



Supplement of

Reduction in vehicular emissions attributable to the Covid-19 lockdown in Shanghai: insights from 5 years of monitoring-based machine learning

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Table S1. Correlation coefficient R and root mean square error (RMSE in $\mu\text{g m}^{-3}$) between predicted and measured NO_x and EC using the random forest model and the multilinear regression model.

R (RMSE)	Random Forest	Multilinear regression
NO_x	0.89-0.98 (12.94-29.34)	0.48 (47.6)
EC	0.90-0.98 (0.27-0.51)	0.45 (0.96)

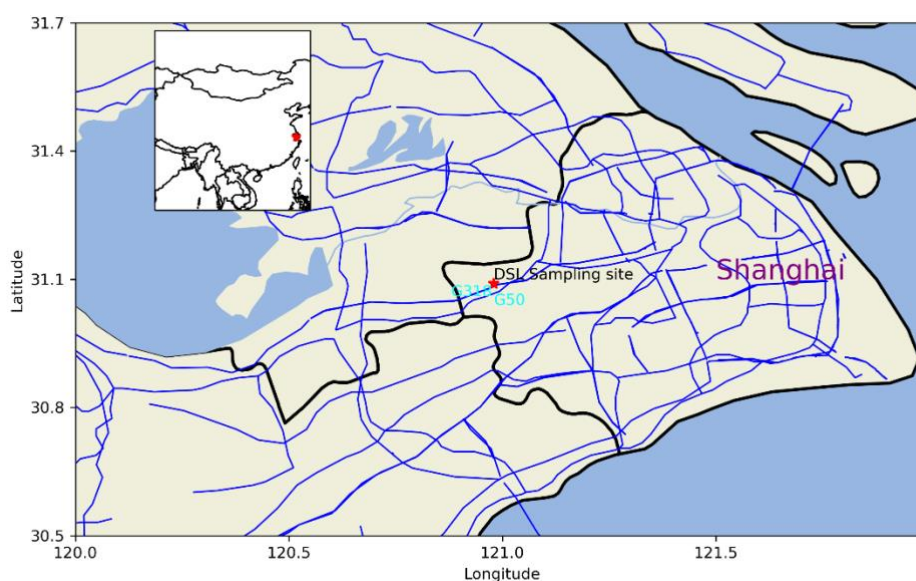


Figure S1. The location of Dianshan Lake (DSL) supersite in west Shanghai. Blue lines represent roads. The DSL sampling site is approximately 1 km away from the nearest highways (G318 and G50). The Inset shows the location of the sampling site in China. (The map was created using a Python package of Cartopy).

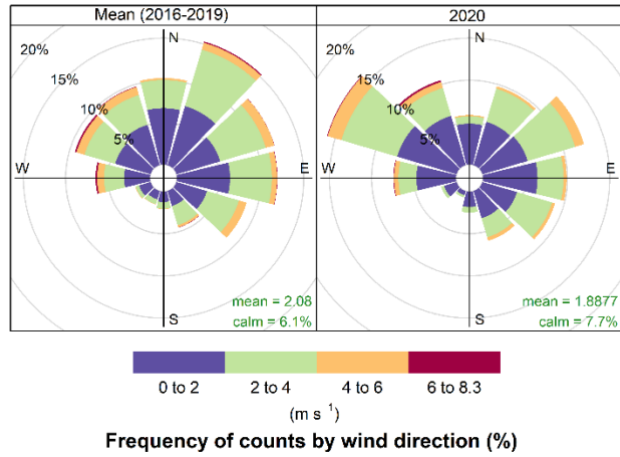
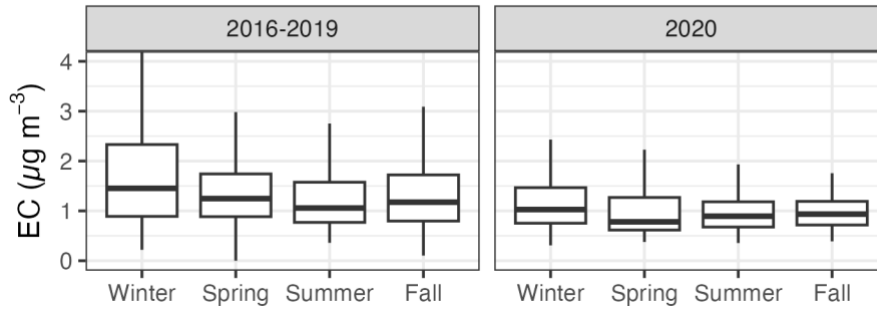


Figure S2. Wind roses for the period between 2016 and 2019 (left), and the year of 2020 with the Covid-19 lockdown (right).

a EC seasonal trend



b NOx seasonal trend

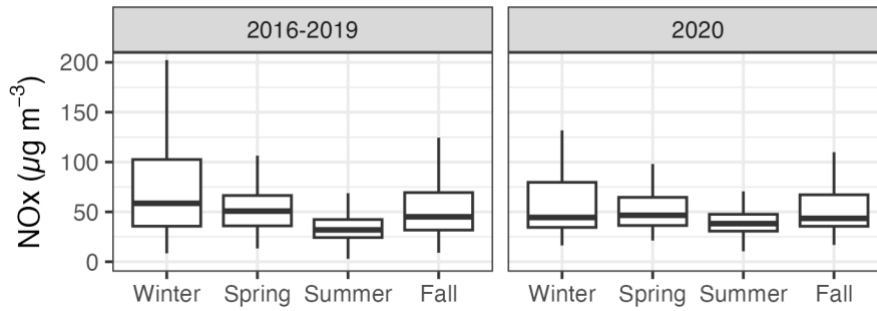


Figure S3. Seasonal variation of EC and NOx for 2016-2019 and 2020.

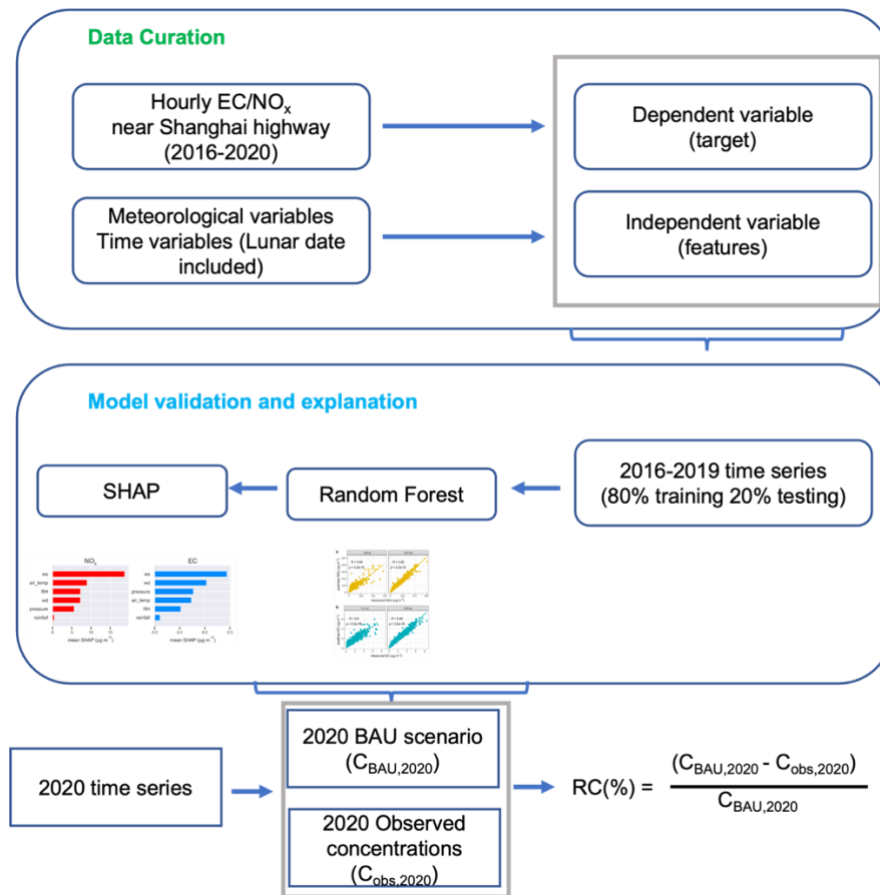


Figure S4. Flow chart of the method. RC standards for relative change in %.

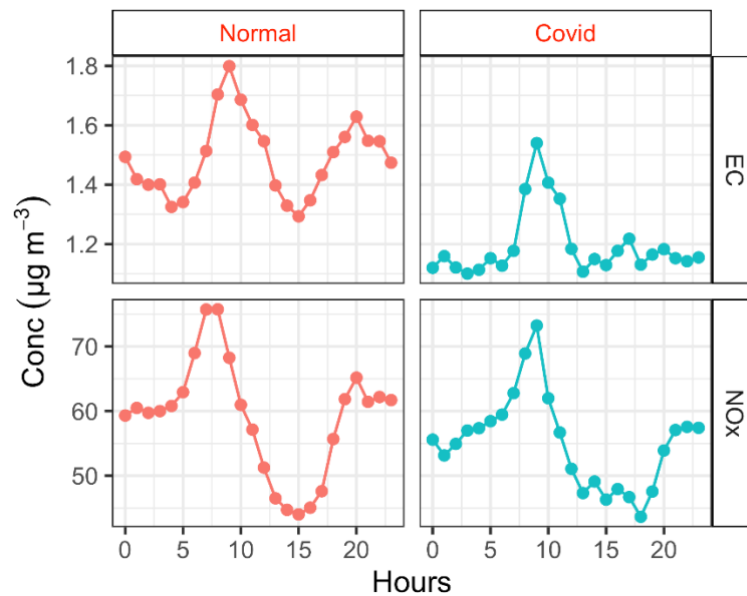


Figure S5. Averaged diurnal cycle of NO_x and EC during the normal years (2016-2019) and 2020 with Covid.

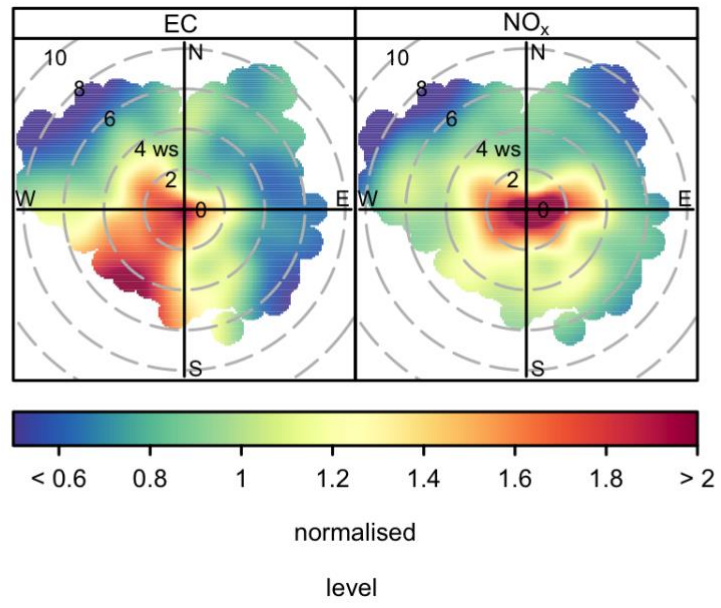


Figure S6. Polar plot of the normalized EC and NO_x as a function of wind speed and direction.