Changes in Inorganic Fine Particulate Matter Sensitivities to Precursors Due to Large-Scale US Emissions Reductions *

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Changes in Inorganic Fine Particulate Matter Sensitivities to Precursors Due to Large-Scale US Emissions Reductions

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Supporting Information

ABSTRACT: We examined the impact of large US emissions changes, similar to those estimated to have occurred between 2005 and 2012 (high and low emissions cases, respectively), on inorganic $PM_{2.5}$ sensitivities to further NO_{x} , SO_{2} , and NH_3 emissions reductions using the chemical transport model GEOS-Chem. Sensitivities to SO_2 emissions are larger year-round and across the US in the low emissions case than the high emissions case due to more aqueous-phase SO_2 oxidation. Sensitivities to NH_3 emissions are larger in the low emissions case, more than $2\times$ those of the high emissions case in parts of the northern Midwest. Sensitivities to NH_3 emissions are smaller (~40%) in the low emissions case, year-round, and across the US. Differences in NO_x and NH_3 sensitivities indicate an altered atmorphasic acidity. Larger sansitivities SO_2 and NO_2 in the low emission.



atmospheric acidity. Larger sensitivities to SO_2 and NO_x in the low emissions case imply that reducing these emissions may improve air quality more now than they would have in 2005; conversely, NH_3 reductions may not improve air quality as much as previously assumed.

INTRODUCTION

Fine particulate matter (PM_{2.5}) adversely affects cardiovascular and respiratory functioning¹ and is a key focus of air quality policies such as the National Ambient Air Quality Standards (NAAQS). Designing effective policies requires knowledge of how PM_{2.5} responds to changes in its precursors-its sensitivity to emissions. The precursors of inorganic PM2.5 are nitrogen oxides (NO_x) , sulfur dioxide (SO_2) , and ammonia (NH_3) . US NO_x and SO_2 emissions decreased by 42% and 62%, respectively, between 2005 and 2012, while NH₃ emissions remained approximately constant.² These emissions changes are potentially large enough to change the sensitivity of PM2.5 to future emissions reductions and hence change the expected benefits of air quality policies. We analyze where and to what extent large changes in precursor emissions, similar to those that occurred between 2005 and 2012, alter US PM2.5 sensitivities to further emissions reductions.

Several chemical mechanisms affect $PM_{2.5}$ concentrations. Nitric acid (HNO₃) is formed from NO₃, and the fraction of NH₃ and HNO₃ in particulates (the partitioning of these species) depends on ambient temperature, humidity, and acidity. NH₃ is the primary basic species, forming ammonium (NH₄⁺) in particles to neutralize acidic nitrate (NO₃⁻) and sulfate (SO₄²⁻, formed from SO₂). PM_{2.5} sensitivities to NH₃ is a cost-effective strategy to reduce PM_{2.5}.^{3–8} Sulfate is not volatile like NH₃ and HNO₃, but it influences the ambient acidity. PM_{2.5} concentrations can even increase as SO₄²⁻ concentrations decrease, by allowing more HNO₃ to condense.^{3,9} However, multiple reactions oxidize SO_2 into SO_4^{2-} , and each reaction responds differently to NO_x and hydrocarbon concentrations.^{10,11}

 $PM_{2.5}$ sensitivities have been estimated using a variety of approaches, including finite differences, ^{12,13} direct decomposition, ^{14–17} and adjoint modeling.^{18,19} Sensitivity estimates calculate derivatives based around atmospheric conditions associated with particular emissions. Extrapolating these estimates to other levels of emissions is associated with some degree of error. Zhang et al.¹⁷ show that a linear extrapolation underestimates the PM_{2.5} response to a 50% decrease in NO_x emissions by 15%, averaged over the US, with local underestimates up to 50%. Higher-order sensitivity analysis can more accurately predict responses beyond the linear range, ¹⁴ but these approaches are computationally demanding.

Studies of health and economic impacts of emissions often apply estimates of $PM_{2.5}$ sensitivities to parametrize how air quality responds to emissions. Muller and co-workers^{12,20,21} use the integrated assessment model APEEP to calculate marginal damages (in US dollars per ton) by increasing emissions from one source by one ton and tracing impacts on human health, agriculture, and other vulnerable structures. The base case in APEEP uses the EPA's 2002 National Emissions Inventory (NEI02), but Muller²¹ implements the 2005 inventory

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(NEI05). Fann, Baker, and Fulcher^{22,23} use the CAMx Particle Source Apportionment Technology to attribute $PM_{2.5}$ concentrations to emissions from each economic sector. Their simulations use NEI05 and projections for 2016 based on proposed air quality rules. Similarly, Fann, Fulcher, and Hubbell²⁴ use NEI02 with projections for 2015 as the emissions inventory for EPA's Response Surface Model of air quality.²⁵ In these studies, the projections based on thenproposed rules exhibit 30% decreases in national NO_x and SO₂ emissions between 2001 and 2010.²⁵ These emissions actually decreased by 34% and 53%, respectively, and NH₃ emissions increased by 17%. Whether sensitivities calculated using older (2002 and 2005) emissions, or even past estimates of current emissions, are sufficiently accurate for health and economic impact assessment depends on the nonlinear response of PM_{2 5}.

We evaluate the influence of large NO_x and SO_2 emissions changes (comparable to those that occurred in the US between 2005 and 2012) on $PM_{2.5}$ sensitivities and identify the most important nonlinear processes. We find that lower NO_x emissions lead to higher SO_2 sensitivities across the US and to higher NO_x sensitivities in winter in cold, humid regions such as the northern Midwest. Lower NO_x and SO_2 emissions also yield smaller sensitivities to NH_3 . Our results suggest that the benefits of NO_x reductions could be much larger now that emissions are lower, especially in winter (when NH_3 emissions were thought to be dominant). We also show that SO_2 controls are still effective, despite >60% reductions nationally. The potential changes identified highlight the need to review the sensitivities used in health, economic, and policy studies and consider a multipollutant approach to air quality policy.

METHODOLOGY

Chemical Transport Model. We use the GEOS-Chem chemical transport model v9-02^{26,27} (http://geos-chem.org/). GEOS-Chem has previously been evaluated against measurements of ozone and hydrocarbon concentrations²⁶ and inorganic PM_{2.5} component concentrations.^{7,28,29} It has been used to study US air quality^{7,8} and climate change impacts on PM_{2.5} formation.^{30–32}

GEOS-Chem simulates ozone-NO_x-hydrocarbon-aerosol chemistry²⁷ coupled to inorganic aerosol thermodynamics, which determines the partitioning of NH₃ and HNO₃. The thermodynamic module is ISORROPIA II,³³ incorporated into GEOS-Chem by Pye et al.³⁰ We use nested-grid simulations^{34,35} with at 0.5° × 0.67° resolution (55 km × 57 km at 40° N) over North America (10°–70° N, 140°–40° W) and 2° × 2.5° resolution (222 km × 213 km) elsewhere. The chemical mechanism has a temporal resolution of 20 min and a vertical grid of 47 layers extending to 80 km, with 30 layers in the lowest 11 km and 14 layers in the lowest 2 km. Our simulations are driven by GEOS5 meteorology from the NASA Global Modeling and Assimilation Office (GMAO). Each simulation uses meteorology for January or July of 2005, representing winter and summer conditions.

GEOS-Chem v9-02 has three new features relevant to our simulations. First, it includes soil NO_x emissions that respond dynamically to meteorology and nitrogen deposition.³⁶ Second, it limits planetary boundary heights from falling lower than a friction velocity-based minimum. This corrects abnormally low nighttime boundary layers in GEOS5 meteorology compared to observations³⁷ and improves the diurnal variability in simulated chemistry.²⁸ Third, it reduces the rate of production of NO₃⁻ from N₂O₅ hydrolysis, reducing some of the high nitrate bias

seen in GEOS-Chem.^{28,38,39} These features improve simulated NO_x concentrations (especially the daytime-nighttime difference) and reduce the high NO₃⁻ bias noted previously.^{28,29,38}

Emissions. We use anthropogenic emissions from the EDGAR and RETRO global inventories plus several regional inventories (over e.g., China, Europe).⁴⁰ The US inventory is EPA's National Emissions Inventory for 2005 (NEI05). NEI05 provides emissions of NO_x, SO₂, NH₃, several hydrocarbon species, and primary PM. We seasonally adjust NH₃ emissions from NEI05 following Zhang et al.,³⁸ decreasing winter emissions to better match observations, consistent with process-based inventories.⁴¹

We created two groups of simulations, based around a high and low emissions case, to test the influence of large-scale emissions reductions on $PM_{2.5}$ sensitivity. The national, annual total emissions of NO_{xy} SO₂, and NH₃ as reported to the US EPA for 2005 and 2012² exhibit a 42% decrease in NO_{xy} a 62% decrease in SO₂, and a 1% increase in NH₃ emissions. We used these ratios to scale the high emissions case (using NEI05) to the low emissions case. Table 1 shows the resulting total

Table 1. Anthropogenic Emissions (in kilotonnes = 10^6 kg) of Inorganic PM_{2.5} Precursor Emissions over the North American Domain in the Base Simulations of Each Case (High and Low Emissions) and the Changes in Emissions ΔE for Sensitivity Calculations

	E (high)		E (low)		ΔE	
	Jan	Jul	Jan	Jul	Jan	Jul
NO_x	1343.3	1248.9	909.6	855.3	206.5	187.5
SO ₂	849.4	834.2	491.0	484.5	115.6	112.8
NH_3	169.9	548.6	170.7	552.3	15.9	73.8

anthropogenic emissions over the nested-grid domain. CO, VOC, and primary $PM_{2.5}$ emissions in the US also changed by -16%, +1%, and +12% between 2005 and 2012. Primary $PM_{2.5}$ emissions have a direct effect on total $PM_{2.5}$ levels, so sensitivities to primary emissions stay constant. Changes to VOC and CO emissions would have effects on both organic and inorganic $PM_{2.5}$ components. Since organic $PM_{2.5}$ is not included in our study, we did not change VOC or CO emissions, but we do discuss the potential oxidative impact of CO emissions changes.

While our scaling approach matches changes in total emissions, the spatial pattern of sources may have changed as well. Reported NO₂ column densities over major cities in the US are between 24% and 48% lower in 2012 than in 2005,⁴² but NO₂ concentrations in the four quadrants of the US are ~37–40% lower in 2011 than in 2005.⁴³ Russell, Valin, and Cohen⁴⁴ estimate that NO_x emissions from major power plants decreased by 26% between 2005 and 2011, while mobile emissions decreased by 34%. Fioletov et al.⁴⁵ find that SO₂ concentrations over major US power plants are consistently ~40% lower in 2008–2010 than in 2005–2007. This previous work suggests that the total emissions decreases are distributed broadly across the country and across sectors. Hence, our scaling approach approximates the actual emissions changes.

We calculate $PM_{2.5}$ sensitivities as the finite difference in $PM_{2.5}$ concentrations between simulations with emissions slightly increased and decreased around the baseline. Specifically, sensitivity is computed as

$$S(c) = \frac{\mathrm{PM}_{2.5}(E(c) + \Delta E) - \mathrm{PM}_{2.5}(E(c) + \Delta E)}{2 \times \Delta E} \tag{1}$$

where c is the case (high or low emissions) with national-total emissions mass E(c); ΔE is the mass change in emissions; and PM_{2.5}(E) is the PM_{2.5} concentration in the simulation with emissions E. Emissions of other species are fixed at their baseline values for that case. The resulting sensitivities have units of ng m⁻³ kt⁻¹, where kt denotes 1000 t of emissions. The mass changes in emissions for each species and each season are in Table 1.

Calculating sensitivities using mass changes helps clarify the mechanisms that contribute most to sensitivity changes. Since $PM_{2.5}$ is the aggregate of multiple species, normalized (%-based) sensitivities can change between the emissions cases even if oxidation, deposition, and transport processes remain constant. We also use a centered finite difference, as opposed to a one-sided difference that may be more reflective of the effects of a regulation (i.e., a decrease in emissions). This allows our results to be comparable to the adjoint and direct decoupled methods of calculating sensitivities, which also produce centered derivatives.

RESULTS

Model Evaluation. We evaluated model performance by comparing inorganic $PM_{2.5}$ component concentrations in our high emissions case (using NEI05 emissions) to measurements in January and July of 2005 from two monitoring networks: the Interagency Monitoring of Protected Visual Environments (IMPROVE) network⁴⁶ and the EPA Air Quality System (AQS⁴⁷). Here we report the coefficient of determination (squared correlation, r^2) and the normalized mean bias (NMB, model mean over observed mean minus one). The statistics use measured and modeled concentrations paired in both space and time. A more detailed evaluation is provided in the SI (Tables S1 and S2 and Figures S3 and S4).

Our simulation correlates reasonably well ($r^2 > 30\%$) with several measurements: January IMPROVE measurements of all species; both IMPROVE and AQS measurements of July SO₄²⁻; and January AQS measurements of NH₄⁺. Modeled SO₄²⁻ is unbiased in January (NMB < 5%) but slightly low in July (NMB ~ -15%) compared to either set of measurements. Modeled January NH₄⁺ and NO₃⁻ concentrations are biased high (NMB = 86%, 89% respectively, compared to AQS; NMB = 51%, 134% compared to IMPROVE) consistent with previous GEOS-Chem analyses.

Simon, Baker, and Phillips⁴⁸ compare published performance statistics from a range of chemical transport models (not including GEOS-Chem). They find that modeled SO_4^{2-} is unbiased (NMB < 15%), whereas NO_3^- is biased high in winter (NMB 0% to 50%) and low in summer (-15% to -75%). Squared correlations for SO_4^{2-} and NH_4^+ are between 25% and 60%, compared to 10%-45% for NO_3^- . Comparing their results to our statistics indicates that GEOS-Chem has a higher bias in winter NO_3^- than is typical but otherwise performs similar to other models.

Within the northern Midwest – the same region as used for the thermodynamic analysis in Figure 3 – the NMB of modeled NO₃⁻ compared to IMPROVE measurements is 109% and r^2 = 42%, showing that GEOS-Chem estimates NO₃⁻ in this area better than in the national average (NMB = 134%, r^2 = 39%). While sulfate biases are generally smaller than nitrate, modeled SO₄²⁻ is low in this area (NMB = -48%), and a simple scaling to lower model SO_4^{2-} would bring the NO_3^- NMB down to 97%. Modeled NO_3^- is thus better simulated in the region where its behavior is most important to our analysis, detailed below.

PM2.5 Concentrations. Figure 1 shows the total inorganic $PM_{2.5}$ concentrations from the high and low emissions cases. Figure S1 shows the components individually.



Figure 1. Spatial maps of the modeled surface concentrations of inorganic $PM_{2.5}$. Columns show the high and low emissions cases; rows show January and July averages.

January $PM_{2.5}$ peaks in the northern Midwest and is elevated over the eastern US. Northern Midwest $PM_{2.5}$ is primarily composed of NH_4^+ and NO_3^- with low SO_4^{2-} . National average NO_3^- , NH_4^+ , and $PM_{2.5}$ concentrations in the low emissions case are 7.7%, 9.5%, and 11.6% lower than in the high emissions case, respectively. However, these decreases are not uniform across the US. The area around Kentucky, Ohio, and Virginia shows *higher* aerosol NO_3^- in the low emissions case than in the high emissions case. Higher NO_3^- is offset by lower SO_4^{2-} , so total $PM_{2.5}$ concentrations are <0.15 μ g m⁻³ higher in the low emissions case.

Summer PM_{2.5} concentrations are generally lower than in winter but still higher over the eastern US than the western US. The largest mass contribution to summer PM_{2.5} comes from SO_4^{2-} , followed by NH₄⁺. SO_4^{2-} , NH₄⁺, and total PM_{2.5} are all ~40% lower in the low emissions case than in the high emissions case, compared to 62% lower SO₂ emissions. Some urban and regional areas (San Diego, New Orleans, Atlanta, Houston/Eastern Texas, Philadelphia/New Jersey, North Carolina, Illinois-Indiana-Ohio) show substantial summer aerosol NO₃⁻ concentrations, up to 5 μ g m⁻³. Aerosol NO₃⁻ levels in these regions are lower in the low emissions case by 30.5% on average, compared to 42% lower NO_x emissions. However, model-measurement correlations of summer NO₃⁻ concentrations are small ($r^2 = 0.8\%$, 3.2% compared to IMPROVE and AQS, respectively).

PM2.5 Sensitivity to Emissions. Here we describe the sensitivities of $PM_{2.5}$ to the different precursor emissions, shown in Figure 2 and computed using eq 1.

January $PM_{2.5}$ is most sensitive to NH_3 emissions (Figure 2, top panel, bottom row) in both the high and low emissions cases. Sensitivity to NH_3 peaks in the northern Midwest at values of 118 ng m⁻³ kt⁻¹ in the high emissions case. In comparison, maximum winter sensitivities to NO_x and SO_2 are 8 and 10 ng m⁻³ kt⁻¹, respectively. Over the southeastern US and the mid-Atlantic coast, winter sensitivities to SO_2 are prominent and can be >10% of the (local) sensitivity to NH_3 .



Figure 2. Sensitivities of $PM_{2,5}$ concentrations to emissions of the precursors NO_{x^2} SO_{2^2} and NH_{3^2} in units of ng m⁻³ of $PM_{2,5}$ per thousand metric tons (kt) of emissions. The top and bottom panels show sensitivities in January and July, respectively. The columns show the high emissions case, the low emissions case, and their difference.

We find slight (<0.65 ng m⁻³ kt⁻¹) negative sensitivities to winter SO₂ emissions near the eastern Great Lakes.

The differences in January sensitivities between the high and low emissions cases show two policy-relevant features. First, sensitivities to NH₃ are much smaller in the low emissions case: 28% smaller in the national average and up to 72% smaller in the northern Midwest. In contrast, January sensitivities to NO_x are larger in the low emissions case: 202% larger in the national average. Second, average sensitivities to January SO₂ emissions are 42% larger in the low emissions case. There are some local exceptions, such as the northeast US, that show slightly smaller sensitivities in the low emissions case.

In contrast to January, the July sensitivities to different precursors have similar magnitudes. July $PM_{2.5}$ sensitivities to SO_2 emissions (Figure 2, lower panel, middle row) are high across the eastern US and peak in the Ohio River Valley. Sensitivities to SO_2 are larger in the low emissions case than the

high emissions case at nearly every grid point: 23% larger in the national average and up to 80% larger in the eastern US.

In the high emissions case, summer sensitivities to NH_3 emissions are large around the Great Lakes region, eastern Pennsylvania, and a few urban regions. These regions also have high summer sensitivities to NO_x emissions, though the magnitudes are different: peak summer sensitivities to NH_3 and to NO_x in the Great Lakes region are 18.6 and 4.9 ng m⁻³ kt⁻¹, respectively. Summer sensitivities to NH_3 are 38% smaller on average in the low emissions case than the high emissions case.

Summer sensitivities to NO_x emissions are evident around the Great Lakes and a few localized areas on the Atlantic and Gulf coasts. Unlike NH_3 , the sign of changes in sensitivity to summer NO_x emissions varies: the Great Lakes region and some urban areas have larger sensitivities in the low emissions case, whereas the southeastern US has smaller sensitivities. In grid cells where sensitivity to NO_x is larger in the low emissions

case, it is 37% larger on average; in cells where the sensitivity is smaller, it is 29% smaller on average.

UNDERLYING PROCESSES

We investigated several processes that could contribute to differences in $PM_{2.5}$ sensitivities between the high and low emissions cases. In this section, we focus on two particularly important aspects. First, the thermodynamics of ammonium nitrate formation determines winter sensitivities throughout the Midwest. Second, the kinetics of SO₂ oxidation help explain changes in sensitivity to both NO_x and SO₂ emissions across the eastern US in both seasons. In the SI, we describe how well a linear model based on the high emissions case sensitivities predicts the PM_{2.5} concentrations in the low emissions case, showing how sensitivity changes impact the performance of a simplified model.

Thermodynamics of Ammonium Nitrate Formation. Cold and humid environments, such as the northern Midwest in winter, promote condensation of NH₃ and HNO₃ into PM_{2.5}. NH₄⁺ and NO₃⁻ concentrations are high in this region (Figure S1) as are sensitivities to NH₃ and NO_x emissions (Figure 2). The differences in NO_x and NH₃ sensitivities between the high and low emissions cases are highly correlated ($r^2 = 94\%$) and opposite. In addition, PM_{2.5} concentrations in this region are similar in the high and low emissions cases -18 and 16 μ g m⁻³, respectively.

To focus on thermodynamic effects, we used ISORROPIA II,³³ the thermodynamic module in GEOS-Chem, to explore how PM_{2.5} concentrations vary with total available HNO₃ and NH₃. SO₄²⁻ concentrations, temperature, and relative humidity are fixed at their average values between 89 and 97°W and 39.75 and 45.25°N: 1.27 μ g m⁻³, -8.1 °C, and 83%. Figure 3



Figure 3. PM_{2.5} concentrations in the thermodynamic model ISORROPIA using temperature, relative humidity, and SO_4^{2-} concentrations typical of the northern Midwest in January. The dashed lines indicate simulated concentrations in the high and low emissions cases. The solid gray line denotes where there is exactly enough NH₃ to neutralize both SO_4^{2-} and NO_3^{-} .

shows $PM_{2.5}$ under these conditions. There is a line at which the system is neutral with respect to NH_3 , i.e. where there is exactly enough NH_3 to fully neutralize both the SO_4^{2-} and NO_3^- , and this neutrality line distinctly separates nitrate-limited and nitrate-saturated regimes. Below the line, $PM_{2.5}$ concentrations depend strongly on HNO₃ concentrations (nitratelimited); above the line, $PM_{2.5}$ concentrations depend only on NH₃ concentrations (nitrate-saturated). Also shown are the NH₃ and HNO₃ concentrations from the high and low emissions cases, demonstrating why there is a large change in sensitivity: the high emissions case is nitrate-saturated, so that large changes in emissions yielded little change in $PM_{2.5}$, whereas the low emissions case is nitrate-limited, implying far greater sensitivity to NO_x emissions. A broad area in the northern Midwest presents conditions for which the line between nitrate-limited and nitrate-saturated is sharp, and nitrate limitation occurs to some degree across the US in the winter low emissions case.

Kinetics of SO₂ Oxidation. The oxidation rate of SO₂ into SO_4^{2-} determines whether $PM_{2.5}$ forms before SO₂ is transported or deposited and hence directly affects $PM_{2.5}$ sensitivities to SO₂ emissions. SO₂ oxidation occurs in both gas and aqueous phases.⁴⁹ Gas-phase oxidation involves the reaction of SO₂ with hydroxyl radicals (OH). In the aqueous phase, SO₂ diffuses into cloud droplets, where it reacts with other soluble gases (mostly hydrogen peroxide, H₂O₂) to form SO₄²⁻. If cloud droplets evaporate instead of precipitate, SO₄²⁻ remains as PM_{2.5}. GEOS-Chem outputs the production rate of SO₄²⁻ from the gas-phase reaction and from three additional reactions (aqueous H₂O₂) aqueous O₃, and on sea salt particles) individually, allowing the formation pathway to be explored in detail.

The rate and location of SO₂ oxidation depends on whether the OH or H_2O_2 reaction dominates. NO_x concentrations determine relative concentrations of OH and the hydroperoxy radical HO₂, which reacts with itself to form H_2O_2 . Lower NO_x concentrations promote less OH and more H_2O_2 , favoring aqueous SO₂ oxidation. More complex organic peroxy radicals can also produce H_2O_2 , and their concentrations also increase as NO_x concentrations decrease.

We explored oxidation changes in the high and low emissions cases by examining the fraction of oxidation occurring in the aqueous phase (Figure 4). The low emissions case has a larger



Figure 4. Fraction of SO_2 oxidation that occurs through the aqueous H_2O_2 reaction. This fraction is based on the column oxidation, i.e., the amount of oxidation that occurs throughout the entire atmosphere overlying a grid cell.

fraction of aqueous-phase SO_2 oxidation, as expected from lower NO_x concentrations. In January, the largest differences are in the northern and northeastern US, where aqueous oxidation is 50% of total oxidation in the high emissions case and 70% in the low emissions case. In July, the largest differences are in the Ohio River Valley: 35% aqueous oxidation in the high emissions case and 55% in the low emissions case. In addition, H_2O_2 concentrations in the eastern US (east of 100° W) are 10.6% higher in the low emissions case than in the high emissions case and OH concentrations are 10.7% lower, supporting the link between NO_x emissions and SO₂ oxidation.

We also investigated whether increased aqueous-phase oxidation would lead to faster SO₂ and SO₄²⁻ rainout due to more sulfur chemistry occurring within cloud droplets. The wet deposition rate (units s⁻¹) is a measure of the speed of rainout (calculation details in the SI). Wet deposition rates are larger in the low emissions case than in the high emissions case by about 5%, compared to 50% larger rates of aqueous SO₂ oxidation. In addition, the differences in PM_{2.5} sensitivities to SO₂ emissions between the high and low emissions cases are spatially correlated ($r^2 = 57\%$) with the fraction of aqueous-phase oxidation. Thus, while faster rainout occurs under low NO_x emissions, it cannot compensate for the increase in aqueous oxidation.

We did not include the 16% decrease in CO emissions between 2005 and 2012 in our simulations. CO reacts with OH to form HO₂ as the counterpart to NO + HO₂ \rightarrow NO₂ + OH. Lower CO emissions would lead to a larger OH/HO₂ ratio, less H₂O₂ production, and more gaseous SO₂ oxidation. Thus, lower CO emissions could partially offset the shift to more aqueous-phase SO₂ oxidation in our simulations. However, Duncan et al.⁵⁰ suggest that much of the US is now in a NO_xlimited ozone formation regime and hence that NO_x exerts more control on HO_x partitioning (and thus the SO₂ oxidation pathway) than CO does.

DISCUSSION

Our study shows large differences in the sensitivities of $PM_{2.5}$ concentrations to precursor emissions between two sets of simulations representing a 2005 baseline (high emissions) and a 2012 analogue (low emissions). We find that winter NO_x reductions represent a potential new opportunity for improving air quality, due to $PM_{2.5}$ being more nitrate-limited under low emissions over much of the US, particularly the Midwest. Lower NO_x emissions also promote aqueous-phase SO_2 oxidation, increasing the sensitivity of $PM_{2.5}$ to SO_2 . Sensitivities to NH_3 emissions are lower in the low emissions case, primarily as a direct response to a less acidic atmosphere.

Results for winter in the northern Midwest are driven by the thermodynamic behavior of ammonium nitrate aerosols and are well-constrained for the meteorological conditions (i.e., cold and moist) that prevail there, where concentrations are highest. Figure 3 shows that nitrate availability will play a major role in determining PM_{2.5} in this region in the near future. Accounting for the model's high NO3⁻ bias can only push the system further into the nitrate-limited regime. Nevertheless, the large absolute sensitivities to winter NO_x emissions through NO_3^- formation are subject to the model bias. Several studies^{28,29,38} have shown that the standard GEOS-Chem simulation overestimates HNO₃ and aerosol NO₃⁻ concentrations compared to both CASTNet and AQS measurements. There is evidence that certain types of NO₃⁻ measurements are biased low due to HNO3 volatilization from filters,⁵¹⁻⁵³ but adjusting for this does not always provide significant improvement.⁵ Studies with other air quality models (notably CAMx⁵ and CMAQ^{55,56}) have emphasized the potential impact of NH₃ emissions controls on PM2.5 concentrations, suggesting that our results are broadly consistent across models.

There are several possible sources of the nitrate bias in GEOS-Chem and other chemical transport models. The dependence of the rate of N_2O_5 hydrolysis on aerosol water, nitrate, chloride, and organic content is uncertain, and nitric

acid concentrations can be highly sensitive to the overall hydrolysis rate.^{39,56–58} Hydrolysis within GEOS-Chem includes limitation by nitrate, but this is a topic of ongoing research.^{57–61} GEOS-Chem produces more HNO₃ from organic species reactions than other chemical mechanisms at low NO_x concentrations but performs comparably at high NO_x.⁶² Heald et al.²⁹ explore several other potential sources of bias, but no solution has yet been found. However, our thermodynamic analysis increases our confidence that the importance of NO_x emissions to winter PM_{2.5} is not an artifact of this model bias.

Model resolution can also affect model performance. Our simulations have a resolution of ~55 km, compared to the 36 or 12 km resolution typical of regulatory models. Li et al.⁶³ show differences in GEOS-Chem simulations of US PM25 at the nested and global resolutions. The maximum values for the three inorganic components are lower at low resolution, but the largest change (29%) is for SO_4^{2-} and changes are most substantial near cities. Thompson, Saari, and Selin⁶⁴ show that PM_{2.5} within CAMx changes by10% across model resolutions from 4 km to 36 km. Punger and West⁶⁵ found that CMAQ simulations produce PM_{2.5} mortality estimates 11% higher at 36 km resolution than at 12 km. Much coarser resolutions (>100 km) show lower $PM_{2.5}$ concentrations, but the relative changes are much larger for other components of PM2.5 than nitrate, sulfate, and ammonia. Our 55 km resolution is therefore sufficient for studying the regional response of inorganic PM₂₅ to large, nation-wide changes in emissions, and the computational efficiency of the lower resolution allowed us to explore sensitivities (requiring several simulations for each case).

The change in sensitivity to NH_3 emissions has several implications. First, NH_3 emissions controls have been identified as a potentially cost-effective way to improve air quality.⁵ We do not analyze the costs of emissions controls, though the costs of SO_2 and NO_x controls have likely changed from the redistribution of sources, but the impacts of NH_3 emissions controls could be much smaller than previously estimated. Second, previous studies comparing modeled and measured $PM_{2.5}$ in the US^{7,29,38,56} have highlighted our generally poor understanding of the magnitude and seasonality of NH_3 emissions. Decreased sensitivity to NH_3 would limit the adverse effects of inaccurate emissions on model performance.

An alternative approach to our sensitivity analysis is to vary emissions based on economic sector (e.g., Caiazzo et al.¹³). However, simultaneous emissions changes in multiple sectors will not have the impact on $PM_{2.5}$ expected from changes in each sector individually. The changes in sensitivities presented here will help identify which sectors could be expected to have strong interactions. For example, broad agricultural NH₃ and NO_x emissions can determine the neutralizing and oxidizing capacity of the background atmosphere and hence the impact of given coal SO₂ emissions on $PM_{2.5}$.

Through this analysis, we find that lower NO_x and SO_2 emissions lead to larger sensitivity to SO_{2j} smaller sensitivity to NH_{3j} and larger sensitivity to winter NO_x emissions in the US. These interactions provide new avenues for effective air quality regulations and emphasize the need to consider multiple pollutants simultaneously.

ASSOCIATED CONTENT

S Supporting Information

 $PM_{2.5}$ component concentration maps, a detailed modelmeasurement comparison, further analysis of the SO₂ oxidation

pathways, and a comparison of a linear extrapolation to the full model. This material is available free of charge via the Internet at http://pubs.acs.org.

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Notes

The authors declare no competing financial interest.

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