (%)	Pre-fire tree cover determines the available biomass for aboveground consumption. Aboveground consumption relates to belowground consumption (Rogers et al., 2014). Pre-fire tree cover is also a proxy of stand age, and thus of the thickness, density and wetness of the soil organic layer (Kasischke and Johnstone, 2005; Beck et al., 2011b; Rogers et al., 2013).	tree cover continuous fields 30 m: Sexton et al. (2013) 250 m: Hansen et al. (2003)
time of burning (day of the year)	Time of burning influences the dryness of the soil organic layer during the season (Turetsky et al., 2011; Kasischke and Hoy, 2012)	derived from MODIS active fire data
dNBR	dNBR assesses pre-/post-fire changes in near shortwave infrared reflectance, which relate to changes in vegetation abundance, charcoal deposition and soil exposure (López García and Caselles, 1991; van Wagtendonk et al., 2004; Key and Benson, 2006)	30 m: MTBS (Eidenshink et al., 2007) 500 m: derived from MODIS surface reflectance

ASTER GDEM 2: Advanced Spaceborne Thermal Emission and Reflection Radiometer Global Digital Elevation Model Version 2 (downloaded from http://reverb.echo.nasa.gov, last access: 25 November 2014).

dNBR: differenced Normalized Burn Ratio.

MODIS: Moderate Resolution Imaging Spectroradiometer.

MTBS: Monitoring Trends in Burn Severity (downloaded from http://www.mtbs.gov/, last access: 25 November 2014).

Tree cover data downloaded from http://glcf.umd.edu/data/landsatTreecover/ (30 m), and http://reverb.echo.nasa.gov (500 m), MODIS surface reflectance data from http://reverb.echo.nasa.gov, MODIS active fire data from ftp://fuoco.geog.umd.edu/modis (all last access: on 25 November 2014).





Figure 1. Fires that occurred in the study area between 2001 and 2012 (yellow perimeters) from the Alaska Large Fire Database. The background tree cover map is the Moderate Resolution Imaging Spectroradiometer Vegetation Continuous Fields product (MOD44B, Hansen et al., 2003) for the year 2000. The colored dots represent the location of field plots from Turetsky et al. (2011) (red), Boby et al. (2010) (blue) and Rogers et al. (2014) (green). Note that at the scale of the map, some field plot locations overlap due to their close proximity to each other.





Figure 2. Workflow used to obtain daily burned area and carbon emissions in the Alaskan Fire Emissions Database (AKFED) (dNBR: differenced normalized burn ratio).











Figure 4. Relative importance of variables assessed from the nonlinear multiplicative models using all different combinations of two or more variables for **(a)** depth of burn and **(b)** belowground carbon consumption. (dNBR: differenced normalized burn ratio, DOY: day of the year, tc: tree cover, elev: elevation)





Figure 5. (a) Total carbon consumption map by fire and **(b)** uncertainty, expressed as the SD of the prediction error, for the spatiotemporal domain.





Figure 6. Daily burned area, carbon consumption and emissions from Alaska derived from the Alaskan Fire Emissions Database for the years 2001–2012.



Interactive Discussion



Figure 7. (a) Inter- and (b) intra-annual variability in burned area and carbon emissions in Alaska. In (b), the mean monthly burned area and carbon emissions between 2001 and 2012 is plotted.





Figure 8. Relationship between annual burned area and mean annual carbon consumption. The error bars represent one SD.







Figure 9. Distribution of (a) carbon consumption and (b) pixel-based uncertainties in carbon consumption from the Alaskan Fire Emissions Database between 2001 and 2012. The x axes are labeled with the center of the binning intervals.



Figure 10. Distribution of belowground carbon consumption of field data and the Alaskan Fire Emissions Database (AKFED) between 2001 and 2012. The x axes are labeled with the center of the binning intervals.

