Intercomparison of air quality simulations in 2019 using three chemical transport models



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Introduction

- Exposure to ambient PM_{25} caused about 2.9 to 4.2 million deaths from 2010 to 2017 and air pollution was ranked the fifth among the global mortality risk factors (Cohen et al., 2017; HEI, 2019), the mortality attributable to air pollution is expected to continue to dominate in Asia at about 75% among the mortality factors by 2050 (Lelieveld et al., 2015).
- Major East Asian countries have implemented strong air pollution control policies over the past decades, the concentration level is still higher than these in other areas or compared to the internationally recommended values (Zhang et al., 2019; Trnka, 2020; Bae and Kim, 2022).
- Chemical transport model (CTM) can be a powerful tool for understanding the physical and chemical processes that affect air pollution, as an approach using multiple models can increase the reliability of the model, it is widely used to understand air pollution problems.
- However, existing CTMs have their own inherent bias errors in the model system, model results are also different depending on the region or period. Therefore, it can be seen that a common understanding of the performance and uncertainty of each model and further efforts to improve the modeling system are important.
- In this study, we compared the air quality simulation results by CMAQ, WRF-Chem and GEOS-Chem models in Northeast Asia including China and Korea for the months representing each season in 2019. And we also analyzed the factors that show performance differences between models.



[Modeling domains and observation site]

[Observation locations]

		Site name	Longitude	Latitude
GTS & ASOS	China	Beijing	116.283	39.933
		Baoding	115.567	38.85
		Dalian	121.633	38.9
		Changdao	120.717	37.9333
	Korea	Baengnyung 124.712		37.974
		Seoul	126.966	37.571
Intensive monitoring sites	China	Beijing	40.042	116.419
		Baoding	38.858	115.486
		Dalian	38.886	121.577
		Changdao	38.189	120.746
	Korea	Baengnyung	37.9647	124.634
		Seoul	37.6098	126.9348

• Data for PM_{2.5} for the 1467 China sites (CNEMC) and the 295 Korean sites (NIER) at surface, and for air pollutants (Sulfate (SO_4^{2-}) , Nitrate (NO_3^-) , Ammonium (NH_4^+) , Organic

[Input options of participating models]					
	CMAQ	WRF-Chem	GEOS-Chem		
Version	5.2	3.9.1	13.3.3		
Horizontal Resolution	9 X 9 km	9 X 9 km	0.25° X 0.3125°		
Grid number	431 X 394	451 X 428	129 X 121		
Vertical levels	23	43	47		
Boundary layer	YSU	YSU	Nonlocal PBL mixing		
Anthropogenic emissions		UNIMIX2019			
Biogenic emission	MEGANv2.1	MEGANv2.0	MEGANv2.1		
Aerosol	AERO6	4 bins-MADE-VBS	Hodzic+aging		
Gaseous chemical mechanism	CB6R3	RACM	GEOS-Chem v13.3.3		
Biomass burning emission	Not used	Not used	GFED4		
CREATEC PM2.5 (averaged for May–June 2016)	UNIMIX - CREATEC dPM2.5 (averaged tor May-June 2016)	CREATEC SO2 (averaged for May–June 2016)	UNIMIX - CREATEC dSO2 (averaged for May-June 2016)		
134 300 12 127 222 11 106 225 9 92 188 77 71 6 150 76 72 75 75 75 73 75 75 75		300 262 225 188 99 190 112 190 190 190 190 190 190 190 190 190 190	114 127 100 113 106 92 85 77 71 44 57 70 44 57 70 44 57 70 44 57 70 44 57 70 44 57 70 44 50 70 70 70 70 70 70 70 70 70 70 70 70 70		



[UNIMIX2019 emission inventory]

aerosol (OA), and Elemental carbon (EC)) at intensive monitoring sites were used for model evaluation. Three chemical transport models (CTMs) participated in this study and were used to conduct air quality simulations focusing on PM2.5 and major chemical compositions for January, April, July, and October, 2019 representing the season

The modeling emission inventory (UNIMIX2019), which reflects the 2019 air pollution emission from anthropogenic sources in China and Korea, was used for all three CTMs to increase the reliability of the multi-model experiment in 2019.

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[Timeseries between simulated and observed meteorological variables at near-surface]

Result 2 : Evaluation of PM_{2.5} surface concentration

• Observed and Simulated monthly PM_{2.5} concentration at the surface



- In January, the NMBs of CMAQ, WRF-Chem, and GEOS-Chem were 28.08%, 6.58%, and 9.79%, respectively, with CMAQ showing the largest overestimation.
- In July, the bias of WRF-Chem was about 3 times higher than that of other models. In October, performance of PM_{2.5} simulated by CMAQ was better than others. The bias of GEOS-Chem was relatively large in mainland China around Shandong Province and Korea.



Results

• As a result of the analysis of ground weather maps, it was found that in July 2019, there was a significant presence of clouds, haze, and precipitation.

 The evaluation of the meteorological fields was conducted at intensive monitoring sites.

• For wind speed, the GEOS-Chem model tended to overestimate, and a relatively low correlation coefficient was observed in July.

			R	NMB	RMSE
Wind speed	China	M1	0.90	11.18	0.66
		M2	0.92	22.28	0.85
		M3	0.92	80.65	2.64
	Korea	M1	0.84	27.26	1.38
		M2	0.82	57.26	2.29
		M3	0.78	114.53	4.09
Temperature	China	M1	0.99	4.43	1.81
		M2	0.99	5.76	1.51
		M3	1.00	3.18	0.99
	Korea	M1	0.99	-3.48	1.63
		M2	0.97	-0.34	2.23
		M3	0.97	-1.68	2.47

*M1 : CMAQ, M2 : WRF-Chem, M3 : GEOS-Chem

		China		Korea	
		R	NMB(%)	R	NMB(%)
201901	M1	0.66	28.08	0.87	7.31
	M2	0.49	6.58	0.84	-37.61
	M3	0.63	9.79	0.86	-6.08
201904	M1	0.56	40.25	0.68	18.50
	M2	0.45	55.82	0.66	26.72
	M3	0.56	41.67	0.64	47.42
201907	M1	0.51	45.63	0.72	-7.08
	M2	0.46	127.01	0.69	48.30
	M3	0.42	44.91	0.74	5.18
201910	M1	0.54	22.89	0.65	4.82
	M2	0.47	49.31	0.57	15.04
	M3	0.53	55.58	0.66	35.63

Spatial distribution of Simulated monthly PM_{2.5} concentration and Performance statistics of PM_{2.5} at the surface]



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• The concentrations of the major chemical composition that contribute to the total PM_{2.5} mass concentration was different.

- WRF-Chem simulated higher SO42- concentration than other models in all seasons.
- In summer, the relative humidity was very high, and there were very many clouds, and haze or precipitation continued from mid-July.

 \Rightarrow We are looking into the causes of the overestimated concentration of each component in summer.

- Although all CTMs used the same anthropogenic emissions inventory, CMAQ simulated lower POA concentrations in all seasons than those of other models.
- The discrepancy is associated with the different treatments of secondary organic aerosol (SOA) production.

CMAQ includes new pathway for potential combustion secondary organic aerosol (pcSOA) (Murphy et al., 2017)

Summary & Future works

– In this study, the seasonal prediction performance of PM_{2.5} and major particulate matter (sulfate, nitrate, ammonium, organic carbon, and elemental carbon) simulated by three CMTs was evaluated in China and Korea.

- The concentrations of PM_{2.5} were typically higher during winter compared to summer, and the CTMs showed good

performance in predicting these seasonal variations. However, there were discrepancies in the simulation results of PM_{2.5}

- The results of these multi-model intercomparisons will be helpful in the future research to increase understanding of the differences between the simulated results by muti-CTMs in Northeast Asia and to seek the ways to improve the model

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