Impact of 2050 climate change on North American wildfire: consequences for ozone air quality

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Abstract

We estimate future area burned in Alaskan and Canadian forest by the midcentury (2046–2065) based on the simulated meteorology from 13 climate models under the A1B scenario. We develop ecoregion-dependent regressions using observed relationships between annual total area burned and a suite of meteorological variables and fire weather indices, and apply these regressions to the simulated meteorology. We find that for Alaska and western Canada almost all models predict significant ($p < 0.05$) increases in area burned at the midcentury, with median values ranging from 150 to 390%, depending on the ecoregion. Such changes are attributed to the higher surface air temperatures and 500 hPa geopotential heights relative to present day, which together lead to favorable conditions for wildfire spread. Elsewhere the model predictions are not as robust. For the central and southern Canadian ecoregions, the models predict increases in area burned of 45–90%. Except for the Taiga Plain, where area burned decreases by 50%, no robust trends are found in northern Canada, due to the competing effects of hotter weather and wetter conditions there. Using the GEOS-Chem chemical transport model, we find that changes in wildfire emissions alone increase mean summertime surface ozone levels by 5 ppbv for Alaska, 3 ppbv for Canada, and 1 ppbv for the western US by the midcentury. In the northwestern US states, local wildfire emissions at midcentury enhance surface ozone by an average of 1 ppbv, while transport of boreal fire pollution further degrades ozone air quality by an additional 0.5 ppbv. The projected changes in wildfire activity increase daily summertime surface ozone above the 95th percentile by 1 ppbv in the northwestern US, 5 ppbv in the high latitudes of Canada, and 15 ppbv in Alaska, suggesting a greater frequency of pollution episodes in the future atmosphere.
Introduction

North American wildfires are important sources of the ozone precursors carbon monoxide (CO), nitrogen oxides (NO\textsubscript{x}), and volatile organic compounds (VOCs). Their emissions can strongly affect air quality locally and, in the case of large fires, in areas thousands of kilometers downwind in the United States and Canada (Wotawa and Trainer, 2000; Morris et al., 2006; Kang et al., 2014), over the mid-Atlantic (Val Martin et al., 2006; Cook et al., 2007), and in Europe (Real et al., 2007). Previous studies have projected increases in the area burned by North American wildfire in the 21st century due mainly to warmer temperatures (Flannigan et al., 2005; Balshi et al., 2009; Wotton et al., 2010; Price et al., 2013; Boulanger et al., 2014), implying further degradation of air quality by wildfire emissions in a changing climate. However, predicted increases in future precipitation in Alaska and Canada (Christensen et al., 2007) may have an opposing effect on future wildfire activity, resulting in large uncertainties in fire projections. In this study, we develop and evaluate stepwise regressions between area burned and observed meteorological variables over North American boreal ecoregions. We apply these regressions with simulated future meteorology from an ensemble of climate models to estimate area burned in Alaskan and Canadian boreal forest by the midcentury. In combination with the projected future area burned in western US by an earlier study, we then quantify fire emissions at the middle of the 21st century and examine its impacts on the ozone air quality in North America.

Wildfires in Canada and Alaska often have much larger size compared with those in the contiguous United States (Stocks et al., 2002; Westerling et al., 2003). Emissions from boreal wildfires can have significant effects on air quality over the contiguous US (Sigler et al., 2003; Miller et al., 2011; Kang et al., 2014). In the summer of 1995, transport of forest fire emissions from northwestern Canada reached as far south as the central and southern US, increasing CO concentrations as much as 200 ppb in that region (Wotawa and Trainer, 2000). The same fires also enhanced ozone in central and southern US by 10–30 ppbv, most of which was associated with NO\textsubscript{x} directly emitted by
the Canadian fires and the remainder with the oxidation of wildfire CO by locally emitted NO\textsubscript{x} (McKeen et al., 2002). The summer of 2004 was one of the most intense fire seasons on record for Canada and Alaska (Turquety et al., 2007; Lavoue and Stocks, 2011). An analysis of flight data over the northeastern US concluded that boreal fire emissions during that summer contributed 10% of the observed CO over the northern United States (Warneke et al., 2006) and enhanced mean summertime ozone there by 1–3 ppbv (Hudman et al., 2009). Smoke plumes occasionally reached Houston that summer, increasing ozone there as much as 30–90 ppbv between the surface and 3 km altitude and likely contributing to violations of the 8 h ozone air quality standard (Morris et al., 2006).

Area burned in North America is influenced by fuel availability, weather, ignition, and fire suppression practices. Many studies, however, have suggested that meteorology is the single most important factor (Hely et al., 2001). For example, Gillett et al. (2004) found that changes in temperature alone explain 59% of the variance of the observed area burned in Canada for 1920–1999. Regression studies using surface meteorological data and fire indices also yield high $R^2$ of 0.4–0.6 for area burned in boreal ecoregions (Flannigan et al., 2005). In addition to the surface weather conditions, the 500 hPa geopotential height is also found to be important in predictions of area burned in boreal forests (Skinner et al., 1999; Wendler et al., 2011), since this variable can indicate the occurrence of blocking highs over the continent, which cause rapid fuel drying (Fauria and Johnson, 2008).

Studies examining climate impacts on wildfire activity in North America have projected increases in area burned over most boreal ecoregions in the 21st century. Flannigan and Van Wagner (1991) developed linear regressions between area burned and fire indices. They applied these regressions with the mean climate simulated by three general circulation models (GCMs) and projected an increase of 40% in Canadian area burned in a doubled CO\textsubscript{2} atmosphere, relative to present day. Flannigan et al. (2005) improved the previous projection with more complete meteorological station data, higher spatial resolution, and a stepwise regression scheme with more poten-
tial regression factors. Their results showed that area burned increases by 70–120 % in boreal ecoregions by 2080–2100, a period with roughly tripled atmospheric CO$_2$ concentrations. However, Balshi et al. (2009) predicted that area burned in Alaska and Canada would double by 2050, a rate more rapid than in the projections by Flannigan et al. (2005). The discrepancies among these studies arise in part from the differences in the climate scenarios as well as the sensitivity of the particular GCMs to increases in greenhouse gases.

In this study, we investigate the impact of changing climate on future Alaskan and Canadian area burned and the consequences for ozone air quality in North America by 2046–2065 under a moderately warming scenario. Wildfires produce abundant ozone precursors, and many, but certainly not all, observational studies of boreal fires suggest subsequent ozone generation either locally or downwind (Jaffe and Wigder, 2012). We build here on our earlier study (Yue et al., 2013), which projected future area burned in the western US using stepwise regressions and the simulated climate from an ensemble of climate models from the World Climate Research Programme’s (WCRP’s) Coupled Model Intercomparison Project phase 3 (CMIP3) multi-model dataset (Meehl et al., 2007a). Yue et al. (2013) predicted that the warmer and drier summer climate over the western US at mid-century would increase area burned there by 60 % and the consequent biomass burned by 77 %.

Yue et al. (2013) further calculated regional increases of 46–70 % in surface organic carbon aerosol and 20–27 % in black carbon aerosol due to the increased fire emissions. For this study, we focus on ozone air quality. We rely on the CMIP3 ensemble of climate models to obtain confidence in projections of boreal area burned, and we combine these results with those of Yue et al. (2013) for the western US. Using the estimated fuel consumption and emission factors for ozone precursors, we calculate future fire emissions over North America. Finally, we quantify the impacts of those emissions on ozone mixing ratios at the midcentury, using the GEOS-Chem chemical transport model (CTM) driven by the Goddard Institute for Space Studies General Circulation Model 3 (GISS GCM3).
2 Data and methods

2.1 Boreal ecoregions

We divide Alaskan and Canadian forests into 12 ecoregions (Fig. 1), following the definitions of the Ecological Stratification Working Group (1996) with modifications by Stocks et al. (2002) and Flannigan et al. (2005). Area burned outside these ecoregions is small. In northern Canada cold weather and the lack of fuel continuity for the tundra and mountainous regions limits fire activity (Stocks et al., 2002), while regulations restrict agricultural burning in the southern part of central Canada.

We describe the 12 ecoregions as follows. Located in the central Alaska, the Alaska Boreal Interior consists mainly of plains and hills and is covered with Arctic shrubs and open coniferous forest. The Taiga Cordillera in western Canada has similar vegetation, although the higher elevation leads to lower temperatures. Three western ecoregions, the Alaska Boreal Cordillera, the Canadian Boreal Cordillera, and the Western Cordillera are located along the Rocky Mountains. The high elevation causes abundant precipitation, especially for the Western Cordillera, resulting in dense forests. In contrast, the two central Canadian ecoregions, the Taiga and Boreal Plains, are at lower altitudes and are characterized by tundra meadow and aspen forest. The Western Taiga Shield is a plain in north central Canada characterized by shrub and conifer forests. The Hudson Plain, to the south of Hudson Bay, is dominated by wetlands. Stocks et al. (2002) defined the Eastern Taiga Shield as covering most of northern Quebec. Here we redefine this ecoregion so that it covers just the southwestern part, where ~ 90 % of the area burned in the original ecoregion occurs. We divide the Mixed Wood Shield, a large ecoregion in southeast Canada, into eastern and western parts. Fire activity in these two subregions is significantly different (Flannigan et al., 2005).
2.2 Fire data

We compile monthly $1° \times 1°$ area burned from 1980 to 2009 based on interagency fire reports. For Alaska, we use incidence reports managed by the National Wildfire Coordinating Group from the Fire and Aviation Management Web Applications (FAMWEB, http://fam.nwcg.gov/fam-web/weatherfirecd/, downloaded on 5 June 2012). Five agencies, the U.S. Forest Service (USFS), Bureau of Land Management (BLM), Bureau of Indian Affairs (BIA), Fish and Wildlife Service (FWS), and National Park Service (NPS), provide $\sim 10 000$ records of fire incidence in Alaska between 1980 and 2009. Each record documents the name, location (latitude and longitude), start and end time, ignition source (lightning or human) and area burned of an individual fire. Duplicates are expected because fires burn in lands managed by different agencies (Kasischke et al., 2011). We identify and delete duplicate records if two or more fires have same names and areas, and occur within a distance of 50 km on the same day. Thus we obtain a corrected subset and compare it with the annual fire report from the National Interagency Coordination Center (NICC, http://www.nifc.gov/nicc/). NICC manages fire reports from federal agencies, states, and private ownership, and so has more complete datasets relative to FAMWEB. NICC, however, provides annual total area burned only back to 1994. The correlation $R$ between FAMWEB and NICC is 1.0 and the differences are within 2% for 1994–2009, giving us confidence in our compilation of FAMWEB area burned.

For Canada, we use fire point data from the Canadian National Fire Database (CNFDB, http://cwfis.cfs.nrcan.gc.ca/ha/nfdb), which is an extension of the Large Fire Database (LFDB) summarized in Stocks et al. (2002). The database provides over 240 000 records of forest fires during 1980–2009, collected from provinces, territories, and Parks Canada. Each CNFDB record includes the name, location, size, and time of one fire. However, duplicates in CNFDB are much fewer, possibly because the redundant records were deleted when the dataset was compiled into a Geographic Information System. Although the total number of fires is immense, only about 5% are greater
than 100 ha. These large fires account for over 99% in area burned in the dataset, as was the case for the LFDB.

We aggregate both the FAMWEB and CNFDB report data onto 1° x 1° grids, based on the location of fires. Area burned is assigned to the start month, as end dates are often uncertain (Kasischke et al., 2011). The monthly gridded area burned is used to derive fire emissions. To develop the fire models, we aggregate the fire report data into boreal ecoregions across Alaska and the Canadian boreal forest (Fig. 1) and then sum the area burned within each ecoregion for the entire fire season (May–October) to reduce noise in the regression.

2.3 Meteorological data and fire weather indices

We use daily observations for 1978–2009 from the Global Surface Summary of the Day dataset (GSOD, http://www.ncdc.noaa.gov/). The length of meteorological data is two years longer than that of fire data, because the regressions employ terms that depend on the weather occurring up to 2 years before the area burned. The GSOD provides 18 daily surface meteorological variables for over 2000 stations in Alaska and Canada. We select 157 sites within the 12 ecoregions that provide observations for at least two thirds of the days during 1978–2009 (Fig. 1). We use daily mean and maximum temperature, total precipitation, and wind speed and calculate relative humidity using daily mean temperature and dew point temperature. We also use the 500 hPa geopotential height from the North American Regional Reanalysis (NARR, Mesinger et al., 2006). Both the site measurements and the NARR reanalysis data are binned into ecoregions to derive monthly averages.

The site observations are also used as input for the Canadian Fire Weather Index system (CFWIS, Van Wagner, 1987). The CFWIS uses daily temperature, relative humidity, wind speed, and total precipitation to calculate three fuel moisture codes and four fire severity indices. The fuel moisture codes indicate moisture levels for litter fuels (Fine Fuel Moisture Code, FFMC), loosely compacted organic layers (Duff Moisture Code, DMC), and deep organic layers (Drought Code, DC). The FFMC is combined
with wind speed to estimate the Initial Spread Index (ISI). The DMC and DC are used to derive the Build-up Index (BUI) to indicate the availability of fuel. The ISI and BUI are then combined to create the Fire Weather Index (FWI) and its exponential form as the Daily Severity Rating (DSR). The CFWIS indices have been widely used in fire-weather research over North America (Amiro et al., 2004; Flannigan et al., 2005; Balshi et al., 2009; Spracklen et al., 2009), and in our previous work (Yue et al., 2013).

2.4 Regression approach

We use total area burned during the fire season as the predictand, and we assume that the influences of both topography and fuels on wildfire activity are roughly uniform across each region. Five meteorological variables (mean and maximum temperature, relative humidity, precipitation, and 500 hPa geopotential height) are used to calculate the seasonal (winter, spring, summer, and autumn), annual, and fire-season means. The mean and maximum values of the seven daily CFWIS indices during fire season are also included in the regressions, resulting in 44 terms dependent on the current year. As in Yue et al. (2013), we also employ variables from the previous two years in the regression, making 132 potential terms for the regression.

We set up two criteria to select a factor as a predictor at each step. First, the chosen factor must have the maximum contribution to the $F$ value, a metric for variance, of the predictand among the unselected factors. Second, this factor must exhibit low correlation with those already selected, with $p$ value $> 0.5$. The first criterion produces a function with the largest possible predictive capability, while the second helps increase the stability of the function by introducing independent predictors (Philippi, 1993).

In Yue et al. (2013), we also developed a parameterization for area burned in the western US. The parameterization was a function of temperature, precipitation, and relative humidity. The same functional form was applied throughout the domain, scaled by an ecoregion-dependent fire potential coefficient. We find that the parameterization approach fails in boreal forests, probably because the driving factors for wildfires vary greatly over the vast boreal areas.
2.5 CMIP3 model data

We use daily output from 13 climate models in the CMIP3 archive (Meehl et al., 2007a) for the fire projection (Table S1 in the Supplement). The variables we select include daily mean and maximum temperature, total precipitation, and surface wind speed. We calculate daily RH for the CMIP3 models using other archived meteorological variables. We also use the monthly mean 500 hPa geopotential heights from all 13 GCMs. We use the output from the 20C3M scenario for the prediction of area burned in the present day (1981–1999). For the future atmosphere (2046–2064), we use the simulated climate under the A1B scenario, which assumes a greater emphasis on non-fossil fuels, improved energy efficiency, and reduced costs of energy supply. CO₂ reaches 522 ppm by 2050 in this scenario (Solomon et al., 2007), resulting in a moderate warming relative to other scenarios (Meehl et al., 2007b). Over this relatively short timeframe, the A1B scenario is consistent with two moderate scenarios in the newer Representative Concentration Pathways, RCP 4.5 and RCP6.0 (Moss et al., 2010). We aggregate all the climate simulations into ecoregions for the projection. In order to reduce model uncertainties, we scale the aggregated variables from each GCM using the mean observations for 1980–2009 from the GSOD sites. The changes in area burned and meteorological variables are examined with a Student t test and only those with p < 0.05 are considered as significant.

2.6 Fuel consumption

Fuel consumption is the amount of both live and dead biomass burned per unit area. It depends on both fuel load and burning severity. In Yue et al. (2013), we estimated fuel load over the western US using the 1 km dataset from the USFS Fuel Characteristic Classification System (FCCS, http://www.fs.fed.us/pnw/fera/fccs/, McKenzie et al., 2007). The FCCS defines ~300 types of fuelbed based on the distribution of vegetation types from the Landscape Fire and Resource Management Planning Tools (LANDFIRE, http://www.landfire.gov/). Each type of fuelbed consists of seven basic
fuel classes (i.e., light, medium, heavy fuels, duff, grass, shrub, and canopy) each with a different load (Ottmar et al., 2007). Here, for Canada, we use the 1 km fuel type map from the Canadian Fire Behavior Prediction (FBP) system, which is derived from remote sensing and forest inventory data and includes just 14 types (Nadeau et al., 2005). For Alaska, we use a fuel map created by the USFS, which also follows the classification scheme of Nadeau et al. (2005). However, the FBP system does not provide fuel load, and so we follow Val Martin et al. (2012), who matched the Canadian FBP fuelbeds with their corresponding types in the US. FCCS and in this way estimated the fuel load for both Canada and Alaska (see their Table A1).

Burning severity indicates the fraction of fuel load burned by fires and varies by moisture state. We follow the approach of Val Martin et al. (2012), who used the USFS CONSUME model 3.0 (Ottmar, 2009) to calculate burning severity and the resulting fuel consumption for a given fuel load. In this approach, the derived FBP fuel loads are applied to CONSUME, yielding reference fuel consumption for five moisture conditions: wet, moist, moderately dry, dry, and extra dry (Val Martin et al., 2012). Here we use a newer model version, CONSUME-python (https://code.google.com/p/python-consume/), which fixes some errors in CONSUME 3.0. The updated reference fuel consumption for different FBP fuel types and moisture states is given in Table S2. Our values for C3 (mature jack or lodgepole pine) and C5 (red and white pine) fuel types are 40–65% lower than those in Val Martin et al. (2012), likely because of errors in the calculation of duff fuel in CONSUME 3.0. We aggregate the new 1 km fuel consumption map to 1° resolution to match that of gridded area burned. Figure 2a shows fuel consumption for moderately dry conditions. The figure shows heavy fuel consumption of > 7 kg dry matter (DM) m⁻² in the Taiga Plain and in the Western and Eastern Mixed Wood Shield, where boreal spruce fuel types (C2) dominate.

We rely on the DC index from the CFWIS in order to assign the moisture condition and determine the monthly fuel consumption. This index is a good indicator for fuel moisture content (Bourgeau-Chavez et al., 1999; Abbott et al., 2007) and has been widely used to calculate fuel consumption (e.g., de Groot et al., 2009; Kasischke and
Hoy, 2012). Higher DC values indicate greater dryness. Figure S1 in the Supplement shows the monthly mean DC in boreal ecoregions for 1980–2009. The values of DC increase gradually from May to September, as fuels become progressively drier. The DC values in western ecoregions are usually higher than those in eastern ones, probably because precipitation in the West (except for the Pacific coast) is much lower relative to that in the East (not shown).

Figure S2 shows the cumulative probability of daily DC in all ecoregions during the fire seasons of 1980–2009. This probability distribution differs somewhat from the distributions in Amiro et al. (2004) who estimated DC for Canadian wildfires larger than 2 km² in different ecosystems during 1959–1999. Such fires typically occur in June to August. In contrast, Fig. S2 shows the DC distribution over the entire fire season, including days in September and October, when DC values are usually very high. We relate burning severity to DC by defining four arbitrary thresholds in the DC probability distribution: 85, 65, 35, and 15 %. The resulting moisture categories and their average DC indices are as follows: extra dry (DC > 85 %, 774), dry (65 % < DC ≤ 85 %, 590), moderately dry (35 % < DC ≤ 65 %, 390), moist (15 % < DC ≤ 35 %, 196), and wet (DC ≤ 15 %, 53). We then calculate the monthly fuel consumption in each ecoregion by matching the DC in that month to these moisture categories and choosing the appropriate fuel consumption (Table S2). In this way, fuel consumption varies yearly and seasonally. Amiro et al. (2004) found that the average DC for Canadian wildfires ranges from 210 to 372 depending on the ecoregion, and the cumulative probability of the DC also varies with ecoregion. Here we have chosen to use a single distribution for the North American boreal region to define the DC thresholds (Fig. S2). As a check, we also compare the fuel consumption derived in this way with that calculated based on the ecoregion-specific DC thresholds.

We assume that the fuel load remains constant for both present day and midcentury, based on the conclusion that changes in forest composition will be a gradual process (Hanson and Weltzin, 2000). Fuel consumption, however, does change in our approach since it depends on the moisture state. We estimate fuel consumption for
both present day and midcentury based on the multi-model median DC in each ecoregion. As a result, the modeled fuel consumption responds to trends in fuel moisture conditions. Amiro et al. (2009) performed a similar estimate of future boreal fuel consumption using modeled monthly mean values of the DC and an empirical relationship derived by de Groot et al. (2009) for forest floor fuel consumption in experimental fires in Canada. However, this empirical relationship has predictive capability only for fires set under experimental conditions, but not for wildfires (de Groot et al., 2009), and we do not apply it here.

2.7 Estimate of gridded fire emissions

We calculate biomass burned as the product of area burned and fuel consumption. The annual area burned estimated with regressions for each ecoregion (Sect. 2.4) is first converted to monthly area burned using the mean seasonality for each boreal ecoregion, on the basis of the observations for 1980–2009. Large fires tend to burn in ecosystems with a history of similarly large fires (Keane et al., 2008). Fuel availability, however, limits reburning in the same location during the forest return interval, which is typically ~200 years for Canadian forests (Ter-Mikaelian et al., 2009; de Groot et al., 2013). We assume a random distribution of area burned within each ecosystem, to allow for these tendencies.

We spatially allocate monthly area burned within each ecoregion to 1° × 1° as follows. In each 1° × 1° grid square we calculate the frequency of fires larger than 1000 ha during 1980–2009; such fires account for ~85% of total area burned in Canada and Alaska over this time period. Accordingly, we arbitrarily attribute 85% of area burned within each ecoregion to fires of 1000 ha in size, and we then allocate these large fires among the 1° × 1° grid squares based on the observed spatial probability of large fires (>1000 ha), which is the percentage of total large fires of the ecoregion located in a specific grid box during this timeframe. We then disaggregate the remaining 15% of area burned into fires 10 ha in size, and randomly distribute these fires across all grid boxes in the ecoregion. We apply this random approach to calculate both present
day (1997–2001) and future (2047–2051) biomass burned. Within each timeframe, the
effect of limited fuel availability in the aftermath of a fire is taken into account by reeval-
uating the spatial probability distribution of area burned at each monthly time step. We
scale the observed probabilities by the fraction remaining unburned in each grid box,
and then use this modified probability distribution to allocate large fires for the remain-
ing months. Using sensitivity tests, we find that specifying different areas burned to the
large fires (100 ha or 10 000 ha rather than 1000 ha) yields < 1 % changes in predicted
biomass burned, suggesting that this approach is not sensitive to the presumed fire
size in the allocation procedure.

We take the emission factors for all ozone precursors except nitric oxide (NO) from
Andreae and Merlet (2001). For NO we average the values from six studies of forest
fires in the western US (Table S3), yielding 2.2 g NO$_x$ kg DM$^{-1}$. Based on the measure-
ments by Hegg et al. (1990), which showed that NO contributes 30 % of fire-induced
NO$_x$, this value is equivalent to 1.6 g NO kg DM$^{-1}$, consistent with the mean emission
ratio of 1.4 g NO kg DM$^{-1}$ derived from measurements from Alaskan fires (Nance et al.,
1993; Goode et al., 2000). Our NO emission factor is ~ 50 % higher than that derived
by Alvarado et al. (2010) from aircraft measurements of boreal fire plumes. They also
found that 40 % of NO$_x$ emissions are rapidly converted to peroxyacetyl nitrate (PAN) in
fresh plumes. We use the emission factor of 1.6 g NO kg DM$^{-1}$ and neglect the rapid for-
mation of PAN for our simulations, recognizing that this likely leads to an overestimate
of ozone formation immediately downwind of the fires.

2.8 GEOS-Chem CTM and simulations

We simulate tropospheric ozone-NO$_x$-VOC-aerosol chemistry using the GEOS-Chem
global 3-D model of tropospheric chemistry version 8.03.01, driven by present-day and
future simulated meteorological fields from the NASA/GISS Model 3 with 4° × 5° reso-
lution (Wu et al., 2007, 2008b). Compared with finer resolution, 4° × 5° resolution does
not induce a significant bias in surface ozone and captures the major synoptic fea-
tures over the United States (Fiore et al., 2002, 2003), though it may underestimate the average ozone level by 1–4 ppbv and predict fewer pollution episodes (Wang et al., 2009; Zhang et al., 2011). The present-day GISS-driven GEOS-Chem simulation as used here has been evaluated by comparison with climatological ozonesonde data and reproduces values throughout the troposphere usually to within 10 ppbv (Wu et al., 2007). For the US, the simulation reproduces the summertime temporal variability of ozone concentrations leading to pollution episodes, although it overestimates surface ozone by up to 10 ppbv in Midwest, due to weakened surface winds in GISS GCM relative to observations (Wu et al., 2008b).

Anthropogenic emissions for ozone precursors, including NO$_x$, CO, and non-methane VOCs as described in Table 1 of Wu et al. (2008b), are kept constant at the level of the year 2000 for both present day and future simulations, to isolate the effects of changes in biomass burning emissions. However, natural emissions of these gases from vegetation, soil, and lightning are computed locally based on the meteorological variables within the model and allowed to change with climate. Emissions of biogenic hydrocarbons are calculated with the Model of Emissions of Gases and Aerosols from Nature (MEGAN), version 2.1 (Guenther et al., 2006). The lightning source of NO$_x$ is computed locally in deep convection events using the scheme of Price and Rind (1992), which relates number of flashes to convective cloud top heights, together with the vertical NO$_x$ distribution from Pickering et al. (1998). Stratosphere–troposphere exchange (STE) is specified by the Synoz flux boundary condition (McLinden et al., 2000) with a prescribed global annual mean flux of 495 Tg ozone a$^{-1}$ for both present day and future simulations. Outside of North America, we use climatological biomass burning emissions derived from the inventory described in Lobert et al. (1999), with seasonality from Duncan et al. (2003) and placed into the boundary layer.

Over North America, we apply the biomass burning emissions predicted by our method. For the western US, we use area burned predicted with regressions from Yue et al. (2013). We update the fire emissions over southern California with our improved fire scheme (Yue et al., 2014). For Canada and Alaska, we use the fire emissions
derived from calculated area burned and the estimated fuel consumption. We do not change the emissions over the eastern US, which are dominated by prescribed agricultural fires (Liu, 2004). As in Leung et al. (2007), we emit 20% of emissions between 3 and 5 km and the rest in the boundary layer, as observations have shown that over 80% of plumes from North America fires are located in the boundary layer (Val Martin et al., 2010).

Surface ozone concentrations in the 21st century will be influenced not just by trends in wildfire emissions, but also by changes in atmospheric transport, temperature, cloudiness, wet and dry deposition, and natural emissions. To isolate the changes due to biomass burning emissions, we conduct an ensemble of 5 year simulations for present day (1997–2001) and the mid-21st century (2047–2051) for a total of 7 sensitivity studies (Table 1). Two simulations, FULL_PD and NOFIRE_PD, are carried out with present-day climate: FULL_PD considers present-day fire emissions from both western US and boreal forests, while NOFIRE_PD omits any fire emissions in these regions. The other five simulations are conducted with future climate. In FULL_A1B, we implement the projected future fire emissions from western US and boreal forests, while NOFIRE_A1B omits these emissions. Simulation WUS_FIRE applies future fire emissions in western US but the present-day emissions in boreal forests. In contrast, BOREAL_FIRE uses present-day emissions in western US but the future ones for boreal regions. The last simulation with future climate, CLIM_CHAN, applies present-day fire emissions everywhere as in FULL_PD.

We examine the differences between FULL_PD and NOFIRE_PD to quantify the impacts of wildfire emissions in the present day, and the differences between FULL_A1B and NOFIRE_A1B to quantify these impacts at midcentury. We use the differences between FULL_A1B and BOREAL_FIRE to isolate the impacts of increased fire emissions in western US at midcentury. The differences between FULL_A1B and WUS_FIRE reveal the effects due to changes of fire emissions in boreal forests, also at midcentury. Finally, the differences between CLIM_CHAN and FULL_PD represent the impacts due solely to climate change on the simulated ozone concentrations. Each
model run was initialized with a 1 year spin-up. Taken together, these 7 cases yield insight into the influence of changing wildfire activity on surface ozone concentrations across North America, and the relative importance of local vs. remote wildfires on US and Canadian ozone air quality.

3 Results

3.1 Regressions and predictions of area burned at present day

Figure 3a shows observed, annual mean area burned for 1980–2009 averaged over the boreal ecoregions. In Canada, the Western Mixed Wood Shield exhibits the greatest area burned of nearly $7 \times 10^5 \text{ha a}^{-1}$. In addition, large area burned of $\sim 4 \times 10^5 \text{ha a}^{-1}$ and $\sim 3 \times 10^5 \text{ha a}^{-1}$ is observed in the Taiga Plain and the Western Taiga Shield. Most fires in these very remote ecoregions are allowed to burn naturally, without intervention. This practice, together with the hot summers typical of continental interiors, leads to large area burned (Stocks et al., 2002). The Western Cordillera shows the least area burned, at $0.4 \times 10^5 \text{ha a}^{-1}$, due to abundant rainfall as well as active fire suppression (Stocks et al., 2002). Fires in Alaska are about three times larger in the Alaska Boreal Interior than in the Alaska Boreal Cordillera, because the summer in interior Alaska is warmer and drier relative to the southern part, which is influenced by moisture from the Pacific (Wendler et al., 2011). In each ecoregion, the top three largest fire years account for 36–67% the total area burned in 1980–2009, with the largest fraction in the Alaska Boreal Cordillera (Fig. 4).

Table 2 shows the regressions we developed between area burned and the suite of meteorological variables and fire weather indices in each ecoregion. These fits explain 34–75% ($p < 0.001$) of the variance in area burned (Fig. 3b). In most ecoregions, the regressions capture well the interannual variations of area burned, although they usually underestimate the values for extreme years (Fig. 4). For the top three large fire years in each ecoregion, the predictions underestimate the total area burned by 22–
57 %, with the worst match in the Hudson Plain. Such failure in predicting extreme fires is a common weakness of fire models, no matter the approach – e.g., regressions (Balshì et al., 2009; Spracklen et al., 2009; Yue et al., 2013), parameterizations (Crevoisier et al., 2007; Westerling et al., 2011), and dynamic global vegetation models (DGVMs; Bachelet et al., 2005).

We find that meteorological variables for the current year are selected as the first term in ten of the twelve ecoregions, indicating that area burned in the boreal forests is most related to current weather (Table 2). In contrast, Westerling et al. (2003) suggested that wildfire activity in shrub ecoregions in the western US is closely related to meteorology in previous years, because the antecedent moisture levels can control fuel growth. In boreal forests, however, fuel load is perennially abundant, and so weather in the current year is more important here. Our regressions show that the 500 hPa height is the dominant factor affecting boreal fires, as it appears in eight regression fits and is selected as the first term for three of them. Temperature, which highly correlates with geopotential height ($R > 0.85$) in spring and summer, is selected as the first term in three other ecoregions. Of the six ecoregions which have either geopotential height or temperature as the first term, five are located in Alaska and western Canada, suggesting that wildfire activity in these areas is greatly influenced by temperature or by blocking highs which lead to persistent hot and dry conditions. Since our regression method does not permit correlation among the predictors, temperature and geopotential height are not selected for the same season and year in any of the ecoregions. Fire indices, which combine the impacts from temperature, humidity, and wind speed, are the dominant predictors in the four central Canadian ecoregions. In three of these four regions, moisture variables such as relative humidity and precipitation are also selected. Our method yields relative humidity as the leading term in the two eastern ecoregions, indicating that the dryness of fuel is most important for wildfire activity there.

Our results confirm that wildfires in Alaska and western Canada are related to geopotential height anomalies, which are associated with the positive phase of either the
Pacific North American (PNA) pattern or the Pacific Decadal Oscillation (PDO; Fauria and Johnson, 2006, 2008). Although such height anomalies may influence wildfire activity in other boreal ecoregions (Fauria and Johnson, 2006), they tend to correlate with fire weather indices or moisture variables and so may not be selected as terms in our regressions.

We compared our results with those in Flannigan et al. (2005), who developed regressions in similar ecoregions. Relative to their $R^2$ of 0.56 and 0.60 in the Taiga Plain and the Western Mixed Wood Shield, where large area burned is observed (Fig. 3a), our regressions yield higher $R^2$ of 0.75 and 0.67. This improvement may result from our use of meteorological data with better spatial coverage or our inclusion of terms dependent on the meteorology in previous years. However, our regressions in the Western Taiga Shield, the Eastern Taiga Shield, and the Hudson Plain explain 34–46 % of the variance in observed area burned, much lower than the 64 % predicted in Flannigan et al. (2005), which aggregated these three ecoregions into one. The larger domain in Flannigan et al. (2005) apparently smoothed spikes in the area burned data (Fig. 4) and as a result increased the $R^2$ for regressions (Spracklen et al., 2009). We treat the three regions separately due to their very different ecologies.

We next calculate present-day (1983–1999) area burned by applying present-day meteorological fields from the 13 GCMs to our regressions. We start with 1983 since we need to apply factors from the previous two years in the regressions. As Fig. 5a shows, in eight ecoregions the median area burned from the ensemble of GCMs matches the observations within ±15 %. However, the predicted area burned is overestimated by 54 % in the Eastern Taiga Shield and underestimated by 30 % in the Taiga Plain. These biases do not derive from the long-term mean model meteorology, since we scale the simulated fields with means from observations. Instead, the biases arise from our use of fire weather indices in the regressions, which depend on the daily variability in meteorology. For example, in the Taiga Plain, the predicted median ISI is lower than observed by 7 %. In the same ecoregion, the site records show that more than 30 % of days have precipitation less than 0.1 mm day$^{-1}$ during fire seasons for 1980–
2009. However, the GCMs predict only 2–13% days with < 0.1 mm day\(^{-1}\), even after scaling with the means from observations. In contrast, they predict 55–65% of days with rainfall of 0.1–1.0 mm day\(^{-1}\), much more than the 37% from observations. The overprediction of drizzle, a common problem in GCMs, results in lower ISI compared with observations. The same problem in modeled precipitation also reduces the predicted DMC\(_{\text{max}}\) in the Eastern Taiga Shield, leading to an overestimate in area burned when applied with a negative coefficient. Flannigan et al. (2005) reported a similar problem in their study, and they subtracted a constant from the GCM precipitation to match the observed rainfall frequency. We do not follow this approach because our predicted present-day median area burned agrees reasonably well with that observed. The non-linear response of fire weather indices to daily meteorology contributes to the uncertainty of predictions, resulting in larger spread of ratios for those ecoregions whose regressions depend on the fire indices (Table 2).

### 3.2 Projection of area burned at midcentury

Figure 6 shows the changes in key meteorological variables at midcentury relative to present day, as predicted by the 13 GCMs. Temperatures across all ecoregions show median increases of ~ 2°C during the fire season, with all models predicting significant changes. Meanwhile, precipitation rates increase by 0.05–0.23 mm day\(^{-1}\) in the median, likely as a result of a poleward shift of mid-latitude storm tracks and precipitation (Yin, 2005). However, these increases in precipitation are significant for only 4 to 8 GCMs, depending on the ecoregion, and in some ecoregions some models project a drier climate by midcentury, reflecting the large uncertainty in model projections of regional hydrology (Christensen et al., 2007). The 500 hPa geopotential heights are predicted to rise by 2050, with median increases of 30–60 m (0.6–1%) and these changes are significant for all GCMs.

We find that the wildfire response to these trends in meteorological variables varies greatly by ecoregion, with large increases in area burned by 2050 in Alaska and west-
ern Canada, but little or no change in area burned elsewhere (Fig. 5b). The median area burned at midcentury increases by 130–350 % in Alaska and the western Canadian ecoregions, relative to present day (Figs. 5b, 7a and Table 3). The greatest increase in area burned occurs in the Alaska Boreal Cordillera, where area burned at the midcentury is more than four times that of the present day. These increases in Alaska and western Canada are largely driven by changes in temperature and/or geopotential height (Table S4), and as a result are statistically robust in 11 to 13 GCMs, depending on the ecoregion (Fig. 7b). The central and southern Canadian ecoregions show more moderate and less robust increases in area burned of 40–90 %, with only 3–8 models projecting significant changes. In these ecoregions, fire activity depends either on hydrological variables (e.g., RH for the Eastern Mixed Wood Shield) or on fire indices that combine effects from temperature and moisture (e.g., the fire indices DSR and FWI in the Boreal Plain and the fire index BUI in the Western Mixed Wood Shield; Table 2). As a result, the effects of increased precipitation in these ecoregions may partly offset the effects of rising temperatures on wildfires.

In some of the most northern ecoregions within the Canadian interior, median area burned decreases in the wetter climate of the midcentury. In the Taiga Plain, the median area burned decreases by 50 % (Table 3, Fig. 7a) despite the 1.7 °C increase in temperature (Fig. 6a). The increase in rainfall decreases the fire index ISI in the Western Taiga Shield but enhances the fire index DMC in the Eastern Taiga Shield (Table S4), with the net result of small median decreases in area burned in these ecoregions that are not significant (Table 3, Fig. 7b). Changes in regression terms for these regions have similar magnitude but opposite signs (Table S4).

Our projection of larger increases in Alaska and western Canadian ecoregions are consistent with the observed trends for 1959–1999 in Kasischke and Turetsky (2006) and with the projection by Flannigan et al. (2005) for 2080 to 2100. However, Flannigan et al. (2005) predicted area burned increases of 40–60 % in the Taiga Plain with 3 × CO₂, where we project a decrease of 50 % with ∼ 1.5 × CO₂. The reasons for this discrepancy are not clear. In our results, a median increase of 0.1 mm day⁻¹ in sum-

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mer precipitation drives the decrease in area burned in the Taiga Plain, but Flannigan et al. (2005) did not report their trend in modeled precipitation. In addition, our regression for the Taiga Plain has ISI as the leading term, while the leading term in Flannigan et al. (2005) is temperature. Based on the same GCM meteorology as Flannigan et al. (2005) and using a similar approach, Amiro et al. (2009) found a modest increase of 10 % in area burned with $2\times CO_2$ for the Taiga Plain, the lowest enhancement among all Canadian ecoregions for that study.

### 3.3 Estimate of future fire emissions

We first compare our derived fuel consumption with previous studies. Figure 8a shows the mean annual biomass burned for 1980–2009, calculated from monthly areas burned and monthly fuel consumption (Sect. 2.6). Figure 2b shows the mean fuel consumption per unit area during the fire season for 1980–2009. We find that the mean fuel consumption per unit area is $\sim 30\%$ less than that for moderately dry conditions for which we assumed an average DC of 390 (Fig. 2). Most boreal area burned occurs during the relatively moist months of June and July (Fig. S1), when the monthly average DC is usually less than 370 (Amiro et al., 2004). In the eastern ecoregions (Hudson Plain, Eastern Taiga Shield, and Eastern Mixed Wood Shield), the values for mean fuel consumption are as much as 50 % less than those for moderately dry conditions due to high moisture content in fuel there (Fig. S1).

In Table 4 we compare our estimates for mean fuel consumption with those from other studies, which were derived from forest inventories and field measurements (French et al., 2000; Balshi et al., 2007), fuel-weather models (Amiro et al., 2001, 2009), and biogeochemical models based on satellite observations (van der Werf et al., 2010). We also compare our results with estimates based on wildfire incidents (Table S5). In the Alaska Boreal Interior, our estimate of 5.5 kg DM m$^{-2}$ is within $\sim 10\%$ of those by Balshi et al. (2007) and van der Werf et al. (2010), but is $\sim 25\%$ lower than that of French et al. (2000). Turetsky et al. (2011) collected data from 178 sites in the Alaskan black spruce ecosystem and estimated that average fuel consumption
is 5.9 kg DM m\(^{-2}\) for early season fires (May–July) but increases to 12.3 kg DM m\(^{-2}\) for late season fires (after 31 July; Table S5). Based on our compilation of fuel consumption (Table 2) and the calculated monthly DC values for Alaska (Fig. S1), we find similar results of 6.1 kg DM m\(^{-2}\) for May–July and 14.6 kg DM m\(^{-2}\) for August–October for C2 fuel (boreal spruce). A recent analysis by French et al. (2011) showed that different models of fuel consumption provide very different results for a given fire, with a range of 2.7–12.2 kg DM m\(^{-2}\) for a major fire in Alaska in 2004 (Table S5). The CONSUME model (v. 3.0) yielded 2.8–4.7 kg DM m\(^{-2}\) for moderate to very dry conditions for that fire, while a field study estimated 5.2 kg DM m\(^{-2}\) (French et al., 2011).

There is less consistency among different estimates of mean fuel consumption in the Canadian ecoregions (Table 4). Our estimates fall in the range of previous work for most ecoregions except for the Western Cordillera and the Taiga Plain, where our values are ~100\% higher than most other estimates. These two ecoregions are located in the western Canada, where seasonal DC is usually high, indicating relatively dry conditions (Fig. S1). Our moisture categories derived from the single DC probability distribution (Fig. S2) may underestimate fuel dryness in the west. On the other hand, our estimates show low fuel consumption in the eastern ecoregions, such as Eastern Taiga Shield, Hudson Plain, and Eastern Mixed Wood Shield, consistent with most of other studies. In a sensitivity test, we derive fuel consumption with regional DC thresholds based on ecoregion-specific probability distributions. This approach reduces western fuel consumption by 8–16\%, but increases eastern values by 2–37\% (Table 4). It also predicts lower Alaskan fuel consumption compared with other studies (Tables 8–9). The boreal biomass burned calculated with this alternative approach is about 156.2 Tg DM a\(^{-1}\) for 1980–2009, almost identical to that estimated using a single probability distribution to define the DC thresholds (Fig. 8a).

We estimate fuel consumption at present day and midcentury with the median DC values from the multi-model ensemble. The present-day values are close to the ones based on observed meteorology (Table 4). By the midcentury, DC values increase in
the warming climate, indicating drying, and fuel consumption increases by 2–22 %, depending on the ecoregion, with a 9 % average enhancement.

Using the random method described in Sect. 2.7, we derive gridded area burned based on the projection with regressions. The estimated biomass burned, averaged over 1997–2001 (Fig. 8b) correlates with observations averaged over 1980–2009 (Fig. 8a) with $R^2 = 0.5$ for $\sim$ 1700 boreal grid squares, indicating that our prediction captures the observed spatial pattern reasonably well. The total biomass burned of 160.2 Tg DMa$^{-1}$ is just 2.5 % higher than that obtained with the observed area burned. Our value of biomass burned using the regression yields emissions of 0.27 Tga$^{-1}$ for NO and 18.6 Tga$^{-1}$ for CO in Alaska and Canada at the present day. By the midcentury, we find that total biomass burned across the boreal ecoregions increases by $\sim$ 90 % (Fig. 8c) due to the $\sim$ 70 % increase in area burned and the $\sim$ 10 % increase in average fuel consumption (Table 4). In Alaska, the maximum increase of 36 Tg DMa$^{-1}$ (168 %) is predicted for the Alaska Boreal Interior, where area burned by the 2050s increases by 146 % (Table 3). In Canada, the Western Mixed Wood Shield has the highest increase of 29 TgDMa$^{-1}$ (64 %). These changes in biomass burned result in increases of 0.24 Tga$^{-1}$ for NO emissions and 17.1 Tga$^{-1}$ for CO in boreal regions. Over the western US, the $\sim$ 80 % enhancement in biomass burned yields an increase in NO emissions, from 0.03 Tga$^{-1}$ in the present day to 0.05 Tga$^{-1}$ in the future climate, and an increase in CO emissions from 1.9 to 3.4 Tga$^{-1}$.

### 3.4 Impacts of wildfire on ozone air quality

Figure 9a shows the simulated daily maximum 8 h average surface ozone (MDA8), averaged over North American in summer (June–July–August, JJA). MDA8 is a metric used by the US Environmental Protection Agency (EPA) to diagnose ozone air quality. We focus on the summer season, when fire activity peaks in both the US and Canada. The figure shows mean MDA8 values of 40–75 ppbv across the US, with the maximum in the East due to the local anthropogenic emissions (Fiore et al., 2002). The concentrations in Alaska and Canada range from 20 to 60 ppbv. However, for most re-
regions north of 55° N, MDA8 is generally less than 40 ppbv. As shown in Fig. 9b, we find that wildfire emissions in these far northern areas contribute 1–10 ppbv to average JJA surface ozone concentrations, with a mean contribution of 4 ppbv. These values are considerably larger than the average 1 ppbv contribution of wildfires to surface ozone that we calculate in the western US (Fig. 9b) because of the much higher biomass burning emission in Alaska. In the eastern US, wildfires make almost no contribution to mean surface ozone in summer.

The increased fire emissions which we calculate at the midcentury results in greater ozone pollution across North America (Fig. 9c). We find a maximum JJA mean perturbation of 22 ppbv along the border between Alaska and Canada, where the largest increase in future area burned is projected (Fig. 7a). In central Canada, the future fire emissions contribute 6–9 ppbv to JJA mean ozone concentrations. For the western US, the fire perturbation for surface ozone is about 2 ppbv, with the largest values of 3–5 ppbv in the Pacific Northwest and Rocky Mountain Forest ecoregions. Relative to the present-day contribution, the fire perturbation at the midcentury enhances JJA mean surface ozone by an additional 5 ppbv in Alaska, 3 ppbv in Canada, and 1 ppbv in the western US (Fig. 9d), indicating a degradation in air quality.

A key question is to what extent boreal fires affect the more populated regions of lower latitudes. In Fig. 10, we investigate the contributions of climate, local and boreal wildfire emissions, and atmospheric transport to JJA mean surface ozone concentrations in the central and western US. Figure 10a shows that all these effects together increase surface ozone in the US by 1–4 ppbv at the midcentury but with large spatial variability. The enhancement in central and southwestern states is mainly associated with climate change (Fig. 10b), which increases temperature-driven soil NO\textsubscript{x} emissions and air mass stagnation (Wu et al., 2008b). In the northwestern coastal states, the impact of these effects is offset by the reduced lifetimes of PAN and ozone in the warmer climate, which diminish the impact of Asian emissions on surface ozone there (Wu et al., 2008b). However, the calculated increase of local wildfire emissions in these coastal states and across the Northwest enhances surface ozone by 1–2 ppbv at mid-
century (Fig. 10c). In the most northern states, this increase is enhanced by another 0.5 ppbv due to transport of pollutants from boreal wildfires (Fig. 10d).

In Fig. 11 we examine the impact of wildfire emissions on the frequency of ozone pollution episodes. In the northwestern US, where the impact of fire emission is especially large (Fig. 10c), surface ozone above the 95th percentile (i.e., on the 5 most polluted days in summer) increases by 2 ppbv at the midcentury (Fig. 11a). Simulations without fire emissions show an increase of 1 ppbv above the same percentile, indicating that the increased wildfire emission alone contributes a 1 ppbv enhancement during ozone pollution episodes in this region. The changes are more significant for Alaska and Canada, where we predict large increases in fire activity (Fig. 9c). As Fig. 11b shows, climate change alone decreases ozone above the 95th percentile ozone by an average $\sim$ 3 ppbv in Alaska, likely because of the effects of enhanced water vapor on background ozone (Wu et al., 2008a). However, when changes in fire emissions are included, the simulation predicts that ozone above the 95th percentile instead increases by 12 ppbv at midcentury, suggesting a positive change of 15 ppbv due to wildfire alone. Over high latitudes in Canada, climate change decreases the 95th percentile ozone by 1 ppbv; however, the inclusion of future fire perturbation enhances it by 4 ppbv (Fig. 11c), indicating that the contribution from wildfire may be as great as 5 ppbv.

4 Discussion and conclusions

We examined the effects of changing wildfire activity in a future climate on June–August MDA8 ozone over the Western US, Canada, and Alaska by the midcentury. We built stepwise regressions between area burned and meteorological variables in 12 boreal ecoregions. These regressions explained 34–75% of the variance in area burned for all ecoregions, with 500 hPa geopotential heights and temperatures the driving factors. With these regressions and future meteorology from an ensemble of climate models, we predicted that the median area burned increases by 150–390% in Alaska and the
western Canadian ecoregions by the midcentury due to enhanced 500 hPa geopotential heights and temperatures. The area burned shows moderate increases of 40–90% in the central and southern Canadian ecoregions, but a 50% decrease in the Taiga Plain, where most of the GCMs predict increases in precipitation at midcentury. Using the GEOS-Chem CTM, we found that fire perturbation at the midcentury enhances summer mean daily maximum 8 h surface ozone by 5 ppbv in Alaska, 3 ppbv in Canada, and 1 ppbv in the western US. The changes in wildfire emissions have larger impacts on pollution episodes, as ozone above the 95th percentile increases by 15 ppbv in Alaska, 5 ppbv in Canada, and 1 ppbv in northwestern US. 

Our study represents the first time that multi-model meteorology has been used to project future area burned in Alaskan and Canadian forest. The individual models in our study predict changes in area burned of different magnitudes or even of opposite sign, but the median values and the spread in model results provide an estimate of both the sign and the uncertainty of these projections. We find the projections are most robust over Alaska and western Canada, where almost all GCMs calculate significant increases in area burned (Fig. 7b; Table 3). For these regions, wildfire activity is largely associated with blocking highs and the resulting hot, dry weather, and both temperature and geopotential height show consistent and significant increases here in all climate models (Fig. 6). However, for northern Canada, where the control of blocking systems on area burned is weaker, we projected a less robust decreasing trend in area burned, due to the competing effects of hotter weather and wetter conditions. The multi-model ensemble approach allows us to identify the most robust changes in the future wildfire activity due to climate change, and as a result should be more reliable than predictions using only 1–2 models, which can yield very different projections especially for northern Canada (e.g., Wotton et al., 2010).

Our approach neglects the impacts of topography, human activity, and fuel changes on wildfire trends. The aggregation method used here for each ecoregion may hide the spatial variation of both area burned and meteorological variables and obscure their relationships (Balshi et al., 2009; Meyn et al., 2010). Changes in fire domain and
climate may lead to changes in forest composition (DeSantis et al., 2011), resulting in different fire severity and spread efficiency (Thompson and Spies, 2009).

For our study, we assumed that fuel load remains constant for 50 years, but we calculated a 9% average increase in fuel consumption in boreal regions. Our assumption of constant fuel load is justified at least for the conterminous US since trends in heavy-fuel load in US forests are likely to be gradual (Hanson and Weltzin, 2000). For boreal regions, recent simulations with DGVMs show that large-scale forest die back may occur in coming decades, due to intense heat and drought (Heyder et al., 2011). In addition, mountain pine beetle outbreaks are important disturbances for both boreal and US forests, leading to changes in fuel load with climatic shifts (Fauria and Johnson, 2009; Jenkins et al., 2014). We did not consider these effects in this study.

Compared with previous studies, our estimate of fuel consumption shows higher values over western Canada (Table 4), where the largest increase in future area burned is predicted (Fig. 7a), suggesting that the boreal fire emissions might be overestimated. However, our estimate of a 9% increase in fuel consumption may, in fact, be conservative. Some DGVM studies predict 30–40% increases in burning severity for US Pacific Northwest forest by the end of the 21st century (Rogers et al., 2011). Moreover, observations have suggested that large area burned sometimes results in burning at greater soil depth than is typical (Turetsky et al., 2011). Thus the projected increase in fire areas may amplify future fuel consumption, leading to even larger emissions than predicted in this study.

The emission of boreal wildfires in our simulation shows limited contributions to ozone concentrations in downwind areas, but causes significant local ozone enhancement in Alaska and Canada. However, observations point to uncertainties in the relationship between wildfire activity and ozone. First, the emission factors of ozone precursors are not well constrained, especially for NO\textsubscript{X}. Using aircraft data from boreal fires, Alvarado et al. (2010) determined an emission factor of 1.1 g NO kg DM\textsuperscript{-1}, lower than our value of 1.6 g NO kg DM\textsuperscript{-1} and much lower than the estimate of 3.0 g NO kg DM\textsuperscript{-1} for extratropical forest fires in Andreae and Merlet (2001). Alvarado et al. (2010) based
their NO$_x$ emission factor on their finding that 40% of wildfire NO$_x$ is rapidly converted to PAN and 20% to HNO$_3$. Second, observations do not consistently reveal ozone enhancements during wildfire events. Jaffe et al. (2008) found a significant correlation between interannual variations of observed surface ozone and area burned in the western US. Using the same ozone dataset, however, Zhang et al. (2014) did not find regional ozone enhancements during wildfire events, when such enhancements would be expected to be large. In their review, Jaffe and Wigder (2012) reported that increased ozone is observed in most plumes, but with huge variability in the ratio of $\Delta$O$_3$/ΔCO. Alvarado et al. (2010), on the other hand, found that only 4 out of 22 plumes showed enhanced ozone. Such discrepancies in plume data may be attributed to differences in plume age (Alvarado et al., 2010), emissions of wildfire NO$_x$ and VOCs (Zhang et al., 2014), or plume photochemistry (Verma et al., 2009; Jiang et al., 2012). Third, the effect of long-range transport of wildfire PAN on ozone downwind is not well known. Observations suggest that PAN forms rapidly in fresh fire plumes and may enhance ozone downwind as it decomposes (Real et al., 2007; Jaffe and Wigder, 2012). In their model study, Fischer et al. (2014) reported a large effect of fires on PAN in the high northern latitudes but limited impacts over the downwind areas in US. As a result, the use of a moderately high NO$_x$ emission factor and omission of PAN formation in our simulations may lead to an overestimate of the fire-induced ozone enhancement, especially for the local areas. In contrast, the coarse model resolution and the use of monthly fire emissions may smooth the regional and daily extremes, resulting in a more conservative estimate of pollution episodes.

Given these limitations, our estimate with a multi-model ensemble consistently shows that wildfire activity will likely increase in North American boreal forest by the midcentury, especially in western Canada and Alaska. Our study suggests that area burned could increase by 130–350% in these two regions, while in central and southern Canada, where most people reside, area burned could increase 40–90%. In north central Canada, the competition between increased temperature and precipitation in the future atmosphere results in uncertainty in the projections for area burned. Overall,
these trends in boreal wildfire activity may amplify the threat of wildfires to Canadian residents, increase the expense of fire suppression, and lead to more ozone pollution both locally and in the central and western US.

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References


Table 1. Summary of simulations in this study.

<table>
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<th>Simulations</th>
<th>Western US fire emissions</th>
<th>Boreal fire emissions</th>
<th>Climate</th>
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<tr>
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</table>

\(^a\) Present-day denotes 1997–2001

\(^b\) Future denotes 2047–2051
### Table 2. Regression fits\textsuperscript{a} for each aggregated ecoregion.

<table>
<thead>
<tr>
<th>Ecoregion</th>
<th>Regressions\textsuperscript{a}</th>
<th>$R^2$</th>
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</thead>
<tbody>
<tr>
<td>Alaska Boreal Interior</td>
<td>$2.2 \times 10^5 T_{\text{max}} \cdot \text{SUM} + 5.7 \times 10^3 \text{HGT} \cdot \text{SUM} - 1.5 \times 10^{-3} \text{ISI}_{\text{max}} - 3.5 \times 10^{-4}$</td>
<td>60%</td>
</tr>
<tr>
<td>Alaska Boreal Cordillera</td>
<td>$5.8 \times 10^3 \text{HGT} \cdot \text{SUM} + 4.8 \times 10^4 T_{\text{max}} \cdot \text{AUT} - 2 + 4.6 \times 10^4 \text{T} \cdot \text{SPR} - 3.3 \times 10^{-4}$</td>
<td>61%</td>
</tr>
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<td>Taiga Cordillera</td>
<td>$5.7 \times 10^4 T_{\text{max}} \cdot \text{ANN} - 2 + 2.8 \times 10^3 \text{HGT} \cdot \text{SUM} - 1.5 \times 10^7$</td>
<td>36%</td>
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<td>Canadian Boreal Cordillera</td>
<td>$7.6 \times 10^3 \text{HGT} \cdot \text{SUM} - 4.2 \times 10^{-7}$</td>
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<tr>
<td>Western Cordillera</td>
<td>$3.5 \times 10^4 T_{\text{max}} \cdot \text{SUM} - 8.3 \times 10^2 \text{HGT} \cdot \text{SPR} + 6.4 \times 10^2 \text{DMC}_{\text{max}} - 3.7 \times 10^{-6}$</td>
<td>53%</td>
</tr>
<tr>
<td>Taiga Plain</td>
<td>$9.8 \times 10^5 \text{ISI} - 5.9 \times 10^5 \text{Prec} \cdot \text{FS} - 1.5 \times 10^6 \text{Prec} \cdot \text{Win} - 4.7 \times 10^3$</td>
<td>75%</td>
</tr>
<tr>
<td>Boreal Plain</td>
<td>$8.8 \times 10^4 \text{DSR}<em>{\text{max}} + 5.1 \times 10^4 \text{RH} \cdot \text{SUM} - 2 + 2.1 \times 10^4 \text{FWI}</em>{\text{max}} - 1 - 4.0 \times 10^6$</td>
<td>52%</td>
</tr>
<tr>
<td>Western Taiga Shield</td>
<td>$1.9 \times 10^5 \text{ISI}_{\text{max}} + 5.7 \times 10^4 \text{RH} \cdot \text{AUT} - 6.0 \times 10^6$</td>
<td>46%</td>
</tr>
<tr>
<td>Eastern Taiga Shield</td>
<td>$5.4 \times 10^4 \text{RH} \cdot \text{WIN} - 2 - 6.2 \times 10^4 \text{RH} \cdot \text{ANN} - 7.7 \times 10^3 \text{DMC}_{\text{max}} - 2 + 1.2 \times 10^6$</td>
<td>38%</td>
</tr>
<tr>
<td>Hudson Plain</td>
<td>$2.4 \times 10^3 \text{HGT} \cdot \text{SUM} - 1.8 \times 10^4 \text{T} \cdot \text{SPR} - 1.6 \times 10^4 T_{\text{max}} \cdot \text{WIN} - 1 - 1.4 \times 10^7$</td>
<td>34%</td>
</tr>
<tr>
<td>Western Mixed Wood Shield</td>
<td>$2.0 \times 10^4 \text{BUI}_{\text{max}} + 8.3 \times 10^3 \text{HGT} \cdot \text{SUM} - 4.7 \times 10^4$</td>
<td>67%</td>
</tr>
<tr>
<td>Eastern Mixed Wood Shield</td>
<td>$-6.7 \times 10^5 \text{RH} \cdot \text{SUM} + 2.8 \times 10^3 \text{HGT} \cdot \text{AUT} - 1 - 1.0 \times 10^7$</td>
<td>43%</td>
</tr>
</tbody>
</table>

\textsuperscript{a} The values (−1) or (−2) after a predictor indicate that the meteorological field is one or two years earlier than current area burned. Variables are $T$ (temperature), $T_{\text{max}}$ (maximum temperature), RH (relative humidity), Prec (precipitation), HGT (geopotential height), and fire indexes from CFWIS, such as Duff Moisture Code (DMC), Build-up Index (BUI), Initial Spread Index (ISI), and Daily Severity Rating (DSR). Meteorological fields are averaged for winter (WIN, DJF), spring (SPR, MAY), summer (SUM, JJA), autumn (AUT, SON), fire season (FS, MJJASO), and the whole year (ANN). The order of the terms indicates their contributions to the $R^2$ in the regression.
Table 3. Observed and projected area burned in boreal ecoregions.

<table>
<thead>
<tr>
<th>Ecoregions</th>
<th>Observed&lt;sup&gt;a&lt;/sup&gt; 1983–1999</th>
<th>Present Day Regression&lt;sup&gt;b&lt;/sup&gt; 1983–1999</th>
<th>Future Regression&lt;sup&gt;b&lt;/sup&gt; 2048–2064</th>
<th>Ratio&lt;sup&gt;c&lt;/sup&gt; (Future/Present)</th>
<th># of models&lt;sup&gt;d&lt;/sup&gt; (&lt;p&gt;&lt;i&gt;0.05&lt;/i&gt;)</th>
<th># of models&lt;sup&gt;e&lt;/sup&gt; (± 30 %)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Alaska Boreal Interior</td>
<td>2.1 ± 3</td>
<td>3.7 ± 2.9</td>
<td>9.7 ± 3.6</td>
<td>2.46</td>
<td>12</td>
<td>6</td>
</tr>
<tr>
<td>Alaska Boreal Cordillera</td>
<td>0.6 ± 1</td>
<td>1.1 ± 1.3</td>
<td>5.3 ± 1.7</td>
<td>4.85</td>
<td>13</td>
<td>10</td>
</tr>
<tr>
<td>Taiga Cordillera</td>
<td>0.9 ± 1.7</td>
<td>0.9 ± 0.8</td>
<td>3.3 ± 0.7</td>
<td>3.26</td>
<td>13</td>
<td>11</td>
</tr>
<tr>
<td>Canadian Boreal Cordillera</td>
<td>1.3 ± 1.3</td>
<td>1.7 ± 1.3</td>
<td>4.5 ± 1.4</td>
<td>2.64</td>
<td>13</td>
<td>13</td>
</tr>
<tr>
<td>Western Cordillera</td>
<td>0.2 ± 0.2</td>
<td>0.3 ± 0.4</td>
<td>0.8 ± 0.4</td>
<td>2.66</td>
<td>11</td>
<td>11</td>
</tr>
<tr>
<td>Taiga Plain</td>
<td>3.8 ± 4.6</td>
<td>2.5 ± 2.7</td>
<td>1.6 ± 1.9</td>
<td>0.48</td>
<td>5</td>
<td>5</td>
</tr>
<tr>
<td>Boreal Plain</td>
<td>2.4 ± 3.5</td>
<td>2.6 ± 2.7</td>
<td>4.7 ± 3.2</td>
<td>1.44</td>
<td>3</td>
<td>8</td>
</tr>
<tr>
<td>Western Taiga Shield</td>
<td>3.7 ± 7.1</td>
<td>4 ± 4.3</td>
<td>4.3 ± 3.3</td>
<td>0.96</td>
<td>0</td>
<td>9</td>
</tr>
<tr>
<td>Eastern Taiga Shield</td>
<td>1.9 ± 4.3</td>
<td>2 ± 1.2</td>
<td>1.6 ± 1.9</td>
<td>1.06</td>
<td>11</td>
<td></td>
</tr>
<tr>
<td>Hudson Plain</td>
<td>1 ± 1.6</td>
<td>0.9 ± 0.5</td>
<td>1 ± 0.5</td>
<td>1.2</td>
<td>2</td>
<td>9</td>
</tr>
<tr>
<td>Western Mixed Wood Shield</td>
<td>6.8 ± 7.4</td>
<td>7.3 ± 4.8</td>
<td>11.5 ± 5.1</td>
<td>1.65</td>
<td>8</td>
<td>9</td>
</tr>
<tr>
<td>Eastern Mixed Wood Shield</td>
<td>1.7 ± 1.8</td>
<td>1.8 ± 1.3</td>
<td>3.3 ± 1.6</td>
<td>1.91</td>
<td>8</td>
<td>8</td>
</tr>
</tbody>
</table>

<sup>a</sup> AB = area burned (10<sup>5</sup> ha<sup>−1</sup>). Results in each ecoregion are shown as AB ± σ. AB is the long-term average of the AB during fire season (May–Oct), and σ is the standard deviation.

<sup>b</sup> Results in each ecoregion are the median values of AB and σ predicted using the meteorological fields from 13 GCMs for the A1B scenario.

<sup>c</sup> Results in each ecoregion represent the median value of the 13 ratios of future AB to present-day AB, calculated with the GCM meteorology.

<sup>d</sup> Number out of 13 models that predict a significant (<i>p < 0.05</i>) increase in AB in each ecoregion, as determined by the Student t test.

<sup>e</sup> Number out of 13 models that predict a ratio within ± 30 % of the median ratio.
Table 4. Fuel consumption in boreal ecoregions, as reported by recent studies.

<table>
<thead>
<tr>
<th></th>
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<th></th>
<th></th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>Alaska Boreal Interior</td>
<td>7.5</td>
<td>N/A</td>
<td>N/A</td>
<td>4.9</td>
<td>5.2</td>
<td>5.5 (4.6)</td>
<td>5.4</td>
<td>5.6</td>
<td>5.6</td>
</tr>
<tr>
<td>Taiga Cordillera</td>
<td>N/A</td>
<td>3.1</td>
<td>N/A</td>
<td>N/A</td>
<td>2.7</td>
<td>3.8 (3.5)</td>
<td>3.6</td>
<td>3.7</td>
<td>3.7</td>
</tr>
<tr>
<td>Can. Boreal Cordillera</td>
<td>5.4</td>
<td>3.2</td>
<td>N/A</td>
<td>7.2</td>
<td>3.5</td>
<td>5.5 (4.7)</td>
<td>5.2</td>
<td>5.2</td>
<td>5.2</td>
</tr>
<tr>
<td>Western Cordillera</td>
<td>N/A</td>
<td>3.9</td>
<td>N/A</td>
<td>N/A</td>
<td>2.7</td>
<td>6.6 (5.9)</td>
<td>6.2</td>
<td>7. 7</td>
<td>7. 7</td>
</tr>
<tr>
<td>Taiga Plain</td>
<td>2.9</td>
<td>2.9</td>
<td>3.5</td>
<td>3.3</td>
<td>5.4</td>
<td>7.2 (6.6)</td>
<td>7.7</td>
<td>8.2</td>
<td>8.2</td>
</tr>
<tr>
<td>Boreal Plain</td>
<td>3.8</td>
<td>2.4</td>
<td>2.8</td>
<td>6.8</td>
<td>2.1</td>
<td>5.6 (5.0)</td>
<td>5.7</td>
<td>5.8</td>
<td>5.8</td>
</tr>
<tr>
<td>W. Taiga Shield</td>
<td>1.</td>
<td>1.9</td>
<td>1.5</td>
<td>1.8</td>
<td>5.3</td>
<td>3.9 (3.9)</td>
<td>4.9</td>
<td>5.4</td>
<td>5.4</td>
</tr>
<tr>
<td>E. Taiga Shield</td>
<td>1.6</td>
<td>1.9</td>
<td>1.7</td>
<td>3.</td>
<td>4.</td>
<td>1.8 (2.2)</td>
<td>2.3</td>
<td>2.8</td>
<td>2.8</td>
</tr>
<tr>
<td>Hudson Plain</td>
<td>1.7</td>
<td>1.9</td>
<td>N/A</td>
<td>2.9</td>
<td>6.7</td>
<td>3.1 (4.1)</td>
<td>3.3</td>
<td>3.8</td>
<td>3.8</td>
</tr>
<tr>
<td>W. Mixed Wood Shield</td>
<td>2.1</td>
<td>2.5</td>
<td>3.</td>
<td>5.7</td>
<td>4.9</td>
<td>6.4 (6.6)</td>
<td>6.4</td>
<td>6.9</td>
<td>6.9</td>
</tr>
<tr>
<td>E. Mixed Wood Shield</td>
<td>2.6</td>
<td>2.</td>
<td>2.4</td>
<td>0.5</td>
<td>2.9</td>
<td>3.0 (4.1)</td>
<td>3.1</td>
<td>3.6</td>
<td>3.6</td>
</tr>
</tbody>
</table>

- Fuel consumption unit is kg DM m\(^{-2}\) burned. For some studies that use units of kg C m\(^{-2}\) burned, we multiply their values by 2 g DM g\(^{-1}\) C. DM denotes dry matter.
- Values are averages of 1959–1995.
- Values are estimated for forest floor fuel consumption in a GCM 1 × CO\(_2\) scenario.
- Values are averages of 1959–2002, estimated with the same burning severity parameters as French et al. (2000) but with modeled vegetation and soil carbon pool.
- Results are the fuel consumption weighted by area burned and drought code (DC) for 1980–2009, using the DC thresholds determined by a single probability distribution for North America. As a comparison, the values calculated with ecoregion-specific DC thresholds are shown in brackets. For PD and A1B, values are calculated using predicted median DC for present day (1996–2001) and midcentury (2046–2051) from the multi-model projection.
Figure 1. Distribution of the 12 ecoregions used for this study. The black triangle symbols indicate the GSOD meteorological data sites in Alaskan and Canadian ecoregions.
Figure 2. Fuel consumption over Alaska and Canada (a) for moderately dry conditions and (b) weighted by the Drought Code (DC) and area burned for 1980–2009. The average values are shown in brackets.
Figure 3. (a) Observed annual area burned and (b) fraction of the variance in observed area burned explained by the regression in each ecoregion for the period of 1980–2009 ($R^2$). The ecoregions are: Alaska Boreal Interior (ABI), Alaska Boreal Cordillera (ABC), Taiga Cordillera (TC), Canadian Boreal Cordillera (CBC), Western Cordillera (WC), Taiga Plain (TP), Boreal Plain (BP), Western Taiga Shield (WTS), Eastern Taiga Shield (ETS), Hudson Plain (HP), Western Mixed Wood Shield (WS), and Eastern Mixed Wood Shield (ES).
Figure 4. Observed (red solid lines) and predicted (blue dashed lines) area burned ($10^5$ ha) for 1980–2009 in boreal ecoregions. The area burned is calculated using the regressions for the fire season (May–October) for each ecoregion. Site-based observations from GSOD are used in the prediction. The fraction of the variance in observed area burned explained by the regression ($R^2$) is shown on each panel.
Figure 5. (a) Ratios of modeled to observed area burned for 1983–1999 and (b) the ratios of midcentury (2048–2064) to the present-day (1983–1999) area burned, as projected by an ensemble of GCMs. The ecoregions are: Alaska Boreal Interior (ABI), Alaska Boreal Cordillera (ABC), Taiga Cordillera (TC), Canadian Boreal Cordillera (CBC), Western Cordillera (WC), Taiga Plain (TP), Boreal Plain (BP), Western Taiga Shield (WTS), Eastern Taiga Shield (ETS), Hudson Plain (HP), Western Mixed Wood Shield (WS), and Eastern Mixed Wood Shield (ES). Different symbols are used for each model. The black bold lines indicate the median ratios. Note the difference in scale between the two panels.
Figure 6. Calculated changes in (a) surface air temperature, (b) precipitation, and (c) geopotential height at 500 hPa during the fire season (May–October) in 2048–2064 relative to 1983–1999. Results are from an ensemble of GCMs for the A1B scenario. The ecoregions are: Alaska Boreal Interior (ABI), Alaska Boreal Cordillera (ABC), Taiga Cordillera (TC), Canadian Boreal Cordillera (CBC), Western Cordillera (WC), Taiga Plain (TP), Boreal Plain (BP), Western Taiga Shield (WTS), Eastern Taiga Shield (ETS), Hudson Plain (HP), Western Mixed Wood Shield (WS), and Eastern Mixed Wood Shield (ES). Different symbols are used for each model. The black bold lines indicate the median changes.
Figure 7. (a) Median ratios of midcentury (2048–2064) to present day (1983–1999) area burned in each boreal ecoregions, as predicted by an ensemble of GCMs and (b) the number of GCMs out of 13 total which predict significant changes of the same sign as the median. The ecoregions are: Alaska Boreal Interior (ABI), Alaska Boreal Cordillera (ABC), Taiga Cordillera (TC), Canadian Boreal Cordillera (CBC), Western Cordillera (WC), Taiga Plain (TP), Boreal Plain (BP), Western Taiga Shield (WTS), Eastern Taiga Shield (ETS), Hudson Plain (HP), Western Mixed Wood Shield (WS), and Eastern Mixed Wood Shield (ES).
Figure 8. Biomass burning (BB) in Alaska and Canada in terms of dry matter (DM) burned per year, calculated as the product of area burned and fuel consumption. Panel (a) shows values based on observations for 1980–2009, (b) the predicted values for 1996–2001, and (c) the projections for 2046–2051. The differences between midcentury and present day (c–b) are shown in (d). Annual mean values summed over the whole domain are shown in brackets. Units: Tg DM a⁻¹.
Figure 9. (a) Simulated present-day MDA8 ozone at the surface in summer (June–August). Panel (b) shows the contribution to MDA8 summertime ozone by wildfire emissions in the present day (FULL_PD – NOFIRE_P), and panel (c) shows the same contribution, but at midcentury (FULL_A1B – NOFIRE_A1B). Panel (d) presents the change in the contribution of wildfires to MDA8 ozone between the two periods (i.e., c–b). Descriptions of the sensitivity simulations are given in Table 1. The color scale saturates at both ends.
Figure 10. (a) Simulated changes in MDA8 ozone at the surface in summer (June–August) at the midcentury relative to the present day (FULL_A1B – FULL_PD) over the western and central United States. The other three panels show the contributions to the changes in panel (a) from (b) climate change (CLIM_CHAN – FULL_PD), (c) changes in fire emissions in the western US (FULL_A1B – BOREAL_FIRE) and (d) changes in fire emissions in Alaska and Canada (FULL_A1B – WUS_FIRE). Descriptions of the sensitivity simulations are given in Table 1.
Figure 11. Simulated cumulative probability distributions of MDA8 ozone at the surface in summer (June–August) over (a) northwestern US (> 40° N), (b) Alaska, and (c) Canada (> 55° N) for different scenarios. Black shows the present-day (1997–2001) climate without wildfire emissions; green shows future (2047–2051) climate without wildfire emissions; blue indicates present-day climate including the associated wildfire emissions; and red indicates future climate including the associated wildfire emissions. Each point represents the value in one grid square within each region for each day during the five model summers (1997–2001 or 2047–2051).