

## ***Interactive comment on “Significant wintertime PM<sub>2.5</sub> mitigation in the Yangtze River Delta, China from 2016 to 2019: observational constraints on anthropogenic emission controls” by Liqiang Wang et al.***

**Liqiang Wang et al.**

shaocaiyu@zju.edu.cn

Received and published: 8 September 2020

General comments: Wang et al. evaluated the effect of long-term and emergent emission control strategies on the PM<sub>2.5</sub> levels in Yangtze River Delta of China, by combining modelling analysis with observations. They found the decline in PM<sub>2.5</sub> concentration during 2016–2019 was mainly due to emission control. The decline would be even greater if the meteorology was not unfavorable. Great potential of further decrease is manifested in analysis of data during G20 period when short-term emergent measures were taken. The discussion is valuable for assessment of past policies and design of

C1

future ones. However, to be publishable in ACP, the current manuscript requires further improvement.

Response: We thank the reviewer for the thoughtful comments on our paper and have addressed them as below.

1. Particularly, inadequate credits are given to the existing literature that performed similar analysis of separating meteorology and emission effects on recent PM<sub>2.5</sub> trend in China. Instead, the authors tried to impress the reader by suggesting that this study is the first to do so. Just to list a few studies in literature (and I believe there might be more), Zhang et al., 2019. Drivers of improved PM<sub>2.5</sub> air quality in China from 2013 to 2017, PNAS; Zhai et al., 2019. Fine particulate matter (PM<sub>2.5</sub>) trends in China, 2013–2018: separating contributions from anthropogenic emissions and meteorology, ACP; Zhong et al., 2018. Distinguishing Emission-Associated Ambient Air PM<sub>2.5</sub> Concentrations and Meteorological Factor-Induced Fluctuations, EST. The authors should review the existing literature and put emphasize on their innovations.

Response: We thank the reviewer for the valuable suggestions on the introduction. We have supplemented the additional discussions to introduce more literature, including all of the above works as well as someone else, in order to substantiate and improve the “Introduction”. Added/rewritten part in Sect. 1: The main challenge involves reliably representing substantial and rapid changes in anthropogenic emissions resulting from both long-term and emergency control measures (Chen et al., 2019; Cheng et al., 2019; Zhang et al., 2014; Yang et al., 2016; Zhai et al., 2019; Zhang et al., 2019; Zhong et al., 2018). To gain timely insight into variations in anthropogenic emissions, considerable efforts went into establishing detailed bottom-up emissions and derived valuable findings (Cheng et al., 2019; Zhang et al., 2019). Yet bottom-up inventories were built on the basis of activity data and emission factors. These input data can be absent or outdated, likely leading to misunderstandings of anthropogenic impacts, particularly in terms of the magnitude (Jiang et al., 2018). Recent studies applied available observations to construct multilinear regression models (emission-based or

C2

meteorology-related), allowing us to separate contributions from anthropogenic emissions and meteorology to some extent (Zhai et al., 2019; Zhong et al., 2018). However, the uncertainties in bottom-up inventories and meteorological fields remained. Here we switched to observational constraints on a state-of-the-art chemical model. This can be a potential way to tackle this challenge.

2. I also have concerns about the methodology. Assimilation is used for calculating the total effect (emis+met) which gives a good representation of PM<sub>2.5</sub> distributions, despite any model errors. But assimilation cannot be used for calculating met-only effect. Therefore, model errors may propagate into the met effect. I wonder what uncertainties this inconsistency in two pairs of simulations would cause for the results. The authors evaluated model emissions and concluded the impact is small. But it is not shown if other model errors may be significant. For example, studies have found that model tends to underestimate sulfate production during high RH in China. More evaluation of the model performance may be useful for interpreting the result.

Response: Thanks. Yes, we agree with the reviewer that more detailed model evaluation might be useful for this study. As previous studies have demonstrated (Cheng et al., 2019; Zhai et al., 2019; Zhong et al., 2018), model uncertainties remain, although we have verified the constrained results. We have supplemented the additional discussions in Sect. 4 for further explanations. For instance, model simulations of aerosol components (e.g., sulfate and nitrate) are still poorly constrained. Moreover, they have not been evaluated due to lack of available observations. Previous studies find that the model tends to underestimate sulfate production during high RH (as pointed by the reviewer) as well as SOA (Li et al., 2017a; Wang et al., 2014; Zhong et al., 2018). As a result, these uncertainties can be propagated into the estimations of meteorological effects. In addition, like other atmospheric chemical transport models, the WRF-CMAQ model cannot provide model uncertainty information in the simulations, while Monte Carlo simulations for complex CTMs would be unrealistic due to extremely high computation loadings (Zhong et al., 2018). Looking forward, more detailed model evalu-

C3

ations, as well as more explicit observational constraints, are of great significance for improving associated understandings, which will be the topic of a next separate study. Added/rewritten part in Sect. 4: Note that, as previous studies have demonstrated (Cheng et al., 2019; Zhai et al., 2019; Zhong et al., 2018), model uncertainties remain, although we have verified the constrained results. We have supplemented the additional discussions in Sect. 4 for further explanation. For instance, model simulations of aerosol components (e.g., sulfate and nitrate) are still poorly constrained. Moreover, they have not been evaluated due to lack of available observations. Yet previous studies find that the model tends to underestimate sulfate production during high RH and SOA (Li et al., 2017a; Wang et al., 2014; Zhong et al., 2018). As a result, these uncertainties can be propagated into the estimations of meteorological effects. Besides, like other atmospheric chemical transport models, the WRF-CMAQ model cannot provide model uncertainty information in the simulations, while Monte Carlo simulations for complex CTMs would be unrealistic due to extremely high computation loadings (Zhong et al., 2018).

3. The inclusion of short-term G20 period is interesting. But I am not completely convinced that the mitigation potential map is useful at all. At a first glance, the map does not seem to be very different from conducting a zero-ish YRD emission simulation with the model and then do a subtraction. The problem is that the authors did not provide information about (1) what types or fractions of emissions were shut down during the event; (2) is the emission shutdown implemented in Hangzhou, or Zhejiang, or YRD? Without this information, it is not possible to interpret the mitigation potential.

Response: Thanks. We have supplemented the information associated with anthropogenic emission control measures during the G20 summit. On that occasion, anthropogenic emission controls were conducted across the whole YRD (including Zhejiang, Jiangsu, and Anhui provinces, and Shanghai municipality), particularly in Hangzhou that served as the host city (Li et al., 2019, 2017b; Ni et al., 2020; Yu et al., 2018). Li et al. (2017) showed that most of anthropogenic emissions (e.g., those from in-

C4

dustry, power plant, residential, and on-road transportation sectors) were reduced by around 50% on the basis of available governmental information. Added/rewritten part in Sect. 1: Those measures were conducted across the whole YRD (including Zhejiang, Jiangsu, and Anhui provinces, and Shanghai municipality), particularly in Hangzhou that served as the host city (Li et al., 2019, 2017b; Ni et al., 2020; Yu et al., 2018). Li et al. (2017) showed that most of anthropogenic emissions (e.g., those from industry, power plant, residential, and on-road transportation sectors) were reduced by around 50% on the basis of available governmental information.

Specific comments: 1. Line 41: Not clear from the text whether “> 14  $\mu\text{g}/\text{m}^3$ , 19%” is PM2.5 levels, or in fact, reduction in PM2.5 concentrations. Please clarify.

Response: Thanks. We have deleted the numbers. Specific numbers have been given in the following part of “Abstract”.

2. Line 42-44: Confusing, as it interrupts the flow and misleads a reader that the decline in Hangzhou (35  $\mu\text{g}/\text{m}^3$ ) is due to G20 control measures. I suggest moving the sentence to either Line 40 after “YRD, China” or to Line 48 before “Compared to the long-term policies...”.

Response: Thanks. We suggest not to move this sentence. It follows behind the statement “For the winter time periods from 2016 to 2019” and is thus linked with the effects of the long-term policies from 2016 to 2019.

3. Line 46: remove “thus”

Response: Thanks. We have removed the word.

4. Line 99: should -> can

Response: Thanks. We have revised the word accordingly.

5. Line 105: “unprecedented” is a too-big word here that I suggest to remove. Same for other occurrences of the word in the paper.

C5

Response: Thanks. We have removed the word throughout the paper.

6. Line 125: Be consistent with the citation format.

Response: Thanks. We have revised the format.

7. Line 137-139: Are meteorological observations assimilated in addition to chemical observations? If so, describe the meteorological observations that are assimilated. If not, I don't think it is sufficient to just use initial and boundary conditions from reanalysis data. The WRF should be run in a nudging mode, so the meteorology is close to reality.

Response: Thanks. Yes, the ECMWF reanalysis datasets were used to constrain meteorological simulations. Therein almost all necessary meteorological factors (nine variables), involving temperature, U wind component, V wind component, pressure, relative humidity, precipitation, short-wave radiation, cloud cover, and planetary boundary layer height (PBLH), were assimilated (<https://apps.ecmwf.int/datasets/data/interim-full-daily/levtype=sfc/>, last access: 7 March 2020). Added/rewritten part in Sect. 2.4: The ECMWF reanalysis datasets accounted for the hourly observational constraints on spatiotemporal meteorological evolutions. Therein almost all necessary meteorological factors (9 variables), involving temperature, U wind component, V wind component, pressure, relative humidity, precipitation, short-wave radiation, cloud cover, and planetary boundary layer height (PBLH), were assimilated (<https://apps.ecmwf.int/datasets/data/interim-full-daily/levtype=sfc/>, last access: 7 March 2020).

8. Line 155: “Prior anthropogenic emissions”? Do you optimize emissions at all? I cannot find such description throughout the text. If not, it should not be called “prior”.

Response: Thanks. We have revised the word accordingly.

9. Line 166: “more” is not a proper conjunction word in formal English writing.

Response: Thanks. We have revised the word to “Moreover”.

C6

10. Line 176: grids -> grid cells

Response: Thanks. We have revised the word accordingly.

11. Line 176: "potential excellent roles" Rephrase it.

Response: Thanks. We have revised the phrase to "potentially excellent roles".

12. Section 2.4: What is the assimilation window? Daily? hourly?

Response: Thanks. The assimilation window is hourly. We have highlighted this in the Sect. 2.4. Added/rewritten part in Sect. 2.4: When ground-level PM<sub>2.5</sub> measurements were assimilated, hourly observations were put into equation (1) to construct the new analysis fields. All-day state variables associated with aerosols in the model were adjusted from their background (simulated) to their analysis (constrained) states using the scaling factors ( $X^a/X^b$ ).

13. Line 212: "the threshold pinpointing the key value of the correlation coefficients ( $e^{-1}$ )". -> e-folding length

Response: Thanks. We have revised the sentence. Added/rewritten part in Sect. 2.3: The results indicated that a correlation length scale of  $\sim 180$  km could be treated as the threshold. It allowed the correlation coefficients to fall within the range of  $e^{-1}$ , defining the effective radius of each individual observation.

14. Line 217-219: well, it is still static in the relative sense. I don't this means anything. I'd suggest removing this statement.

Response: Thanks. We have removed the sentence.

15. Line 231: "unify the chemical inputs for the WRF-CMAQ model". What does this mean?

Response: Thanks. We have revised the sentence to further clarify the meaning. Added/rewritten part in Sect. 2.4: These configurations unified both chemical (i.e.,

C7

emission inventories) and meteorological input data for the WRF-CMAQ model.

16. Line 237-240: Up to this point, we still do not know what method the authors use to separate effects of meteorology and emission. A clear description of the method is needed before this point.

Response: Thanks. We have revised the sentence to further clarify the meaning. Added/rewritten part in Sect. 2.4: Specifically, the differences in the constrained PM<sub>2.5</sub> concentrations between DA\_2016 and DA\_2019 reflected the net effects of anthropogenic emission controls and meteorological perturbations between 2016 and 2019, while meteorological impacts therein were calculated as the differences in simulated PM<sub>2.5</sub> concentrations between NO\_2016 and NO\_2019 (Chen et al., 2019a). Hence, by subtracting meteorological impacts from the net effects, we can isolate the effects of anthropogenic emission controls attributable to the long-term strategies.

17. Line 257: It is cursory to conclude these three periods have similar meteorology based on Fig. S1. The validity of the analysis is relied on the assumption that they are similar. E.g., one factor that is not analyzed is wind direction. Showing maps of circulation pattern will also help.

Response: Thanks. We have supplemented the map of atmospheric synoptic circulation patterns in Figure S1 accordingly (Dong et al., 2020; Liu et al., 2019).

18. Fig. 4. Is Fig. 4 useful? It is no surprising that the assimilated simulation could better reproduce observations, which are used in assimilation. It means nothing.

Response: Thanks. This figure is used to verify whether the model coupled with the OI method could reproduce the measurements. While 244 monitoring stations reside in 6660 grid cells, 16 grid cells have two to three monitors in them. For these grid cells, only one averaged measurement was used for DA. However, all the observations were compared against the constrained results. Hence, we suggest not to remove the figure and have supplemented the additional discussions in the Sect. 2.4. Added/rewritten

C8

part in Sect. 2.4: While 244 monitoring stations reside in 6660 grid cells, 16 grid cells have two to three monitors in them. For these grid cells, only one averaged measurement was used for DA. However, all the observations were compared against the constrained results in the analyses.

19. Line 307-311. There is a jump in the logic of this sentence. I'd remove it.

Response: Thanks. We have removed the sentence.

20. Line 350-352: Many studies have properly separated the effects from meteorology and emissions, though with different approaches. I don't think the statement is fair.

Response: Thanks. We have revised the sentence. Added/rewritten part in Sect. 3.2: This also indirectly implied the importance of assimilating meteorology, which, however, were generally neglected by previous studies (Chen et al., 2019).

21. Line 353: It's unclear to me whether 5% and 3% are relative to mean PM2.5 concentration or mean reduction of PM2.5. Be more explicit.

Response: Thanks. We have added the absolute concentrations to make it clear. Added/rewritten part in Sect. 3.2: As shown in Figure S10 and Figure 5, even with the largest adjustment (i.e., -40%), such interferences could be well controlled within the 5% ( $< 3 \mu\text{g}/\text{m}^3$ ) scope, let alone other tests (i.e.,  $< 3\%$ ,  $< 2 \mu\text{g}/\text{m}^3$ ).

22. Line 355: how do you prove it was "under the same meteorological condition".

Response: Thanks. We have revised the sentence to further clarify the meaning. Added/rewritten part in Sect. 3.2: Moreover, these findings are consistent with previous analyses (Cheng et al., 2019; Zhang et al., 2019). They generally revealed that reasonable changes in the bottom-up emissions, together with the same meteorology input data, would not remarkably alter the simulated results associated with meteorological effects on surface PM2.5 ( $< 5\%$ ).

23. Line 370: what is "concurrent meteorology"? better to rephrase it.

C9

Response: Thanks. We have revised the phrase to "meteorological conditions therein".

24. Line 375-376: Of course, the long-term strategies are emission oriented. You cannot change weather easily after all ... I guess the author wanted to say the long-term decrease in PM2.5 was driven mainly by decreased emissions.

Response: Thanks. We agree. We have revised the sentence. Added/rewritten part in Sect. 3.2: This indicates that the impacts of the long-term strategies are mainly driven by anthropogenic emission mitigation.

25. Line 387: what is "stably spatiotemporal state". Rephrase it.

Response: Thanks. We have revised the sentence. Added/rewritten part in Sect. 3.3: We found that such impacts were of relatively low standard deviations ( $< 5\%$ ) and kept stable over time.

26. Line 412: remove "unprecedented".

Response: Thanks. We have removed the word throughout the paper.

27. Line 414: what is "stable supply-side structures"? Not directly related to air quality to me.

Response: Thanks. We have revised the phrase to "stable structures of anthropogenic emissions".

28. Line 418: There have been quite a few papers discussing the effect of emission control strategy. To list a few, Zhang et al., 2019. Drivers of improved PM2.5 air quality in China from 2013 to 2017, PNAS Zhai et al., 2019. Fine particulate matter (PM2.5) trends in China, 2013–2018: separating contributions from anthropogenic emissions and meteorology, ACP Zhong et al., 2018. Distinguishing Emission-Associated Ambient Air PM2.5 Concentrations and Meteorological Factor-Induced Fluctuations, EST Although the method may not be the same, the authors should give credit to these studies rather than claim this is the first study trying to separate effects of met and

C10

emission.

Response: Thanks. We have addressed this issue in the response for the general comment (1). We have supplemented the additional discussions to introduce more literature, including all of the above works as well as someone else, in order to substantiate and improve the “Introduction”. Added/rewritten part in Sect. 1: The main challenge involves reliably representing substantial and rapid changes in anthropogenic emissions resulting from both long-term and emergency control measures (Chen et al., 2019; Cheng et al., 2019; Zhang et al., 2014; Yang et al., 2016; Zhai et al., 2019; Zhang et al., 2019; Zhong et al., 2018). To gain timely insight into variations in anthropogenic emissions, considerable efforts went into establishing detailed bottom-up emissions and derived valuable findings (Cheng et al., 2019; Zhang et al., 2019). Yet bottom-up inventories were built on the basis of activity data and emission factors. These input data can be absent or outdated, likely leading to misunderstandings of anthropogenic impacts, particularly in terms of the magnitude (Jiang et al., 2018). Recent studies applied available observations to construct multilinear regression models (emission-based or meteorology-related), allowing us to separate contributions from anthropogenic emissions and meteorology to some extent (Zhai et al., 2019; Zhong et al., 2018). However, the uncertainties in bottom-up inventories and meteorological fields remained. Here we switched to observational constraints on a state-of-the-art chemical model. This can be a potential way to tackle this challenge.

29. Line 444: “rudimentary” may not be a proper word here.

Response: Thanks. We have removed the word.

30. Line 449: the statement that “the biogenic emissions are unimportant for IAV of PM<sub>2.5</sub>” may be true for YRD, but may not be “generally” true for elsewhere in the world. I’d suggest being more specific.

Response: Thanks. We have revised the sentence to make it more specific. Added/rewritten part in Sect. 4: Moreover, the former is generally of minor significance

C11

for interannual PM<sub>2.5</sub> variations for the YRD.

References: Chen, D., Liu, Z., Ban, J., Zhao, P. and Chen, M.: Retrospective analysis of 2015–2017 wintertime PM 2.5 in China: response to emission regulations and the role of meteorology, *Atmos. Chem. Phys.*, 19(11), 7409–7427, 2019a.

Chen, D., Liu, Z., Ban, J., Zhao, P. and Chen, M.: Retrospective analysis of 2015–2017 wintertime PM<sub>2.5</sub> in China: response to emission regulations and the role of meteorology, *Atmos. Chem. Phys.*, 19(11), 7409–7427, 2019b.

Chen, D., Liu, Z., Ban, J. and Chen, M.: The 2015 and 2016 wintertime air pollution in China: SO<sub>2</sub> emission changes derived from a WRF-Chem/EnKF coupled data assimilation system, *Atmos. Chem. Phys.*, 19(13), 8619–8650, 2019c.

Cheng, J., Su, J., Cui, T., Li, X., Dong, X., Sun, F., Yang, Y., Tong, D., Zheng, Y., Li, Y. and others: Dominant role of emission reduction in PM<sub>2.5</sub> air quality improvement in Beijing during 2013–2017: A model-based decomposition analysis, *Atmos. Chem. Phys.*, 19(9), 6125–6146, 2019.

Dong, Y., Li, J., Guo, J., Jiang, Z., Chu, Y., Chang, L., Yang, Y. and Liao, H.: The impact of synoptic patterns on summertime ozone pollution in the North China Plain, *Sci. Total Environ.*, 735, 139559, doi:<https://doi.org/10.1016/j.scitotenv.2020.139559>, 2020.

Jiang, Z., McDonald, B. C., Worden, H., Worden, J. R., Miyazaki, K., Qu, Z., Henze, D. K., Jones, D. B. A., Arellano, A. F., Fischer, E. V and others: Unexpected slowdown of US pollutant emission reduction in the past decade, *Proc. Natl. Acad. Sci.*, 115(20), 5099–5104, 2018.

Li, G., Bei, N., Cao, J., Huang, R., Wu, J., Feng, T., Wang, Y., Liu, S., Zhang, Q., Tie, X. and Molina, L. T.: A possible pathway for rapid growth of sulfate during haze days in China, *Atmos. Chem. Phys.*, 17(5), 3301–3316, doi:[10.5194/acp-17-3301-2017](https://doi.org/10.5194/acp-17-3301-2017), 2017a.

Li, H., Wang, D., Cui, L., Gao, Y., Huo, J., Wang, X., Zhang, Z., Tan, Y., Huang, Y.,

C12

Cao, J. and others: Characteristics of atmospheric PM<sub>2.5</sub> composition during the implementation of stringent pollution control measures in Shanghai for the 2016 G20 summit, *Sci. Total Environ.*, 648, 1121–1129, 2019.

Li, P., Wang, L., Guo, P., Yu, S., Mehmood, K., Wang, S., Liu, W., Seinfeld, J. H., Zhang, Y., Wong, D. C. and others: High reduction of ozone and particulate matter during the 2016 G-20 summit in Hangzhou by forced emission controls of industry and traffic, *Environ. Chem. Lett.*, 15(4), 709–715, 2017.

Liu, N., Zhou, S., Liu, C. and Guo, J.: Synoptic circulation pattern and boundary layer structure associated with PM<sub>2.5</sub> during wintertime haze pollution episodes in Shanghai, *Atmos. Res.*, 228, 186–195, doi:<https://doi.org/10.1016/j.atmosres.2019.06.001>, 2019.

Ni, Z.-Z., Luo, K., Gao, Y., Gao, X., Jiang, F., Huang, C., Fan, J.-R., Fu, J. S. and Chen, C.-H.: Spatial-temporal variations and process analysis of O<sub>3</sub> pollution in Hangzhou during the G20 summit, *Atmos. Chem. Phys.*, 20(10), 5963–5976, doi:10.5194/acp-20-5963-2020, 2020.

Wang, Y., Zhang, Q., Jiang, J., Zhou, W., Wang, B., He, K., Duan, F., Zhang, Q., Philip, S. and Xie, Y.: Enhanced sulfate formation during China's severe winter haze episode in January 2013 missing from current models, *J. Geophys. Res. Atmos.*, 119(17), 10,410–425,440, doi:10.1002/2013JD021426, 2014.

Yang, Y., Liao, H. and Lou, S.: Increase in winter haze over eastern China in recent decades: Roles of variations in meteorological parameters and anthropogenic emissions, *J. Geophys. Res. Atmos.*, 121(21), 13,13–50,65, doi:10.1002/2016JD025136, 2016.

Yu, H., Dai, W., Ren, L., Liu, D., Yan, X., Xiao, H., He, J. and Xu, H.: The effect of emission control on the submicron particulate matter size distribution in Hangzhou during the 2016 G20 Summit, *Aerosol Air Qual. Res.*, 18, 2038–2046, 2018.

C13

Zhai, S., Jacob, D. J., Wang, X., Shen, L., Li, K., Zhang, Y., Gui, K., Zhao, T. and Liao, H.: Fine particulate matter (PM<sub>2.5</sub>) trends in China, 2013–2018: separating contributions from anthropogenic emissions and meteorology, *Atmos. Chem. Phys.*, 19(16), 11031–11041, doi:10.5194/acp-19-11031-2019, 2019.

Zhang, R., Li, Q. and Zhang, R.: Meteorological conditions for the persistent severe fog and haze event over eastern China in January 2013, *Sci. China Earth Sci.*, 57(1), 26–35, doi:10.1007/s11430-013-4774-3, 2014.

Zhang, Q., Zheng, Y., Tong, D., Shao, M., Wang, S., Zhang, Y., Xu, X., Wang, J., He, H., Liu, W. and others: Drivers of improved PM<sub>2.5</sub> air quality in China from 2013 to 2017, *Proc. Natl. Acad. Sci.*, 116, 24463–24469, 2019.

Zhong, Q., Ma, J., Shen, G., Shen, H., Zhu, X., Yun, X., Meng, W., Cheng, H., Liu, J., Li, B., Wang, X., Zeng, E. Y., Guan, D. and Tao, S.: Distinguishing emission-associated ambient air PM<sub>2.5</sub> concentrations and meteorological factor-induced fluctuations, *Environ. Sci. Technol.*, 52(18), 10416–10425, doi:10.1021/acs.est.8b02685, 2018.

---

Interactive comment on *Atmos. Chem. Phys. Discuss.*, <https://doi.org/10.5194/acp-2020-510>, 2020.

C14