Using machine learning and the GOCI satellite instrument to map fine particulate matter air quality in East Asia

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Abstract. We use 2011-2019 aerosol optical depth (AOD) observations from the Geostationary Ocean Color Imager (GOCI) instrument over East Asia to infer 24-h daily surface fine particulate matter ($PM_{2.5}$) concentrations at continuous 6x6 km² resolution over eastern China, South Korea, and Japan. This is done with a random forest (RF) algorithm applied to the gap-filled GOCI AODs and other data, including information encoded in GOCI AOD retrieval failure, and trained with PM_{2.5} observations from the three national networks. The predicted 24-h GOCI PM_{2.5} concentrations for sites entirely withheld from training in a ten-fold crossvalidation procedure correlate highly with network observations ($R^2 = 0.89$) with single-value precision of 26-32% depending on country. The area-weighted and population-weighted trends of GOCI PM_{2.5} concentrations for eastern China, South Korea, and Japan show steady 2015-2019 declines consistent with surface networks. Further examination of GOCI PM_{2.5} fields for South Korea identifies hotspots where surface network sites were initially lacking and shows 2015-2019 $PM_{2.5}$ decreases across the country except for flat concentrations in the Seoul metropolitan area.

The random forest algorithm

Step 1: Draw a bootstrap sample Each tree is trained with replacement from the training data **Original Data** ootstrap sample

Step 2: Grow different decision trees for each bootstrap sample

Training data

We train our random forest (RF) machine learning algorithm to predict 24-hr surface PM_{2.5} observed at sites in eastern China, South Korea, and Japan. The RF algorithm predicts surface PM2.5 using:

- layer height, 2m temperature, relative humidity, 10m u/v winds, sea level pressure)
- etc)





sensitive to input data

Step 3: Average tree output to make prediction.



Trees have high variance but on average they are unbiased.

Averaging many trees should give an accurate estimate

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Random forest model accuracy

Below is a scatterplot, color-coded by count, comparing surface observations of 24-h PM_{2.5} to the predicted GOCI PM_{2.5} values in grid cells whose records are entirely withheld from training in a crossvalidation procedure.

When aggregated to annual resolution, RMSE between observed and predicted $PM_{2.5}$ is 3.3 µg m⁻³ corresponding to a relative RMSE of 14%. The prediction captures 96% of observed annual variability. Equivalent 24-h statistics in figure.

Read the AMT paper



National trends in PM_{2.5}

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The PM_{2.5} networks show decreasing trends in all three countries and these trends are consistent with the GOCI $PM_{2.5}$ for both areal and population-weighted means. However, the PM₂₅ networks in eastern China and South Korea underestimate the population-weighted means.







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Contact information and links

Please use this data!