

Stjern Camilla Weum (Orcid ID: 0000-0003-3608-9468)
Samset Bjørn Hallvard (Orcid ID: 0000-0001-8013-1833)
Myhre Gunnar (Orcid ID: 0000-0002-4309-476X)
Forster Piers (Orcid ID: 0000-0002-6078-0171)
Andrews Timothy (Orcid ID: 0000-0002-8248-8753)
Boucher Olivier (Orcid ID: 0000-0003-2328-5769)
Faluvegi Greg (Orcid ID: 0000-0001-9011-3663)
Iversen Trond (Orcid ID: 0000-0001-6875-2979)
Kasoar Matthew (Orcid ID: 0000-0001-5571-8843)
Kharin Viatcheslav V. (Orcid ID: 0000-0002-3439-9609)
Lamarque Jean-Francois (Orcid ID: 0000-0002-4225-5074)
Richardson Thomas Benjamin (Orcid ID: 0000-0002-3947-9915)
Sand Maria (Orcid ID: 0000-0003-0256-7468)
Shawki Dilshad (Orcid ID: 0000-0002-0013-5955)
Shindell Drew T. (Orcid ID: 0000-0003-1552-4715)
Smith Christopher J (Orcid ID: 0000-0003-0599-4633)
Takemura Toshihiko (Orcid ID: 0000-0002-2859-6067)
Voulgarakis Apostolos (Orcid ID: 0000-0002-6656-4437)

Arctic amplification response to individual climate drivers

Camilla Weum Stjern¹, Marianne Tronstad Lund¹, Bjørn Hallvard Samset¹, Gunnar Myhre¹, Piers M. Forster², Timothy Andrews³, Olivier Boucher⁴, Gregory Faluvegi^{5,13}, Dagmar Fläschner⁶, Trond Iversen⁷, Matthew Kasoar⁸, Viatcheslav Kharin⁹, Alf Kirkevåg⁷, Jean-François Lamarque¹⁰, Dirk Olivié⁷, Thomas Richardson², Maria Sand¹, Dilshad Shawki⁸, Drew Shindell¹¹, Christopher J. Smith², Toshihiko Takemura¹², Apostolos Voulgarakis⁸

¹ CICERO Center for International Climate and Environmental Research, Oslo, Norway

² University of Leeds, Leeds, United Kingdom

³ Met Office Hadley Centre, Exeter, UK

⁴ Institut Pierre-Simon Laplace, Univ. P et M. Curie / CNRS, Paris, France

⁵ Center for Climate Systems Research, Columbia University, New York, USA

⁶ Max Planck Institute for Meteorology, Hamburg, Germany

⁷ Norwegian Meteorological Institute, Oslo, Norway

⁸ Imperial College London, London, United Kingdom

⁹ Canadian Centre for Climate Modelling and Analysis, Victoria, BC, Canada

¹⁰ NCAR/UCAR, Boulder, USA

¹¹ Duke University, Durham, NC, USA

¹² Kyushu University, Fukuoka, Japan

¹³ NASA Goddard Institute for Space Studies, New York, USA

Corresponding author: Camilla Weum Stjern (camilla.stjern@cicero.uio.no)

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Abstract

The Arctic is experiencing rapid climate change in response to changes in greenhouse gases, aerosols and other climate drivers. Emission changes in general, as well as geographical shifts in emissions and transport pathways of short-lived climate forcers, make it necessary to understand the influence of each climate driver on the Arctic. In the Precipitation Driver Response Model Intercomparison Project, ten global climate models perturbed five different climate drivers separately (CO_2 , CH_4 , the solar constant, BC and SO_4). We show that the annual mean Arctic amplification (defined as the ratio between Arctic and the global mean temperature change) at the surface is similar between climate drivers, ranging from 1.9 (\pm an inter-model standard deviation of 0.4) for the solar to 2.3 (\pm 0.6) for the SO_4 perturbations, with minimum amplification in the summer for all drivers. The vertical and seasonal temperature response patterns indicate that the Arctic is warmed through similar mechanisms for all climate drivers except BC. For all drivers, the precipitation change per degree global temperature change is positive in the Arctic, with a seasonality following that of the Arctic amplification. We find indications that SO_4 perturbations produce a slightly stronger precipitation response than the other drivers, particularly compared to CO_2 .

Key Points:

- Arctic amplification of surface warming is similar between global drivers of climate change
- Black carbon induces differing vertical and seasonal amplification patterns
- Sulfate affects Arctic precipitation responses more strongly than other drivers, particularly in the summer season

1 Introduction

Both observations and climate models indicate that the temperature increase in the Arctic has been substantially larger than the globally averaged warming since preindustrial times (Cowtan and Way, 2014; Hartmann et al., 2013; Holland and Bitz, 2003). This phenomenon is known as Arctic amplification (AA). Locally, the Arctic amplification involves pronounced reductions in sea ice cover (AMAP, 2017; Kay et al., 2011; Notz and Stroeve, 2016), increasing Arctic methane emissions from permafrost degradation (Anthony et al., 2016), changes to the mass balance of glaciers and ice sheets (Shepherd et al., 2012) and detrimental influences on Arctic ecosystems (Hinzman et al., 2013) and indigenous peoples (IPCC, 2014). Future changes in shipping and other anthropogenic activities in the Arctic (Peters et al., 2011) can induce further climate change and health impacts in this sensitive region. The strong Arctic warming can also have far-reaching global implications (Zappa et al., 2018). For instance, there are signs that enhanced melting of Arctic sea ice and glaciers influences the Atlantic overturning circulation (Caesar et al., 2018; Yang et al., 2016). Cohen et al. (2014) and Coumou et al. (2018) point out that the AA can influence mid-latitude weather via changes in storm tracks, the jet stream, and planetary waves and their associated energy propagation. Similarly, teleconnections between the Arctic and the Asian continent have been demonstrated through influences on the winter monsoon (Gong et al., 2001) and on the frequency of unusually hard winters (Kug et al., 2015) in East Asia.

The AA is caused by local forcings and feedbacks as well as by remote forcing, but the relative importance of these impacts is still debated. While previous investigations commonly attributed the main cause of AA to the snow and ice albedo feedbacks (Hall, 2004; Manabe and Stouffer, 1980), an idealized model experiment by Alexeev et al. (2005), in which the surface albedo was fixed, demonstrated that poleward energy transport from lower latitudes alone could create an AA. The importance of lower-latitude impacts on the Arctic through transport of heat and moisture has been stressed by many studies (Graversen and Wang, 2009; Screen et al., 2012; Yoshimori et al., 2017), and a recent analysis by Yoshimori et al. (2017) found that the effect of remote forcing is predominant in the Arctic warming. Other studies report that the AA is dominated by local forcings and feedbacks (Stuecker et al. 2016), and point to the importance of processes such as the water vapor and cloud feedbacks (Graversen and Wang, 2009) or the lapse-rate and Planck feedbacks (Pithan and Mauritsen, 2014), in addition to the ice albedo feedback (Laîné et al., 2016; Screen and Simmonds, 2010). Whether or not local processes dominate, the possibility of an influence from forcing at lower latitudes remain. This makes the origin of this forcing relevant, as homogeneously distributed greenhouse gases and highly regional aerosols may influence the AA differently.

While several multi-model studies based on CMIP5 models have focused on the Arctic (Barnes and Polvani, 2015; Clara and Dirk, 2017; Franzke et al., 2017; Huang et al., 2017), none have, to our knowledge, looked at different drivers independently. The Precipitation

Driver Response Model Intercomparison Project (PDRMIP) (Myhre et al., 2017) provides a unique dataset that allows for investigations into climate responses to separate and clearly defined climate drivers, such as greenhouse gases or aerosols, in a multi-model framework. This has led the way for several studies that have investigated various aspects of climate driver-response relationships from a global perspective (Liu et al., 2018; Samset et al., 2018; Samset et al., 2016; Stjern et al., 2017). Here we use the PDRMIP data set to analyze the climate response in the Arctic (defined as the region north of 60°N). Our main objective is to see whether clear differences exist with respect to how different climate drivers influence the Arctic. We do this by providing descriptive accounts of how changes to precipitation and the Arctic amplification vary between the drivers. Our main focus is on experiments where the drivers are perturbed globally, but we also examine the response to isolated perturbations of aerosols in two different source regions. For aerosols and other short-lived climate drivers, the location of emissions can be important to the subsequent impact on global and regional climate (Lund et al., 2014; Sand et al., 2013; Sand et al., 2016; Screen et al., 2012; Shindell et al., 2008; Yang et al., 2018). Wobus et al. (2016) find that different future emission scenarios produce large variations in the resulting Arctic temperature change. Hence, as future changes in emissions and transport pathways (Jiao and Flanner, 2016) may shift the role of drivers, and their sources, in Arctic climate change, quantifying regional dependence of the forcing and response relationship is important. Combined, this will improve our understanding of the processes governing Arctic climate change today and allow us to make more informed predictions for future changes under various emission pathway scenarios.

The next section gives an account of the data and methodology used in the present analyses. Section 3 shows spatially and temporally resolved Arctic temperature responses to global perturbations of five climate drivers, including a quantification of the slow temperature-driven feedback response. A similar account of precipitation responses is given in Section 4, including an energy budget analysis of differences between drivers. A separate analysis of Arctic responses to regional aerosol perturbations is presented in Section 5. Results are discussed in Section 6 and summarized in Section 7.

2 Methods

The PDRMIP dataset is described in detail by Myhre et al. (2017). Briefly, ten global climate models performed experiments where five different climate drivers were perturbed separately. Relative to a baseline, the perturbations involved a doubling of CO₂ concentrations (CO₂x2), a tripling of CH₄ concentrations (CH₄x3), a two percent increase in the solar constant (SOL), a tenfold increase in BC concentrations or emissions (BCx10), and a fivefold increase in SO₄ concentrations or emissions (SO₄x5), respectively. In addition, seven of the models (see Table S1) performed three regional experiments, which simulated the effects of a tenfold increase in BC over Asia (BCx10asia) and in SO₄ over Asia and Europe

(SO₄x10asia and SO₄x10europe). Here, Asia was defined as 10-50°N and 60-140°E, and Europe as 35-70°N and 10-40°E. These results will be described in Section 5.

All perturbations were instantaneous, and simulations were performed in two modes for both the baseline and perturbed states;

- 1) A fully coupled mode, for which equilibrium simulations were run for 100 years. Years 51-100 were used in our analyses. These simulations give the total climate response.
- 2) A set-up with fixed sea surface temperatures (fixed-SST), which were run for 15 years. Years 6 to 15 were used in the analyses. These simulations give the fast (on the order of days to months) climate response.

The baseline represents present-day conditions focused at the beginning of this century (except HadGEM2-ES, which used preindustrial conditions), but the exact year and emission dataset used varies slightly between models (see Table S1). For the aerosol experiments, five of the models (Table S1) performed concentration-driven simulations using identical aerosol concentrations based on the year 2000 monthly multi-model mean of the models in AeroCom Phase II (Myhre et al., 2013; Samset et al., 2013) as a baseline. Of the remaining models, one was not able to perform aerosol perturbations at all, while four used present-day aerosol emissions as a baseline. In the emission-driven models, responses to the increase in aerosols (e.g., increased precipitation) can feed back on the aerosol concentrations (e.g., increased wet removal of the aerosols) and potentially dampen or amplify the change. Sand et al. (2015) found in a comparison of emission-driven and concentration-driven BC perturbation experiments that Arctic responses were stronger in the emission-driven simulations. These model differences will therefore be kept in mind in the subsequent analyses.

The ten models represent a selection of Earth System Models with a broad variation in schemes and formulations. Three of the models include a representation of the radiative impact of BC on snow (Table S1). Furthermore, microphysical aerosol interactions with clouds (aerosol indirect effects) is accounted for in eight models (three of these also include an indirect effect of BC).

The fixed-SST simulations were utilized to calculate effective radiative forcing (ERF) as recommended in Forster et al. (2016), i.e., the change in net radiative long-wave (LW) plus short-wave (SW) fluxes at the top of atmosphere (TOA). As indicated above, the fixed-SST simulations are also used to study the fast (on the order of days to months) response of the climate system to the initial perturbation, also referred to as rapid adjustments. The total climate response, often (but not always) dominated by the surface-temperature change, is derived from the coupled simulations. These two sets of simulations then allow us also to extract the pure surface temperature-driven feedback response (also referred to as the slow response) as the total response minus the fast response.

3 Arctic temperature responses to global perturbations

In this section, we document and discuss the Arctic temperature responses to the global perturbations. We cover annual mean and seasonally resolved surface responses, the separation between rapid responses and global surface temperature driven changes, and the vertical structure of the temperature response. We then conclude the section with a discussion on inter-model differences due to the inclusion of radiative effects of BC on snow in some of the models.

3.1 Annually and seasonally resolved temperature responses

As shown in previous PDRMIP studies (Myhre et al., 2017; Samset et al., 2016; Stjern et al., 2017), the different perturbations cause very dissimilar responses in global mean temperature and ERF, with model mean warming (and ERF) varying from 2.45 K (3.78 Wm^{-2}) for CO₂x2 to 0.61 K (0.66 Wm^{-2}) for BCx10. To facilitate comparison between the different climate drivers, Fig. 1 shows changes per degree global temperature change in Arctic surface temperature (the AA) and precipitation for each given model and experiment. Hatched bars indicate the regional aerosol perturbation experiments, which will be discussed in Section 5. For absolute changes for individual models the reader is referred to Fig. S1. Note here and throughout the paper that an increase in SO₄ gives a positive normalized temperature (and precipitation) change since the negative response it causes in the Arctic is normalized by a negative global temperature change. The horizontal grey line shows the average AA of the five global perturbations.

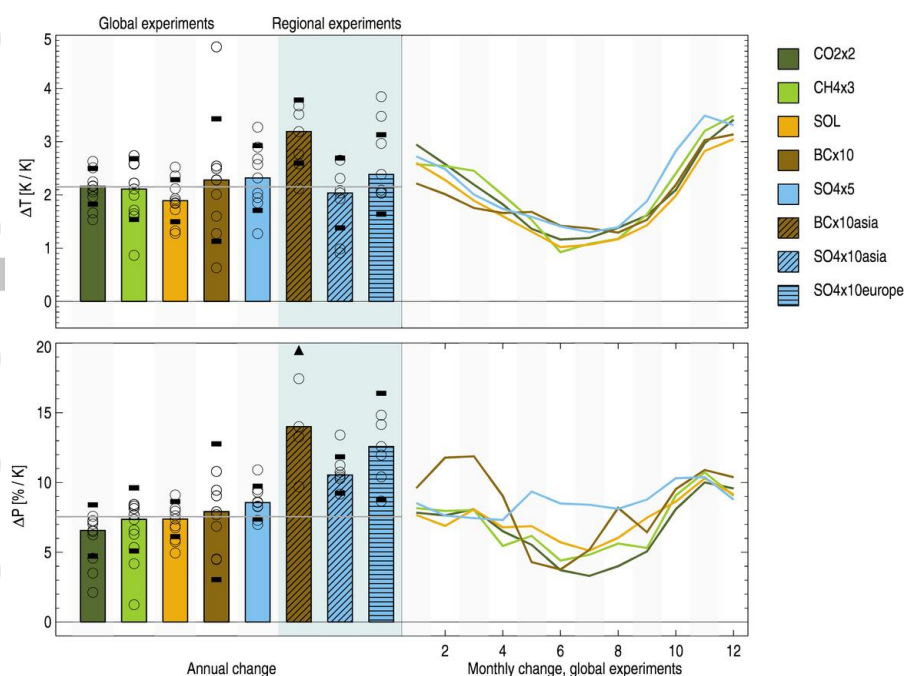


Figure 1. Responses in temperature and precipitation (normalized by the global mean temperature change) to perturbations in the five climate drivers, averaged over the Arctic (all area north of 60° N). Bars show annual model median values, individual model values are shown as circles. The mean of the model-median values for the global cases is shown as a thick grey line. Horizontal bars indicate \pm one

standard deviation, and triangles are used to indicate when standard deviations are outside the range of the axes limits. Graphs show monthly mean changes (for the global experiments only).

We find that the AA is not significantly different (at the 95 % confidence level, by using a Student's t-test) between any of the global experiments. The multi-model median AA ranges from 1.9 (SOL; \pm an inter-model standard deviation of 0.4) to 2.3 (SO₄x5; \pm 0.6) between the drivers. The horizontal grey line shows the average AA of the five global perturbations, and none of the solid bars deviate much from this line. Note, however, that the inter-model spread is substantially larger than the differences in model median AA between experiments. This inter-model range is partly due to the large spread in Arctic responses, but also due to large model differences in the global response. For instance, the outlier in the upper panel of Fig. 1 is CESM-CAM5, for which a very small global response to the BCx10 experiment produces an anomalously high AA. CESM-CAM5's weak temperature response is linked to the model's short BC lifetime for the BCx10 experiment (Stjern et al., 2017). With a few exceptions, we find that models with an above-average AA for one driver tends to have high values for all drivers (and vice versa). This means that the tendency of similar AA regardless of driver is found both for the multi-model medians and for individual models.

The AA is smallest during summer for all the climate drivers (Fig. 1, upper rightmost panel). That amplification is weakest during Arctic summer is well documented elsewhere: In studies based on realistic historical or future scenarios with emission changes in both greenhouse gases and aerosols (Acosta Navarro et al., 2016; Laîné et al., 2016; Screen et al., 2012), in single-perturbation experiments for e.g. SO₄ (Yang et al., 2018) or CO₂ (Yoshimori et al., 2017), and in observations (Graversen et al., 2008). The positive ice albedo feedback is strongest in spring and summer, but is counteracted by strong energy uptake by the Arctic ocean (Laîné et al., 2016), as well as by an increase in evaporation (Acosta Navarro et al., 2016; Laîné et al., 2016) and potentially also cloud responses (Crook et al., 2011; Yoshimori et al., 2017). During winter, on the other hand, the warming from heat release by the ocean is strengthened by positive cloud and lapse rate feedbacks (Laîné et al., 2016; Sejas et al., 2014). The dominating mechanisms contributing to this seasonal cycle differ between land and ocean surfaces, with the tendency for a stronger seasonal AA cycle over oceans (Laîné et al., 2016). Here, we confirm that this seasonality is not a greenhouse gas-specific response but is in fact climate driver independent. To study the land-/ocean-contrast in seasonality, we have repeated the analysis behind Fig. 1 for Arctic land-only and ocean-only responses. Figure S2 shows that the annually averaged AA are consistently stronger over ocean than over land, supporting Laîné et al. (2016). This is true for all drivers. We note that the land/ocean difference can partly be an effect of latitude: the higher Arctic (where AA is found to be stronger, see Fig. 2) consists mostly of oceans while lower latitudes include much more land. Interestingly, over Arctic land surfaces there seems to be a secondary maximum in spring/early summer for BCx10. This is potentially related to the inclusion of the radiative effect of BC on snow (which is strongest in this season) in some models – see

Section 3.4 for further discussion.

3.2 Separating Arctic amplification into rapid responses and feedbacks from global temperature change

A key methodology in PDRMIP is to separate the climate response to an instantaneous perturbation into rapid responses (diagnosed from fixed sea-surface temperature simulations), and the slow feedback response to a change in global mean surface temperature. In Figure S3, we perform this separation for the AA shown in Fig. 1.

Overall, we find weak rapid response contributions to AA. However, when removing the fast response to isolate the global mean temperature induced AA (the slow feedback response), the driver differences seen in Fig. 1 are reduced (Fig. S3). While the standard deviation of the multi-model median AA between these four drivers is 0.16 for the total AA (Fig. 1), the feedback response has a much smaller standard deviation of 0.03. SO_4 , on the other hand, which had a non-significantly stronger AA than the other drivers for the total response, has an AA feedback response that is 40 % stronger than the other drivers, and the difference is now statistically significant.

The only driver to show a rapid adjustment AA significantly different from zero is CO_2 . We interpret this to be due to its atmospheric absorption, which will act to heat the atmospheric column. The other driver with significant absorption, BC, has relatively little change in the Arctic, as discussed below.

Recently, Huneus et al. (2014) compared climate effects of perturbations to CO_2 and the solar constant. They found that particularly for the feedback response, the similarity was remarkable, while rapid adjustments introduced more dissimilar responses. Overall, our results are consistent with this finding.

3.3 Geographical temperature response patterns

In Figure 2, we show the multi-model geographical pattern of AA. Overall the different drivers produce qualitatively similar patterns, with a maximum around the northernmost parts of the Barents Sea, corresponding to the region where all climate drivers produce the greatest loss in sea ice (Fig. S4). It should be noted that the location of this maximum warming may – at least partly – be an artifact of the tendency for models to have a low temperature bias in this region. For instance, Chapman and Walsh (2007) found that 12 out of 14 models have a cold temperature bias in the Barents sea due to too extensive sea ice, and that the same models showed strongest projected temperature changes in this region. Still, the main point – that warming patterns are similar between drivers – remains. Maps of ERF demonstrate that the radiative forcing for all drivers is larger at the lower latitudes than in the Arctic, meaning that the regional ERF pattern is not an important contribution to the

AA (Fig. 3). This is also true when we look at the pattern of ERF at the surface (not shown), which might be more indicative of regional responses than the TOA forcing. The similarity in AA patterns and the decoupling of the AA from the original forcing pattern further imply that the AA is not driver specific.

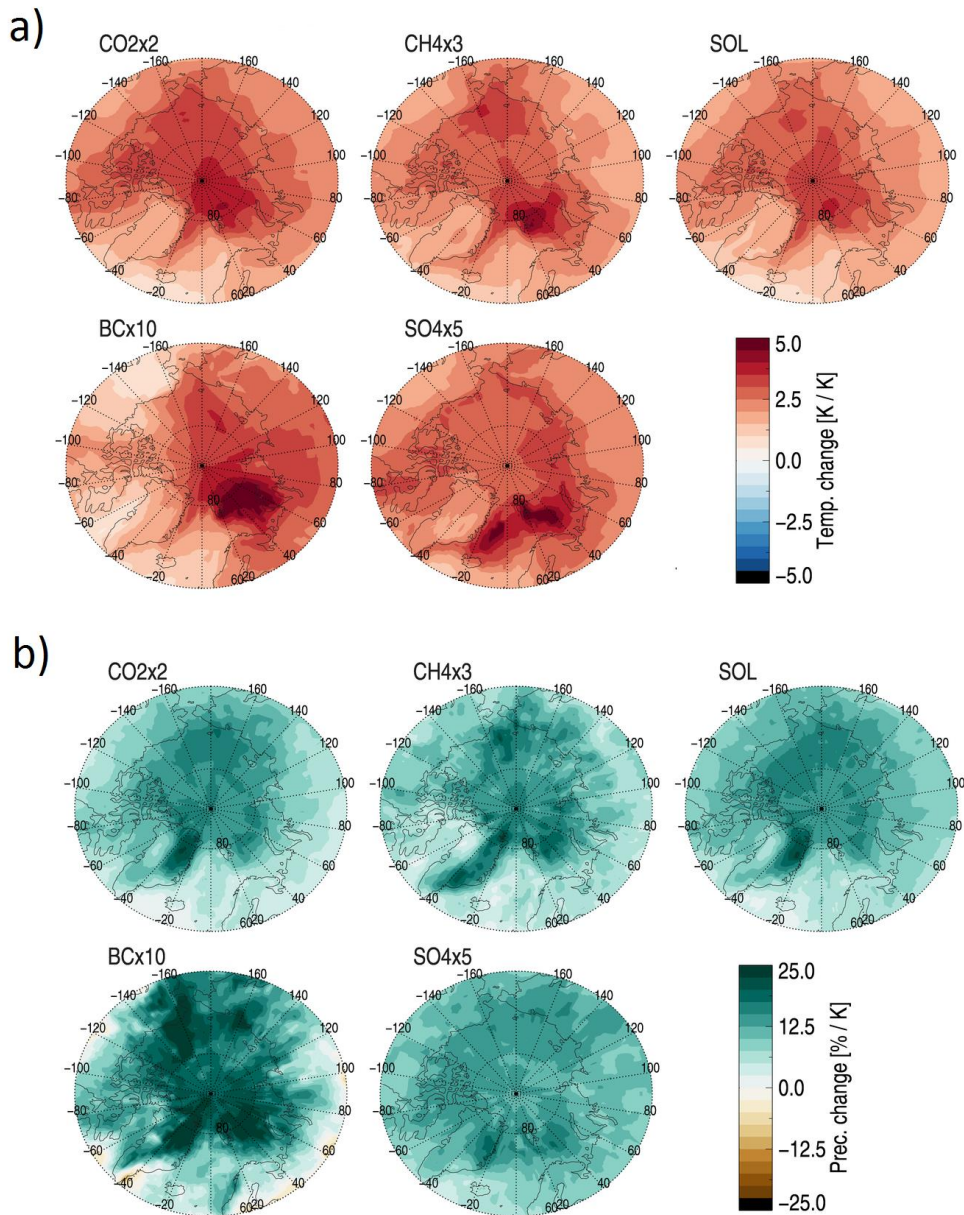


Figure 2. Multi-model median spatial patterns of a) AA and b) precipitation changes normalized by global mean temperature change in the Arctic region, for each of the global PDRMIP experiments.

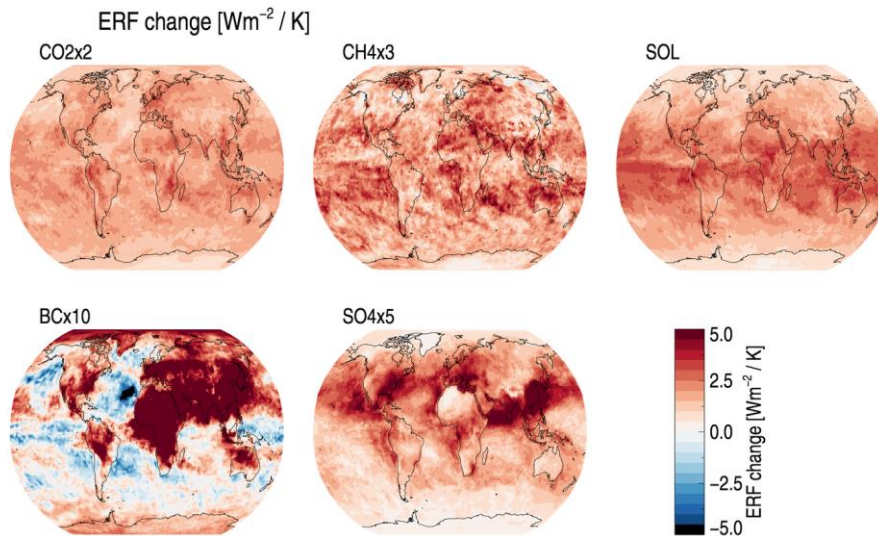


Figure 3. Multi-model median change in ERF, normalized by the global mean temperature change, for the global PDRMIP experiments.

3.4 Vertical structure of the Arctic amplification

In Figure 4, we show the vertical structure of the Arctic amplification. Some studies have suggested that the vertical dimension can provide further insights into the processes driving surface heating in the Arctic region. Sea ice or sea surface temperature changes are typically found to induce strongest response near the Arctic surface (Screen et al., 2012), while changes driven by processes such as changes in poleward heat transport or concentrations of water vapor or clouds will have a maximum response in the mid-troposphere (Chung and Räisänen, 2011; Graversen and Wang, 2009; Yang et al., 2010). Note, however, that local responses and feedbacks will complicate such a clear interpretation.

As seen from Fig. 4a, increases in CO_2 , CH_4 , the solar constant and SO_4 cause very similar vertical patterns in the Arctic, with a strong surface AA and an upper-level (around 200 hPa) negative AA, visible as areas over Arctic latitudes with blue shades in the figure. This upper tropospheric cooling is a typical sign of a warming climate and an associated lifting of the tropopause, which causes temperatures to drop at pressure levels that used to be a part of the warmer stratosphere in the baseline simulations, but are included in the (colder) upper troposphere in the perturbed climate (see e.g. Gillett et al., 2003). This is not to be confused with the well-known stratospheric cooling following increased CO_2 (Ramanathan and Dickinson, 1979), which is also visible in Fig. 2a for $\text{CO}_2 \times 2$, but at higher altitudes. The upper tropospheric cooling is not visible for BC in the annual mean, but in Fig. 4b, where we show monthly mean Arctic changes, we find that BC causes a similar upper tropospheric cooling as the other climate drivers during January through March.

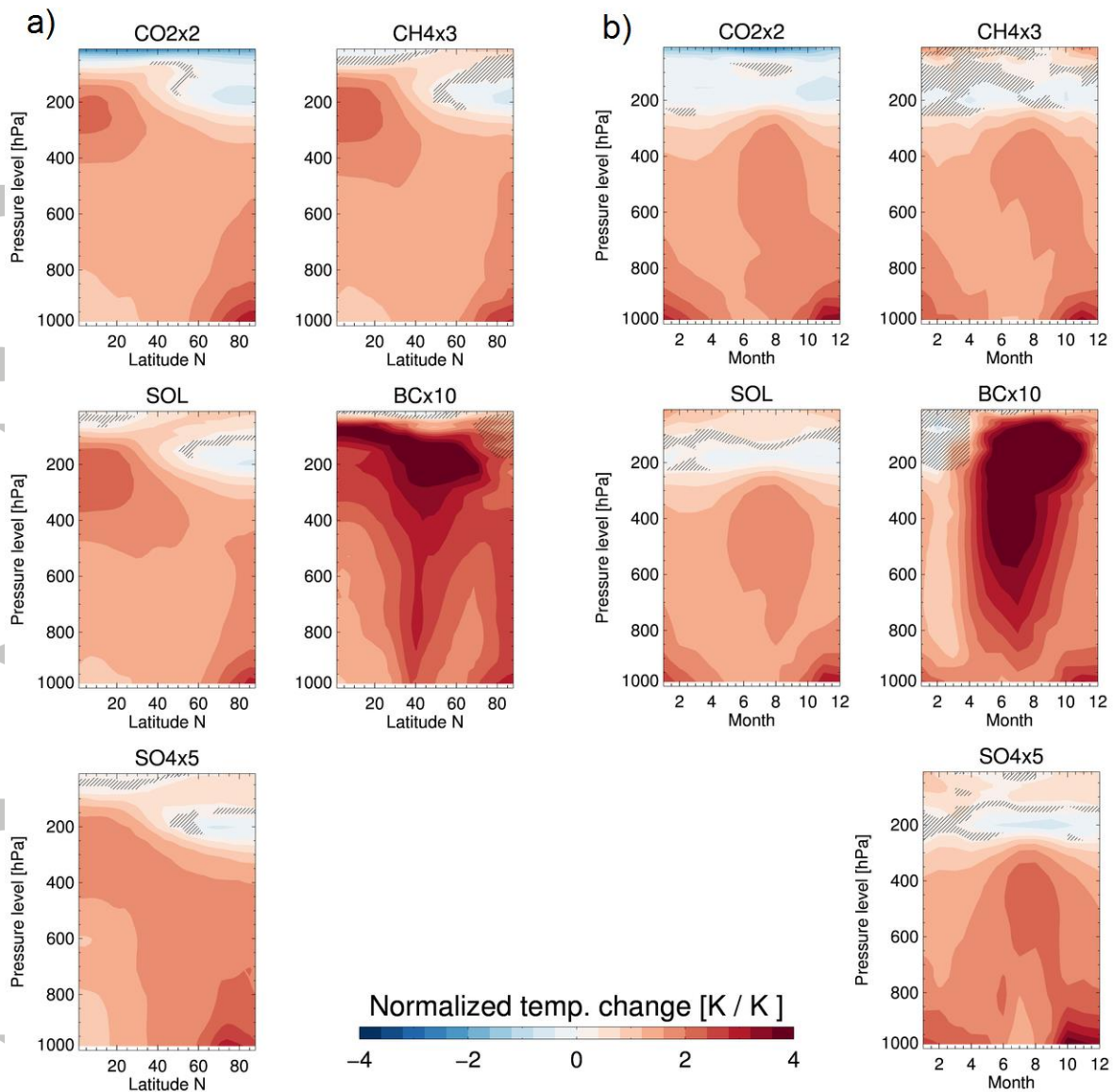


Figure 4. Multi-model median vertically resolved change in Arctic amplification for a) the annual mean for latitudes 0 to 90°N and for b) the Arctic (60 to 90°N) average for each month of the year. Hatching indicates where fewer than 75 % of the models agree on the sign of the change.

For the rest of the year, however, the warming from BC extends throughout the atmospheric column. BC is known to induce efficient local heating of the atmosphere, and has an absorption efficiency that increases with height (Samset and Myhre, 2011). In the Arctic, the atmospheric warming from BC absorption is particularly strong due to the high surface albedo. The warming pattern is therefore linked to the spatiotemporal distribution of BC concentrations. Figure S5a shows the BC x10 mass mixing ratio that the concentration-based models use as input. The largest BC concentration changes occur, as expected, at the surface around 40°N, where emissions are largest. At these latitudes, air masses are effectively lifted by convection, resulting in elevated concentrations also at higher altitudes, where aerosols are transported northwards. This transport is most effective during summer

and fall, as seen from the Arctic BC concentration changes in Figure S5b and the corresponding warming in Fig. 4b. The strong decline in available SW radiation towards fall and winter also means that the radiative effect of increasing concentrations (i.e., absorption of SW and consequent atmospheric warming) will be highest in the summer. At lower altitudes, the polar dome limits the meridional transport into the Arctic, in particular during winter (Stohl, 2006). In addition, the aerosols that are lofted at lower latitudes give strong high-altitude warming, as illustrated in Fig. S5c, where the seasonal warming averaged over East Asia, the major source region (Fig. S5a), for the global BCx10 experiment is shown. This heat (and the associated increase in water vapor) therefore does not need to be elevated from the surface through convection but can be transported directly poleward.

Comparing the AA profile patterns between 0 to 90°N for winter and summer months separately (Fig. S6), we find that the warming pattern for BCx10 has a much stronger poleward extension in summer, consistently with the strong within-Arctic summertime warming around 500 hPa. Graversen et al. (2008) found a similar upper-level summertime amplification in reanalysis data. They pointed out that the source of this high-altitude AA could not be surface driven (i.e., snow and ice feedbacks) and attributed it instead to summertime changes in the northward energy transport. This upper-level AA is however not visible in all reanalysis data sets (Bitz and Fu, 2008; Grant et al., 2008). While most pronounced for BC, the pattern of upper-level summertime warming is present for all climate drivers in our simulations. Dedicated studies are needed to further quantify the role of northward heat transport to the consistent heating around 500 hPa.

For all drivers, the surface AA is lowest in summer. Previous studies have found that one reason for this is that (surface) warming from the ice albedo effect can be partially cancelled by an increase in summertime cloud cover (Crook et al., 2011). However, while an increase in summertime cloud cover is an expected response to the reduced sea ice cover in climate models (Vavrus, 2004; Yoshimori et al., 2014), observations indicate that the summer is the only season which lacks a cloud-response to sea ice changes (Morrison et al., 2018), and that the cloud response tends to be strongest in fall (Kay and Gettelman, 2009). Here, we find that while the sea ice reductions are similar between the drivers (Fig. S4), changes in cloud amounts vary (Fig. 5), with no general tendency of a summertime increase for any of the drivers except BC. BCx10 causes a 5 % increase in summertime low cloud cover (defined as the vertical average of cloud fractions between the surface and 680 hPa), contrasting a very weak increase or (for CO₂x2) a reduction for the other climate drivers. The BC-induced increase in low level clouds would act both to dampen the summertime surface warming, and (by reflecting incoming SW back towards the aerosol layer and adding to the amount of SW available for absorption) enhance the strong upper-level warming. Note also that around 500 hPa, all climate drivers cause a reduction in clouds, consistent with the upper-level summertime warming (Fig. 4b). This response is strongest for BC, for which the warming is greatest.

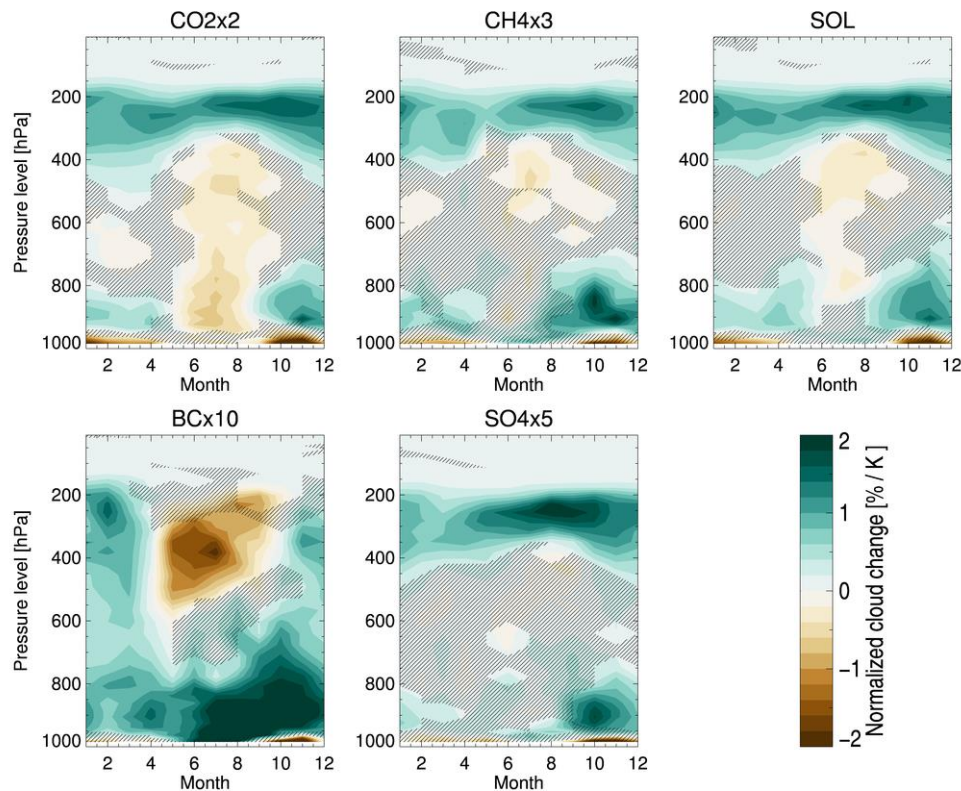


Figure 5. Multi-model mean vertically resolved cloud cover change [%] for the Arctic (60 to 90°N) average for each month of the year, normalized by global mean temperature change.

In the wintertime, there is an upward shift the lower-most clouds in response to all drivers, conceivably associated with an increase in specific humidity and therefore the lifting condensation level, as the atmosphere warms. A similar Arctic cloud response was found in a modelling study by (Gillett et al., 2003) as a response to CO₂ doubling. Li et al. (2014) noted a link between reduced Arctic boundary layer stability and a deepening of the boundary layer in satellite data. The same features are found in our results during winter, seen as a thickening of the low-level cloud layer. This change in low-level wintertime cloud is important as it will primarily act to enhance the surface warming through their long wave warming effect in the absence of incoming solar radiation.

3.5 Model differences in BC-induced Arctic Amplification

Light-absorbing particles, such as BC, deposited on snow and ice may reduce the albedo and enhance melting, with potentially large impacts on the radiative budget. Of the ten PDRMIP models, three include the radiative effect of BC on snow. Figure 6 shows AA responses to the BCx10 experiment from individual models. Model names are marked with [S] if they include BC on snow and are shown as solid lines in the righthand panel. We focus on land-average AA, where BC snow-albedo effects are expected to be largest (Goldenson et al., 2012). Note that the CESM-CAM5 model has a particularly strong annual mean AA partly due to its very small global warming.

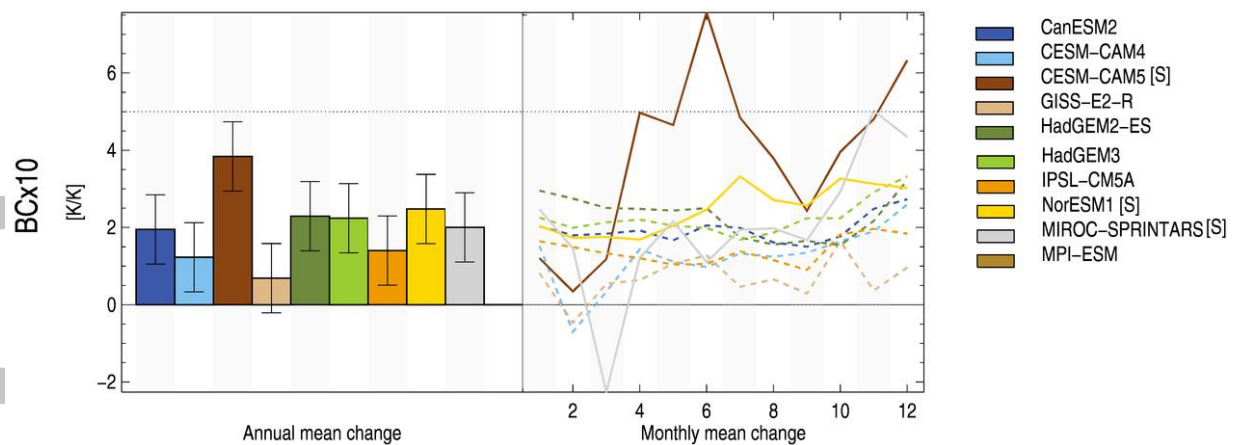


Figure 6: Annual mean (bars) and monthly mean (graphs) changes in land surface (normalized by global mean temperature change) AA for the BCx10 experiment, for all contributing models. [S] indicates models with effect of BC on snow/ice. These are marked with solid lines, while dashed lines show models without the effect of BC on snow.

There does not seem to be a systematic difference in the seasonal AA cycle between the models with and without BC-on-snow treatment, except for a general tendency for higher AA in late fall/early winter. The exception is CESM-CAM5, which has a very strong BC-induced amplification in spring/summer. This is consistent with Qian et al. (2014), who found that the Arctic forcing from BC on snow in the CAM5 model is substantial, and is at its maximum around mid-summer. There is a similar although weaker summertime maximum in NorESM1, but no such feature is discernible in MIROC-SPRINTARS. However, due to the wide variety of parameterizations and limited number of available models, these results should be interpreted with care. Further investigation of this effect requires more a dedicated study, where a larger number of models perform sensitivity studies with and without the effect of BC on snow.

4 Arctic precipitation responses to global perturbations

In this section, we discuss the Arctic precipitation response to the five global perturbations. As for temperature, we begin with the annual and seasonal mean responses and the impact of rapid adjustments. Then, we discuss precipitation change through the application of energy and moisture balance analyses.

4.1 Annually and seasonally resolved precipitation responses

While the Arctic surface temperature amplification has been thoroughly investigated, less attention has been given to changes in Arctic precipitation. Here we find that Arctic precipitation change per degree warming, often termed the apparent hydrological

sensitivity (HS), is similar (6-8 %/K) for all the global perturbations (Fig. 1). Samset et al. (2016) analyzed the globally averaged precipitation change for the PDRMIP data set and found a stronger apparent HS for changes in aerosols and solar insolation (around 3 %/K) than for greenhouse gas perturbations (around 2 %/K). It was also found that the global average response to BCx10 was negative (i.e., a decrease). The Arctic hydrological sensitivity is thus found to be stronger than the global averages, consistent with previous findings from simulations using the RCP4.5 and RCP8.5 emission scenarios for a range of CMIP5 models (Bintanja and Selten, 2014). Here, we are able to conclude that this is a robust feature among all drivers. Testing the ten annual mean model values of SO₄-induced precipitation change against the other drivers using Student's t-test we find that SO₄ is associated with a slightly higher precipitation change, which is significantly (at the 95% level) different from the three other non-aerosol drivers.

As trends in Arctic precipitation follows the Arctic warming (Bintanja and Selten, 2014), the seasonal variation in the change generally resembles that of the AA, with a maximum in winter and smallest change during summer. Deser et al. (2010) performed idealized model experiments isolating the impact of future sea ice loss and found the same seasonality in sea ice and precipitation changes, indicating a close link. Our results also show the strongest precipitation changes coincident with the regions of strongest warming (Fig. 2) and loss in sea ice (Fig. S4), and we find that this also is where surface evaporation increases the most (not shown). The seasonal cycle in precipitation changes is most similar between the non-aerosol drivers (CO₂x2, CH₄x3 and SOL). For BCx10 the changes are stronger in late winter/spring than in November/December. For SO₄x5 there is no clear summer minimum, and SO₄-induced precipitation changes in April through October are the highest of all drivers. All drivers have a strong relative increase in summertime convective precipitation (not shown), particularly over land, except for BC which causes a strong reduction due to increased stabilization connected with the strong upper-level warming shown in Fig. 4b. However, the Arctic summertime precipitation is dominated by stratiform precipitation formation, so changes in convective clouds play a smaller role in the seasonal cycle of total precipitation. Instead, the seasonality can likely be attributed to seasonal variation in surface evaporation or moisture transport.

4.2 Separating Arctic precipitation amplification into rapid responses and feedbacks from global temperature change

Like for temperature, we can separate the precipitation response into rapid adjustments and a slow change that scales with global mean surface temperature change. See Figure S3. Here, the BC perturbation produces precipitation response in the Arctic of opposite sign from the global response (Samset et al., 2016; Stjern et al., 2017). Globally, the BC-induced drying has been shown to originate from the strong atmospheric absorption by the aerosols, which produces a very strong negative rapid adjustment that outweighs the final surface-temperature driven increase (Andrews et al., 2010; Kvalevåg et al., 2013; Samset et al.,

2016). However, Samset et al. (2016) demonstrated a regional dependency on the distribution between slow and fast changes, with a dominance of the slow response (increasing precipitation) in the Arctic region. This characteristic is confirmed in Fig. S3, where the rapid precipitation response for BCx10 is close to zero.

4.3 Arctic precipitation changes through an energy balance perspective

To understand the mechanisms behind the Arctic precipitation changes for individual drivers, we utilize the energy budget approach (O’Gorman et al., 2011; Richardson et al., 2016). When we calculate the atmospheric energy budget, contributions to the total precipitation change can be subdivided into changes in longwave atmospheric cooling LWC, shortwave atmospheric heating SWH, the net upward sensible heat at the surface, SH, and the dry static energy flux divergence H_{dry} :

$$LdP = dLWC - dSWH - dSH + dH_{dry} \quad (1)$$

Above, L is the latent heat of condensation of water vapor and P is the surface precipitation. The column-integrated dry static energy flux divergence change dH_{dry} is calculated as the residual between the precipitation change LdP and the other terms, and contributes positively to precipitation change if more dry static energy is transported out of the Arctic atmospheric column in the perturbed climate. Alternatively, we may use a simplified moisture budget, following Bintanja and Selten (2014):

$$LdP = dE + dF_{moisture} \quad (2)$$

Here, E denotes the evaporation from the surface, and the transport of moisture into the Arctic ($dF_{moisture}$) is calculated as the difference between LdP and dE .

The energy and moisture budget analyses of total Arctic precipitation changes are shown in Fig. 7a, again normalized by the global mean temperature change. Note that LdP indicates the total Arctic precipitation change per global warming, but in contrast to Fig. 1, it shows model means (as opposed to model medians) and absolute changes (as opposed to relative changes). This is necessary to produce meaningful energy budgets and calculate residuals but gives rise to some differences between this figure and Fig. 1. Immediately apparent from Fig. 7a is how different the energy budget components are for the global BC perturbation compared to the other drivers. Even so, the total precipitation change is similar between BC and the other drivers, as in Fig.1.

Short wave heating is shown with a negative prefix in Fig. 7a, since a reduction in short wave heating contributes to increased precipitation. For all drivers, this budget term is negative, indicating increases in short wave atmospheric heating. In general, short wave heating is driven by elevated atmospheric water vapor concentrations in the warmer climate (recall

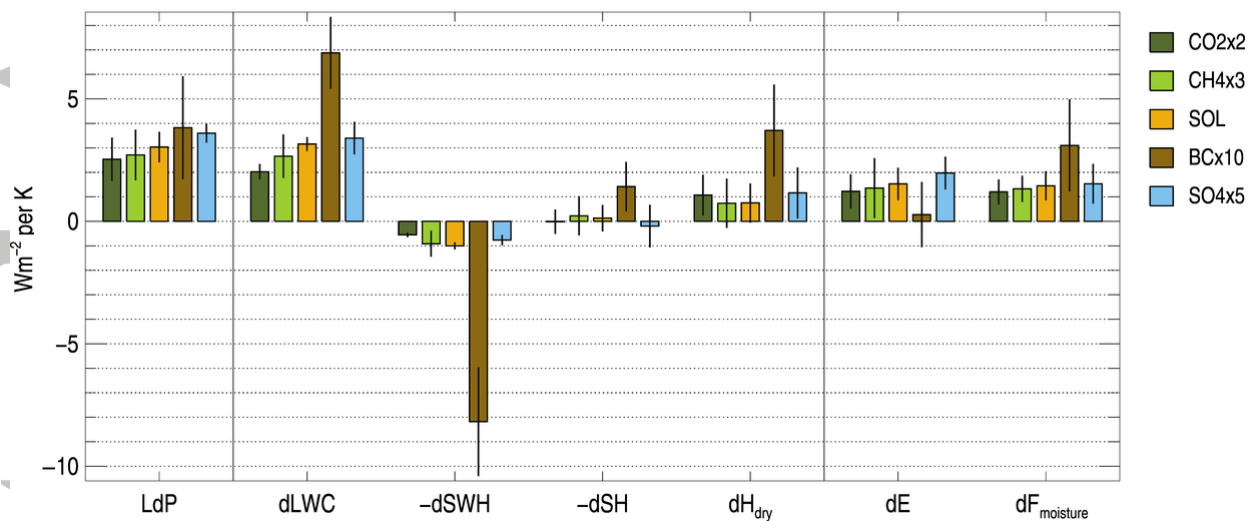
that the sign of the SO_4 responses are turned in the normalization by the global mean temperature change). However, due to the strong additional absorption by atmospheric BC particles, this term is by far strongest for BC. It is also stronger over land than over oceans (not shown) where BC concentrations are highest. Changes in long wave cooling are positive for all drivers, again as a general response to the warming climate. The long wave cooling is strongest for BC since long wave radiation is re-emitted from high-altitude BC particles after their absorption of solar radiation. The smaller long wave cooling of $\text{CO}_2 \times 2$ than $\text{SO}_4 \times 4$ is because in the Arctic, as opposed to in the global average, the rapid response to an increase in CO_2 is a reduction in long wave cooling, which lowers its total increase. The sensible heat flux is affected by changes in the temperature difference between the surface and the overlying air, as well as by changes in turbulence and convection. Globally, the change in this quantity is minor for all drivers but BC, as shown in a previous PDRMIP study (Myhre et al., 2018). We find that this is true also in the Arctic. The strong atmospheric warming caused by BC (Fig. 4) leads to a marked reduction in the sensible heat flux, which is stronger over land than over oceans. Finally, all drivers cause an increase in the dry static energy flux divergence. This term is negative (indicating energy transport into the Arctic) in the baseline climate but slightly less negative in the perturbed climate, meaning that the poleward energy transport is reduced. This is a natural consequence of the Arctic amplification, by which the high latitudes warm more than the lower latitudes and the meridional temperature gradient is reduced. For all drivers, but especially for BC, the reduced dry static energy flux divergence is strongest in the summer season (not shown).

Previous studies highlight the role of moisture transport from lower latitudes to Arctic precipitation changes (Bengtsson et al., 2011; Kug et al., 2010). Still, indications from both observations (Kattsov and Walsh, 2000) and model simulations (Bintanja and Selten, 2014) are that the dominating cause of the precipitation change may be the strong intensification of local surface evaporation. Here, we find that the change in moisture transport (dF_{moisture}) is of similar magnitude as increased evaporation (dE) for $\text{CO}_2 \times 2$, $\text{CH}_4 \times 3$ and SOL. For SO_4 , contribution from evaporation is higher than that from moisture transport and for BC there is a statistically significant higher impact from the latter, whereas the evaporation change is negligible. This means that the dominating processes behind Arctic precipitation changes may shift depending on what climate driver has the dominating impact in the Arctic.

We next examine the seasonal variations in the dF_{moisture} and dE terms, and their contribution to the total precipitation change. In Fig. 7b the seasonal absolute change in precipitation is shown as a solid black line and the monthly contribution of evaporation and moisture transport changes to this precipitation change is indicated by the bars. For $\text{CO}_2 \times 2$, $\text{CH}_4 \times 3$, SOL and $\text{SO}_4 \times 5$, the increase in surface evaporation is greatest in June, with a secondary maximum in late fall/early winter. BC, on the other hand, causes a small increase in evaporation in all seasons except summer, when there is a notable reduction. This is when available incoming solar radiation and enhanced BC concentrations combine to produce the seasonally strongest reduction in the amount of short-wave radiation that

reaches the surface, an effect known as solar dimming. For BC, the moisture transport term is much larger than for the other drivers, and it is the dominant cause of the BC-induced Arctic precipitation changes. During winter, the relative importance of evaporation and moisture transport changes for SO_4 are comparable to the other (non-BC) drivers, while evaporation changes dominate in the summer. Recall, however, that the sign is turned by the normalization by a negative global mean temperature change. In reality, increased concentrations of SO_4 causes reduced moisture transport and evaporation throughout the year, but like for BC, there is a particularly strong reduction in evaporation in the summer months due to enhanced solar dimming. Conversely, CO_2 induces the weakest evaporation change of all the drivers in this season due to a negative rapid response (not shown) that lowers the total positive change. These differences explain the stronger precipitation-response to SO_4 than to the other non-aerosol drivers, as shown in Fig. 1.

a)



b)

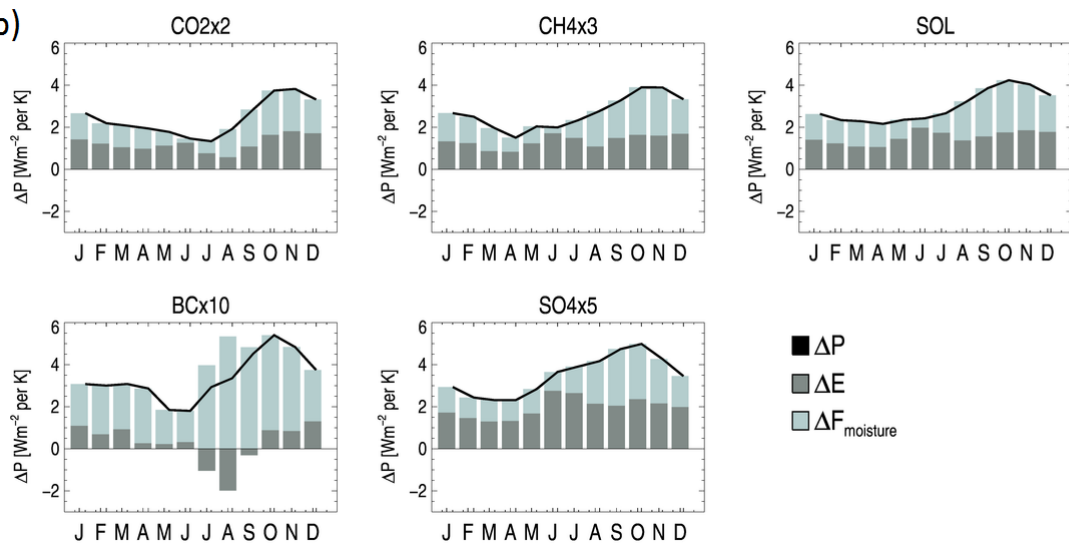


Figure 7. a) Model mean atmospheric energy budget and moisture budget analysis for the Arctic, normalized by global mean temperature change. The sign convention is such that positive values indicate

that the term contributes to increased precipitation. LdP is the total precipitation change. Components of the energy budget are shown in between the two vertical lines, where $dLWC$ is the change in longwave atmospheric cooling, $dSWH$ is the change in shortwave atmospheric heating, dSH is the change in sensible heat at the surface and dH_{dry} is the change in dry static energy flux divergence. Components of the moisture budget are at the rightmost side, where dE is change in surface evaporation and $dF_{moisture}$ is moisture transport into the Arctic. Vertical lines show the inter-model standard deviation. b) Contribution of the moisture budget terms to the absolute monthly precipitation change.

For all drivers we find a maximum change in poleward moisture transport around August through October. This summer/autumn maximum is likely a contributing cause for the strong seasonal upper-level warming for BC, for which the moisture transport term is particularly strong. Moisture from lower latitudes can be a source of heat within the Arctic, in the form of latent heat release, increased water vapor greenhouse effect or changes to clouds. Sedlar and Tjernström (2017) found a clear pattern of increased lower-level and decreased upper-level clouds in months with particularly strong moisture advection into the Arctic, similar to the cloud change pattern we see during summer for BCx10 (Fig. 5).

We note for consistency that Bintanja and Selten (2014) found strongest late-century increases in moisture transport in summer/early autumn for a range of CMIP5 models under the RCP4.5 and RCP8.5 emission scenarios. The later seasonal maximum in our data is related to our broad definition of the Arctic as all area north of $60^{\circ}N$: using $70^{\circ}N$ instead shifts the maximum transport change towards mid-summer (see Fig. S7).

5 Arctic responses to regional perturbations

Seven of the ten models performed additional simulations, where BC was perturbed in a subregion of Asia and SO_4 was perturbed in parts of Asia and Europe (see Section 2 for detailed description of these regions). Two of these models ran emission-based simulations, leading to additional model differences due to differences in transport, wet removal and other processes. The remaining five used concentration-based set-ups, which diminishes the sources for model differences, and allows for more focus on dynamical responses. Given these differences, we have here chosen to present the concentration-based and the emission-based results separately, even if the latter will only show the mean of two models. Figure 8 shows the equivalent of Fig. 1, where solid and hatched bars represent global and regional perturbations, respectively, but for the subset of five concentration-based models in Fig. 8a and two emission-based in Fig. 8b.

Starting with the concentration-driven models, we find that the Asian BC perturbation produces a stronger Arctic amplification, as well as a stronger precipitation response, than the global perturbation. This contrasts the response to the regional SO_4 perturbations,

which both cause an AA more like the global experiment. There is a tendency for a stronger AA and stronger precipitation response from the European than from the Asian SO_4 perturbation, but the differences are not statistically significant. Figure 9a shows vertical AA profiles by latitude north, similar to Fig. 4a, for the concentration-driven models. The warming pattern for BCx10asia is more focused around the latitudes of the regional emissions (around 40°N), but otherwise resembles the global pattern. The AA signals from the regional SO_4 perturbations have much clearer maxima over the emission latitudes than the global signal, but as opposed to the regional BC perturbation they do not show the same clear poleward extension of the AA.

The two emission-driven models (Fig. 8b) have stronger AA than the concentration-driven models, both for the global and the regional experiments. As mentioned in Section 2, emission-driven models have the potential for an amplified or dampened response since the model's climate response may feed back on the BC concentrations, as found in Sand et al. (2015). Notably, while regional BC perturbations caused stronger AA than global for the concentration-driven models, the opposite is true for the emission-driven, due to a very strong AA for the global perturbation. However, this includes results from NCAR-CAM5, which was seen to be an outlier (Fig. 1). With the exception of the BCx10 experiment, the AA responses of the two individual emission-driven models are very similar (note the circles in the figure, indicating individual model values).

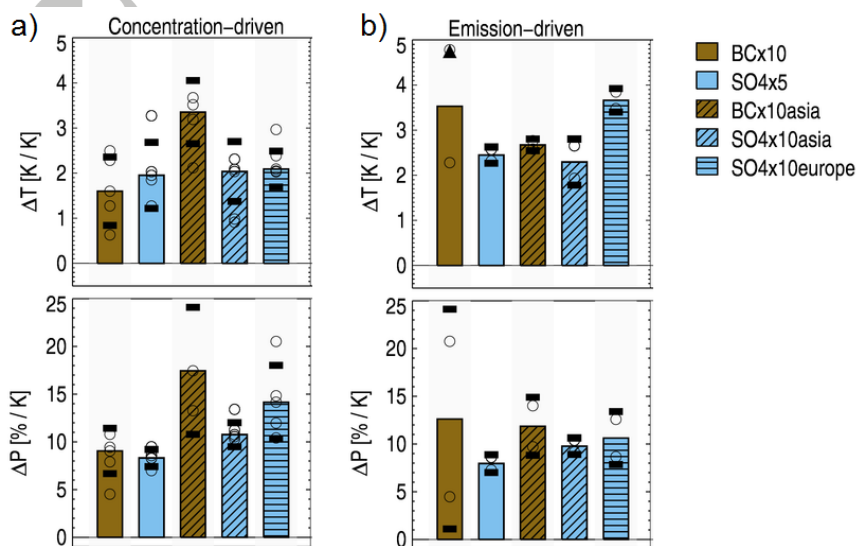


Figure 8. Responses in temperature and precipitation (normalized by the global mean temperature change) to global (solid bars) versus regional (hatched bars) perturbations in BC and SO_4 , averaged over the Arctic, for a) five concentration-based models (see Table S1) and b) two emission-based models. Bars show annual model median values, individual model values are shown as circles. Horizontal bars indicate \pm one standard deviation, and triangles are used to indicate when standard deviations are outside the range of the axes limits.

While the concentration-driven models had no discernible AA difference between the two regional SO_4 perturbations, the tenfold increase in SO_4 in Europe causes a significantly stronger AA than a tenfold increase in Asia for the emission-driven models. This difference is particularly clear in the summertime, while wintertime responses are more similar (not shown). The similarity in wintertime responses can be explained by that fact that during winter, Europe and Eurasia are both located within the polar dome (i.e., the surfaces of constant potential temperature). This, and the fact that parts of these regions are covered in snow in winter, which cools the near-surface air and eases transport into the polar dome, allows air masses from both these regions to reach the Arctic through low-level transport (Stohl, 2006). Air masses that are too warm (i.e., air originating from lower latitudes in summer) for isentropic transport reach the Arctic through lifting outside the polar dome and may subsequently slowly descend within. Indeed, comparing seasonal vertical warming patterns between the European and the Asian SO_4 perturbations in Fig. S6, we see that while the summertime Arctic response to SO_4 perturbations in the (warmer, more remote and with more effective convective lifting) Asian region is mostly confined to the mid-troposphere, the impact from European SO_4 extends all the way to the surface. This produces notable differences in surface AA (Fig. 8b) between the two experiments. A wintertime warming, on the other hand, is visible at the Arctic surface in both the European and the Asian SO_4 perturbation experiments. The Europe-only perturbation also yields a higher precipitation response than the Asian-only perturbation.

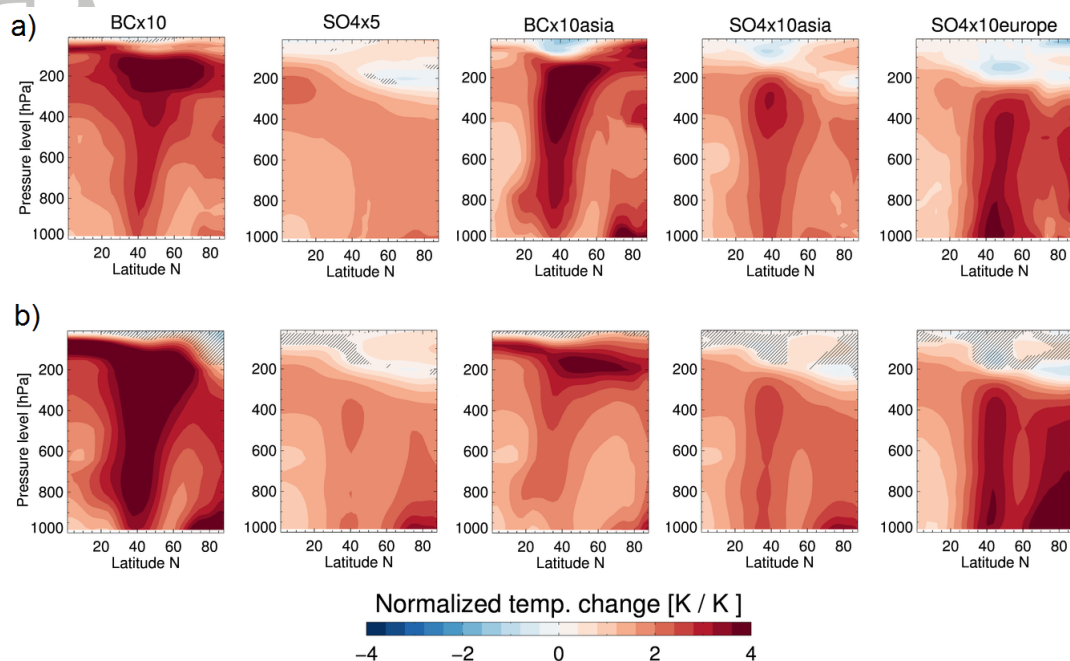


Figure 9. Multi-model median vertically resolved Arctic amplification for the annual mean for latitudes 0 to 90°N and for the model median on a) five concentration-based models and b) two emission-based models. Hatching indicates where fewer than 75 % of the models agree on the sign of the change.

Our findings for the emission-driven models, that Arctic temperature and precipitation responses are significantly stronger for the European than for Asian perturbations, are in line with previous findings (Aamaas et al., 2016; Bellouin et al., 2016; Sand et al., 2016; Stjern et al., 2016; Yu et al., 2013). For instance, Sand et al. (2016) shows that Europe and Russia are the regions that influence Arctic temperatures the most per unit emissions, and the large sulfate emission change in Europe over the past decades have been linked to changes in Arctic sea ice cover (Acosta Navarro et al., 2016; Gagné et al., 2017; Mueller et al., 2018). The greater impact of European emissions is likely due to factors connected with the nearer proximity of Europe to the Arctic, but may also be related to a saturation of aerosol-cloud interactions over East Asia, as well as a greater climatological cloud cover, which could mask the direct aerosol forcing in that region (Kasoar et al., 2018). In another PDRMIP analysis Liu et al. (2018) also found that the forcing from European SO₄ increases had a stronger efficacy in terms of global temperatures and precipitation effects than that from the Asian region. In the present study we find the concentration-based models do not show this difference, at least in the Arctic. However, the regional PDRMIP simulations is limited in terms of number of regions, making it difficult to confidently conclude on regional impacts on Arctic climate.

6 Discussion

Common for the Arctic impact studies mentioned in the previous section is that they focus on responses per unit emissions. In reality, SO₂ emissions in Europe are relatively low today, while SO₂ emissions in India are currently increasing strongly (Li et al., 2017). Sand et al. (2016) stress that due to the large absolute emissions of SLCF (particularly BC) from the Asian continent, this is presently the largest source of SLCF-driven Arctic warming – both due to long-range transport and due to localized radiative impact and subsequent transport of heat and moisture. However, large emission changes in this region are ongoing (van der A et al., 2017). Recent studies suggest that emissions of sulfur dioxide in China has decreased by 75 % over the past decade and that India is taking over as the largest single-region emitter (Li et al., 2017). In contrast, emissions of BC and CO₂ have continued to increase in both regions (e.g., Hoesly et al., 2017). Persad and Caldeira (2018) looked at impacts from anthropogenic aerosol (including both SO₄ and BC) changes in a range of regions, using the NCAR-CAM5 model, and found substantial variations in how emissions from different regions influence the global temperature. They found that regions which historically or presently account for the majority of anthropogenic aerosol emissions have the largest impacts on temperature, while regions where emissions currently have or are projected to have strong trends (i.e., India or East Africa) have smaller temperature impacts. This suggests that, globally averaged, the future distribution of main emission regions may have a smaller cooling potential than historically. Importantly, however, one reason for the small global temperature impact of Indian emissions was a regionally heterogeneous pattern of

both positive and negative temperature responses. And while increased aerosol emissions in general cause a cooling, the Arctic response to Indian emission changes in this study was a warming.

Energy and moisture transport from lower latitudes play an important role in the impact on Arctic climate. Recent research has looked closer into how poleward energy transport contributes to current and future AA, and to how it will react to the reduced meridional temperature gradient following an AA (Graversen, 2006; Jonas et al., 2016; Kapsch et al., 2013). They find that it is the transport of latent heat, and not of dry-static energy, that dominates the contribution to AA. As for future changes in these energy transport terms, simulations show that while models do not agree on the sign of the change in the total poleward heat transport, they all predict a reduction in the dry static energy transport (as the faster Arctic warming reduces the temperature gradient between high and low latitudes), and an increase in the latent heat transport (Hwang et al., 2011). Here, we find that the contribution from the dry-static energy flux divergence to the Arctic precipitation change differs between climate drivers, but is particularly important for BC. It is however positive for all drivers and all individual models, indicating a reduction in northward energy transport. We also find, for all drivers, an increase in the northward transport of moisture, consistent with Hwang et al. (2011). We note that the ocean has been shown to play an important role in the poleward heat transport (Nummelin et al., 2017), but here, we have not looked at changes in this component. A separate study investigating how different climate drivers influence the various elements of the poleward energy transport could improve our ability to predict how the Arctic would respond to future emission and climate changes.

Many of the processes involved when simulating climate responses in the Arctic are associated with high levels of uncertainty. One example is the effect of BC aerosols on ice and snow, with estimates ranging from relatively weak (Bond et al., 2013) to stronger (Flanner et al., 2007) effects. Here, three of the contributing models (NorESM1, CESM-CAM5 and MIROC-SPRINTARS) include estimates of the radiative effect of BC on snow in their models, but in only one of these (CESM-CAM5) do we see a clear signal from this effect in the seasonal warming in the Arctic. Differences in how models treat processes such as BC on snow nevertheless contributes to be a source of uncertainty and deviations between models. Previous multi-model studies have found that the spread in model responses is particularly large in the Arctic (Sand et al., 2017), and this is also the case here: Table S3 shows that the inter-model standard deviation for both temperature and precipitation is typically a factor three or more larger in the Arctic than globally.

7 Summary and Conclusion

The Arctic is influenced by both local and remote emissions (Acosta Navarro et al., 2016; Gagné et al., 2017). Presently, global and regional trends in emissions are shifting rapidly. While global CO₂ emissions are still increasing (Le Quéré et al., 2018), there are significant regional differences in air pollutant trends (Hoesly et al., 2017; Li et al., 2017). Combined with underlying climate change, potentially affecting transport pathways (Jiao and Flanner, 2016), this may shift the relative importance of individual climate drivers for Arctic climate change. Moreover, aerosol emissions in individual regions may have a different impact on Arctic climate to global perturbations.

The present study analyzes the response of perturbations to five individual climate drivers (CO₂, CH₄, the solar constant, SO₄ and BC), investigating effects on Arctic temperature and precipitation. While some notable differences exist, particularly for BC, response patterns in the vertical temperature profile are broadly similar between climate drivers. The Arctic amplification in surface temperature ranges only from 1.9 [± 0.4] (SOL) to 2.3 [± 0.6] (SO₄x5) between drivers, and inter-driver differences are not statistically significant. As indicated in previous studies, the majority of the Arctic response seems to originate from a generalized dynamical response to the climate warming. This is also reflected in the present analysis through the similar surface Arctic amplification (AA) pattern across all drivers and in the fact that all drivers cause the strongest surface warming in winter. All drivers display a similar pattern of upper-level (~400 hPa) AA in the summer, and a reduction in cloud cover in the same altitude and season. For BC, we see a much stronger upper-level AA than for the other climate drivers. This is likely connected to increased high-level concentrations of BC in summer, as well as a very strong BC-induced summertime increase in poleward moisture transport, which may induce additional warming through short wave absorption by BC particles and an added water vapor greenhouse effect. This warming does not propagate down to the Arctic surface. A stronger increase in summertime low cloud cover (which tends to cool the surface) for the BC perturbations may instead act as a buffering effect, so that all climate drivers end up causing relatively similar surface AA, with maxima in winter and minima in summer.

Global temperature change-normalized precipitation responses are also relatively similar between drivers, although less so than for temperature. SO₄ causes significantly stronger (normalized) precipitation increases than the other non-aerosol drivers, driven by a much stronger evaporation change. This difference is particularly pronounced in the summer. An energy budget break-down of the contribution of different processes to the precipitation change shows that the BC response is substantially stronger than those of the other climate drivers in all these processes. In total, however, the different contributions cancel out to leave a precipitation change similar to that from the other climate drivers. While changes in poleward moisture transport and changes in evaporation have similar contributions to the total precipitation change for CO₂x2, CH₄x3 and SOL, the aerosol-driven precipitation

changes differ, with a dominance of moisture transport in the case of BCx10 and a dominance of evaporation changes for SO₄x4.

Our study presents a first, comprehensive overview of how Arctic temperature and precipitation responds to individual climate drivers. While further analysis is needed to fully disentangle all the processes that act to shape the Arctic climate response, we conclude that overall, responses are similar enough to provide confidence in scenario-based projections for Arctic climate change regardless of the detailed balance between components of future emissions.

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Supporting Information for

Arctic amplification response to individual climate drivers

Camilla Weum Stjern¹, Marianne Tronstad Lund¹, Bjørn Hallvard Samset¹, Gunnar Myhre¹, Piers M. Forster², Timothy Andrews³, Olivier Boucher⁴, Gregory Faluvegi^{5,13}, Dagmar Fläschner⁶, Trond Iversen⁷, Matthew Kasoar⁸, Viatcheslav Kharin⁹, Alf Kirkevåg⁷, Jean-François Lamarque¹⁰, Dirk Olivié⁷, Thomas Richardson², Maria Sand¹, Dilshad Shawkil⁸, Drew Shindell¹¹, Christopher J. Smith², Toshihiko Takemura¹², Apostolos Voulgarakis⁸

¹ CICERO Center for International Climate and Environmental Research, Oslo, Norway

² University of Leeds, Leeds, United Kingdom

³ Met Office Hadley Centre, Exeter, UK

⁴ Institut Pierre-Simon Laplace, Univ. P et M. Curie / CNRS, Paris, France

⁵ Center for Climate Systems Research, Columbia University, New York, USA

⁶ Max Planck Institute for Meteorology, Hamburg, Germany

⁷ Norwegian Meteorological Institute, Oslo, Norway

⁸ Imperial College London, London, United Kingdom

⁹ Canadian Centre for Climate Modelling and Analysis, Victoria, BC, Canada

¹⁰ NCAR/UCAR, Boulder, USA

¹¹ Duke University, Durham, NC, USA

¹² Kyushu University, Fukuoka, Japan

¹³ NASA Goddard Institute for Space Studies, New York, USA

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Tables S1 to S3

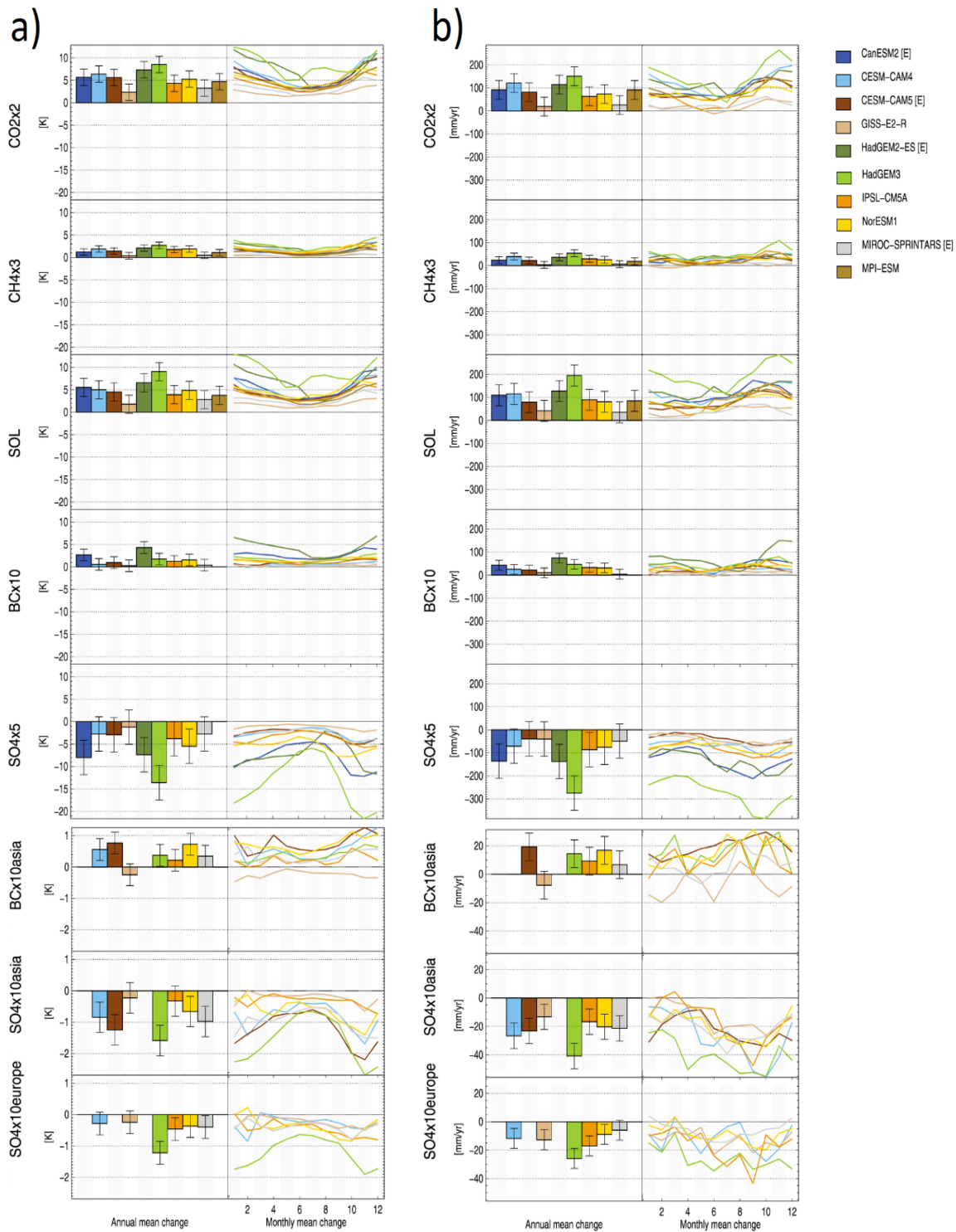


Figure S1: Individual model response in Arctic (a) temperature [K] and (b) precipitation [mm/yr] averaged over the year, as well as monthly mean changes. Note that the y axes in the lowermost regional panels have different scales.

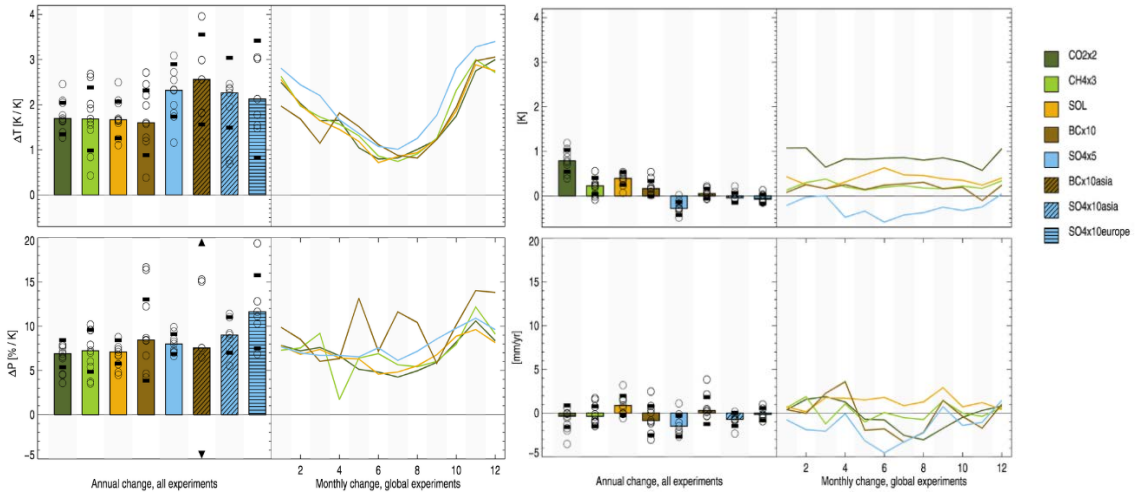


Figure S2. Like Fig. 1 but for slow responses (left) and fast response (right) in temperature (upper) and precipitation (lower) averaged over the Arctic (all area north of 60° N). Slow responses are normalized by the global mean temperature change.

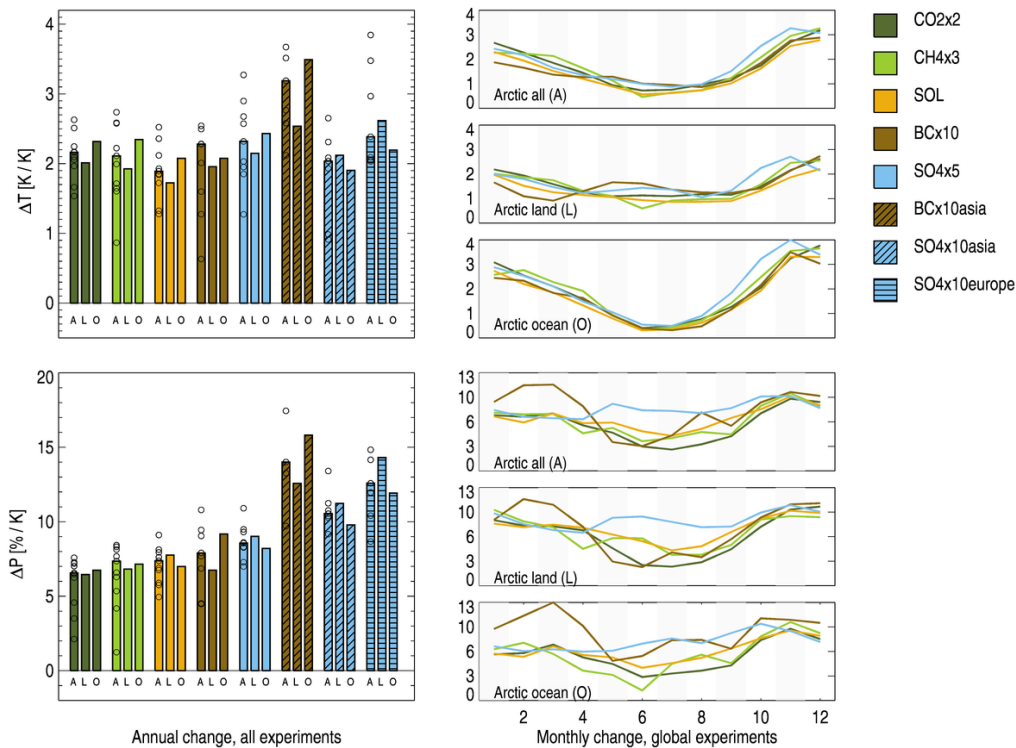


Figure S3. Basic responses in temperature and precipitation (normalized by the global mean temperature change) to perturbations in the five climate drivers, averaged over all Arctic (all area north of 60° N; A), Arctic land (L) and Arctic Ocean (O), respectively. Bars show annual model median values, individual model values (for Arctic all only) are shown as circles. Graphs show monthly mean changes (for the global experiments only).

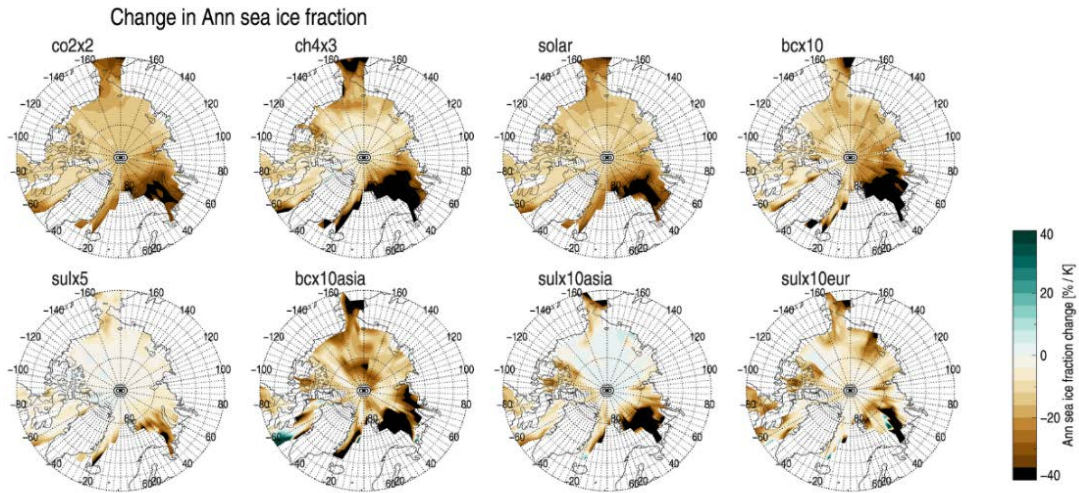


Figure S4. Multi-model median change in annual mean sea ice fraction for all experiments, normalized by the global temperature change.

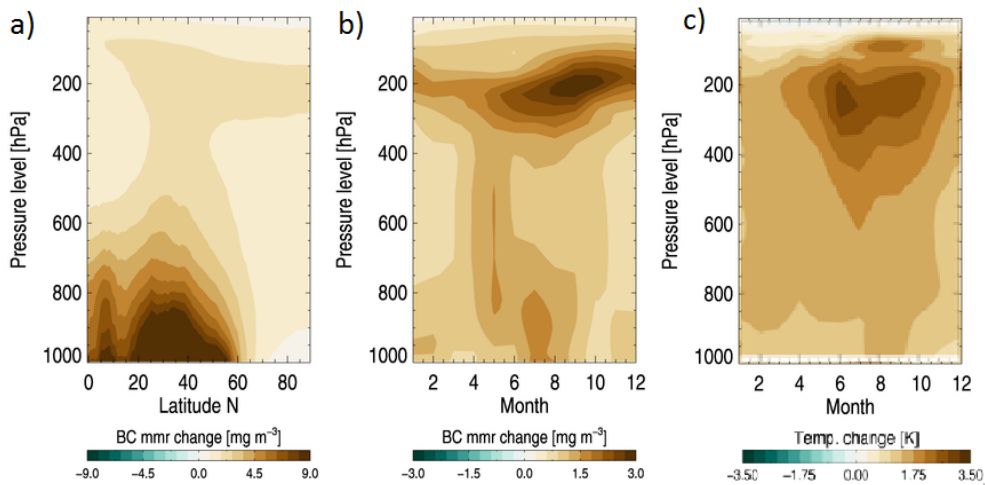


Figure S5. Change in a) BC mass mixing ratio for the BCx10 experiment for the annual mean for latitudes 0 to 90°N, b) BC mass mixing ratio for the Arctic (60 to 90°N) average for each month of the year and c) warming over the East Asia region (10-50 °N, 60-140 °E) in response to the global tenfold increase in BC (experiment BCx10).

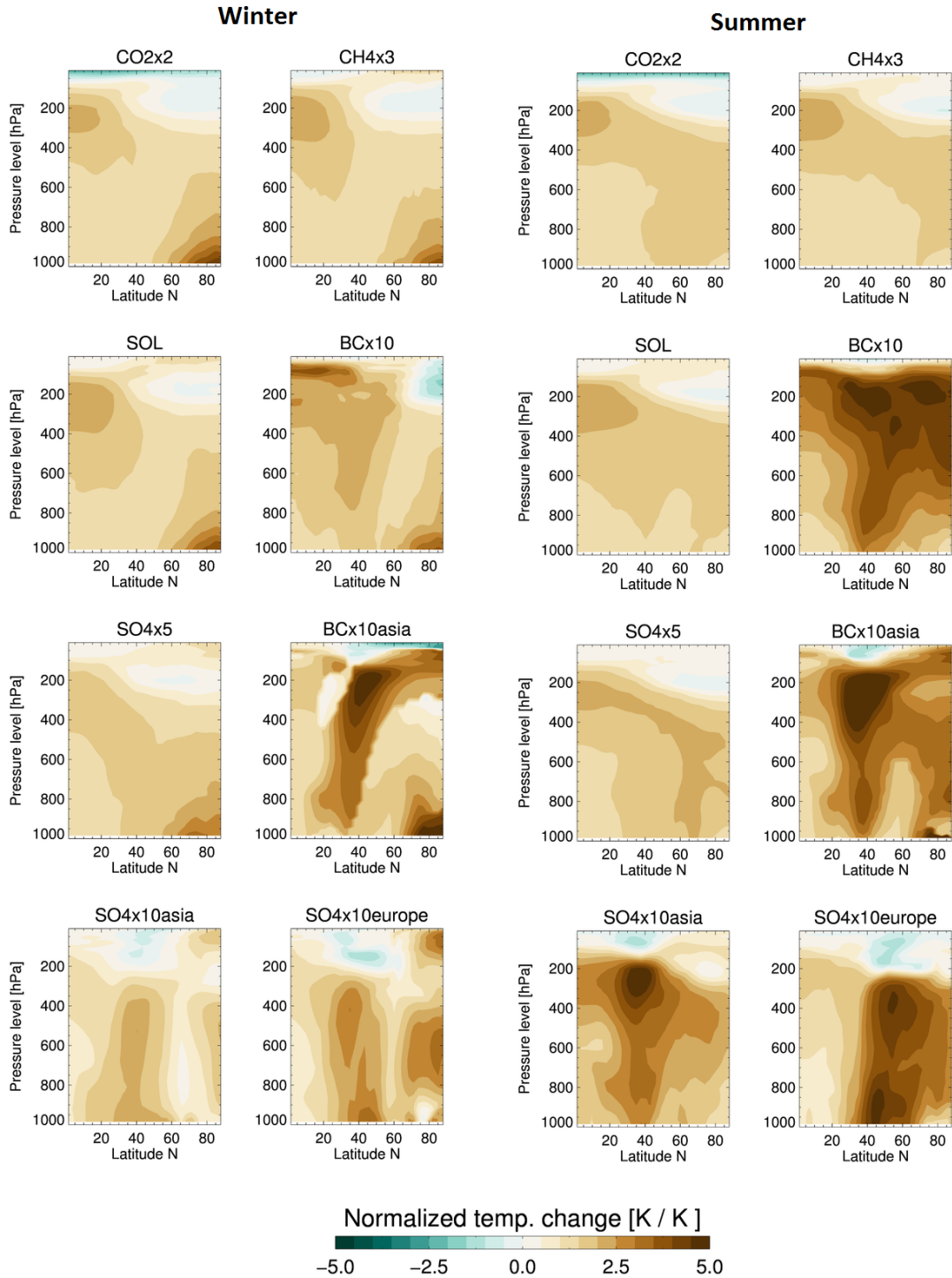


Figure S6. Warming over Northern Hemisphere latitudes for all experiments, averaged over winter (December, January, February) and summer (June, July, August) months, and normalized by the global mean temperature change.

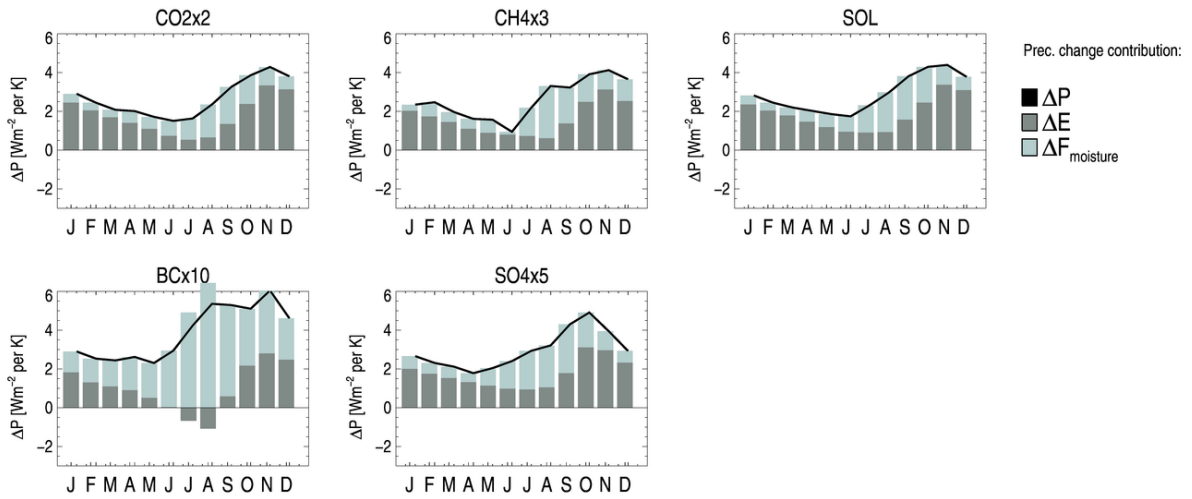


Figure S7. Like Fig. 7b, only for a narrower definition of the Arctic, where Arctic is the area north of 70°N .

Table S1. Overview of models included in the study. Models driven by fixed concentrations in the aerosol experiments are marked by a subfix [C] while the emission-driven models are marked with [E] (MPI-ESM did not perform the aerosol experiments and is therefore marked as neither). An asterisk means that indirect aerosol effects of BC aerosols are not included in the model's microphysical scheme.

	Emissions		Performed the regional experiments	Treats BC on snow	Includes indirect aerosol effects
	BASE	Perturbation experiments			
CESM-CAM4 [C]	Mostly year 2000, Lamarque et al. (2010), although some variation, see <i>Myhre et al.</i> [2013].		Yes	No	Yes*
GISS-E2-R [C]			Yes	No	No
HadGEM3 [C]			Yes	No	Yes*
IPSL-CM5A [C]			Yes	No	Yes
NorESM1 [C]			Yes	Yes	Yes
CanESM2 [E]	---	---	No	No	Yes*
CESM-CAM5 [E]	Year 2005, Lamarque et al. (2010)	Year 2005 x factor	Yes	Yes	Yes*
HadGEM2-ES [E]	Year 1860, Lamarque et al. (2010)	Year 2000 x factor	No	No	Yes*
MIROC-SPRINTARS [E]	Year 2010, <i>Janssens-Maenhout et al.</i> [2015]	Year 2010 x factor	Yes	Yes	Yes
MPI-ESM-LR	---	---	No	No	No

Table S2. Arctic amplification (AA; calculated as Arctic mean divided by global mean temperature change) for each model and experiment. MMM shows the multi-model mean. Models driven by fixed emissions in the aerosol experiments are marked by a suffix [E] while the models which include treatment of BC on snow are marked with [S]. MPI-ESM did not perform the aerosol experiments.

	CO2x2	CH4x3	SOL	BCx10	BCx10asia	SO4x5	SO4x10asia	SO4x10europe
CanESM2 [E]	2.10	2.03	1.87	2.04		2.95		
CESM-CAM4	1.99	1.74	1.74	1.28	3.64	1.87	2.05	1.99
CESM-CAM5 [E S]	2.21	2.66	1.94	4.98	2.85	2.62	2.71	
GISS-E2-R	1.57	0.93	1.32	0.66		1.30	0.95	2.03
HadGEM2-ES [E]	2.68	2.62	2.57	2.59		2.72		
HadGEM3	2.29	2.27	2.16	2.55	2.14	2.05	2.36	3.01
IPSL-CM5A	1.68	1.68	1.33	1.63	3.32	1.97	1.01	2.11
NorESM1 [S]	2.55	2.79	2.40	2.30	3.74	3.33	2.10	2.35
MIROC-SPRINTARS [E S]	2.23	1.62	1.98	2.37	2.62	2.36	1.97	3.96
MPI-ESM	2.13	2.15	1.88	---	---	---	---	---
MMM	2.14	2.05	1.92	2.27	3.05	2.34	1.88	2.58

Table S3. Inter-model standard deviation in global versus Arctic mean change in temperature (T) and precipitation (P), for all experiments.

	Global, T	Arctic, T	Global, P	Arctic, P
CO2x2	0.7	1.8	1.5	6.9
CH4x3	0.3	0.7	1.0	2.5
SOL	0.8	2.0	2.0	7.4
BCx10	0.5	1.3	0.9	3.7
SO4x5	1.8	3.8	4.5	12.7
BCx10asia	0.1	0.3	0.2	1.9
SO4x10asia	0.1	0.5	0.4	1.3
SO4x10europe	0.1	0.4	0.3	1.0