

Assessing the health estimation capacity of air pollution exposure prediction models

Additional file

Contents

Section A. Simulated seasonal, monthly, and acute time trends	1
Figure S1. Comparing simulated observed time series and model predictions using overall model performance (orange circles) and frequency band k model performance (triangles) for correlation r, $RMSE$, and LVR. For model predictions, classical error $w^k(t)$ was added to simulated observations at frequency bands $k = 1, \dots, 6$ with magnitude of error $\sigma_c = \{0.2, 0.4, 0.6, 0.8\}$. Results are shown for frequency band k ($k = 1, 2$) model performance.	2
Figure S1a. Frequency band $k = 1$ model performance (purple triangles)	2
Figure S1b. Frequency band $k = 2$ model performance (pink triangles)	3
Figure S2. Comparing simulated observed time series and model predictions using overall model performance (orange circles) and frequency band $k = 6$ model performance (green triangles) for correlation r, $RMSE$, and LVR with magnitude of error $\sigma_c = \{0.2, 0.4, 0.6, 0.8\}$. This plot shows the impact of changing the variance of the seasonal component of the simulated observed time series, $\text{Var}(x_{(1)}(t)) \in \{1^2, 1.5^2, 2^2\}$.	4
Figure S3. Map of 17 $PM_{2.5}$ monitors across 8 US cities.	5
Table S1. For each monitor ID, table of the corresponding city and state, number of days of data, number and percent of imputed monitor observations, and median (interquartile range) $PM_{2.5}$ ($\mu g/m^3$) for both the monitoring data and FAQSD.	5
References	6

Section A. Simulated seasonal, monthly, and acute time trends

To create timescale components for $k = 1, 2, 6$, we specified $x^k(t) = \frac{\tilde{x}^k(t)}{\text{Var}(\tilde{x}^k(t))}$ where

$$\tilde{x}^k(t) = \cos\left(\frac{2\pi\gamma}{365} \times t\right)$$

corresponding to γ cycles per year. For the seasonal trend ($k = 1$), we set $\gamma = 1$ such that one complete cycle was completed each year. For the monthly ($k = 2$), we set $\gamma = 8$, which fits within the second frequency band of [6-12] cycles per year from Dominici et al. 2003 (Dominici et al. 2003). Last, for the acute trend ($k = 6$), we specified $\gamma = 180$, or approximately a cycle every two days.

Figure S1. Comparing simulated observed time series and model predictions using overall model performance (orange circles) and frequency band k model performance (triangles) for correlation r , $RMSE$, and LVR . For model predictions, classical error $w^k(t)$ was added to simulated observations at frequency bands $k = 1, \dots, 6$ with magnitude of error $\sigma_c = \{0.2, 0.4, 0.6, 0.8\}$. Results are shown for frequency band k ($k = 1, 2$) model performance.

Figure S1a. Frequency band $k = 1$ model performance (purple triangles)

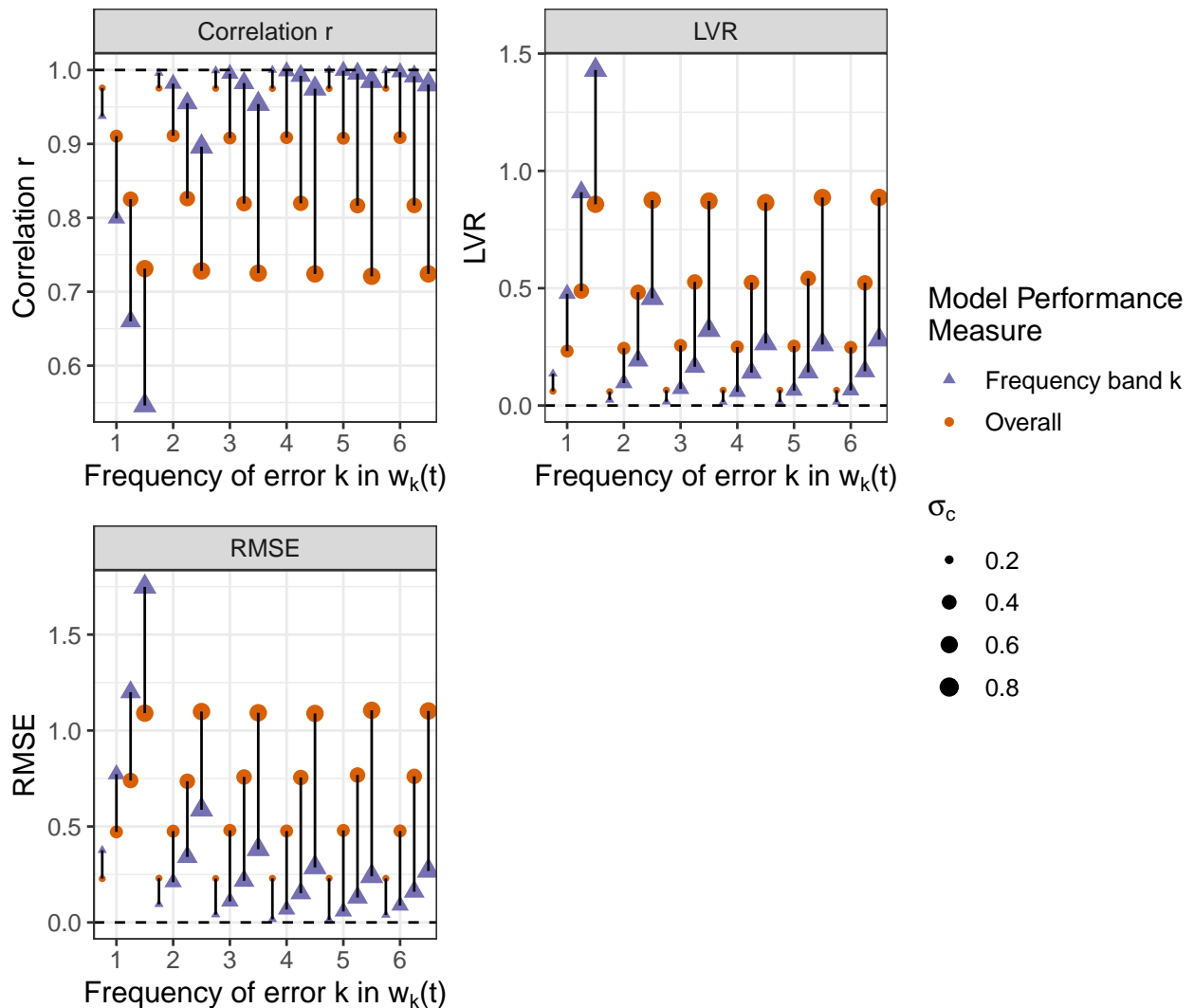


Figure S1b. Frequency band $k = 2$ model performance (pink triangles)

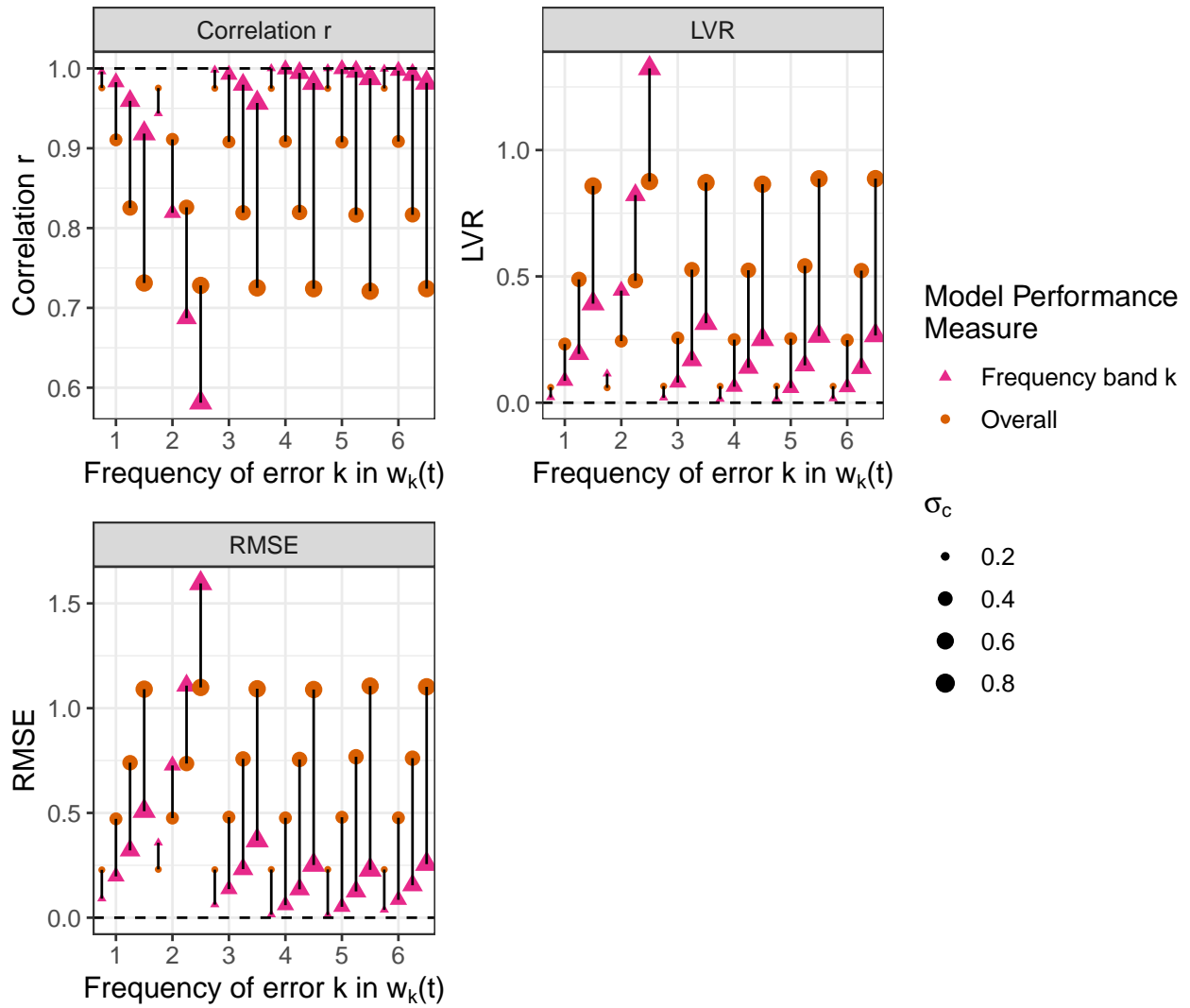


Figure S2. Comparing simulated observed time series and model predictions using overall model performance (orange circles) and frequency band $k = 6$ model performance (green triangles) for correlation r , $RMSE$, and LVR with magnitude of error $\sigma_c = \{0.2, 0.4, 0.6, 0.8\}$. This plot shows the impact of changing the variance of the seasonal component of the simulated observed time series, $\text{Var}(x_{(1)}(t)) \in \{1^2, 1.5^2, 2^2\}$.

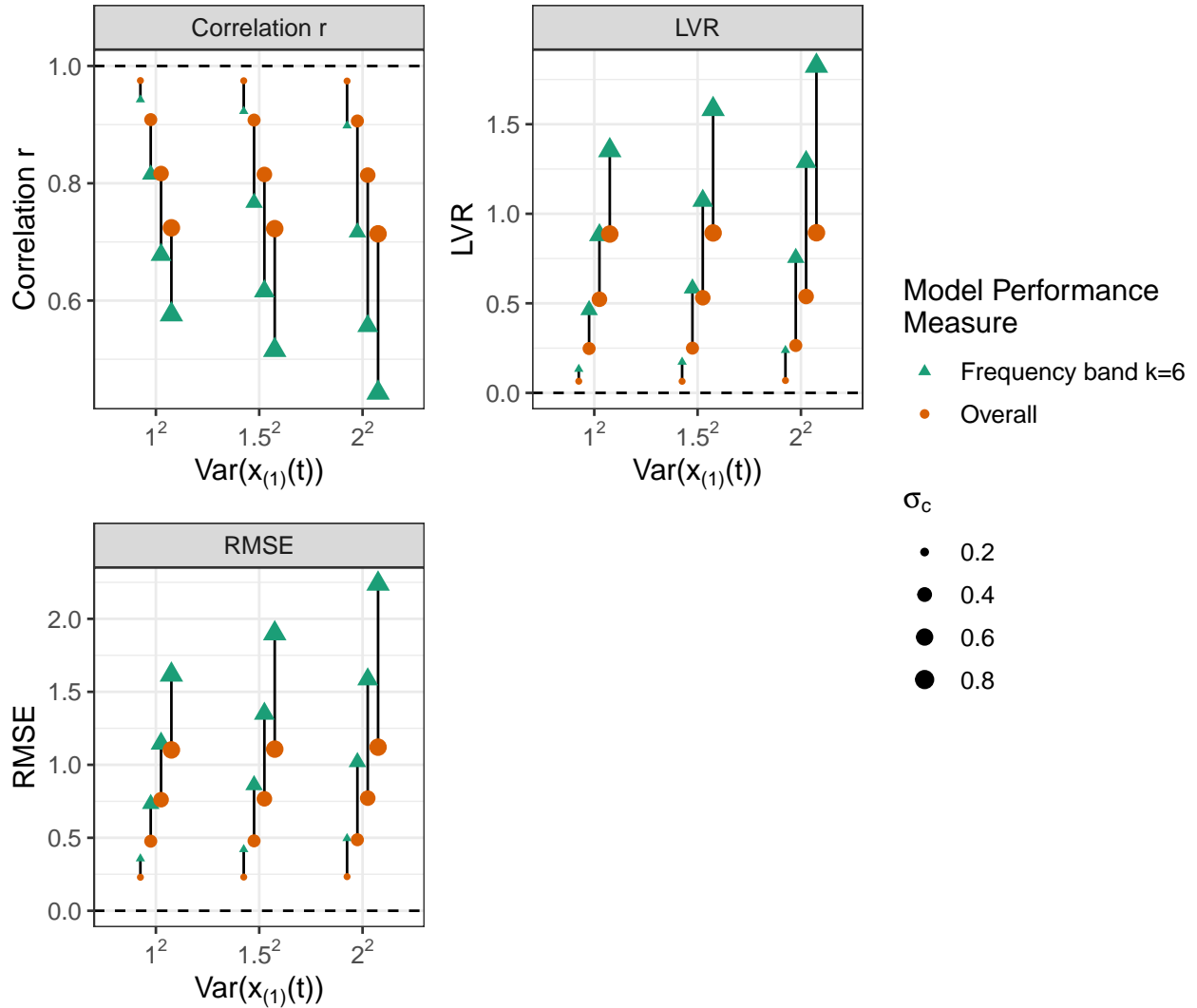


Figure S3. Map of 17 PM_{2.5} monitors across 8 US cities.



Table S1. For each monitor ID, table of the corresponding city and state, number of days of data, number and percent of imputed monitor observations, and median (interquartile range) PM_{2.5} ($\mu\text{g}/\text{m}^3$) for both the monitoring data and FAQSD.

Monitor ID	City	State	Days	Imputed (N (%))	Monitor (Median (IQR))	FAQSD (Median (IQR))
060371002	Los Angeles	California	990	58 (5.9)	11.8 (5.3)	11.16 (5.14)
060371103	Los Angeles	California	1422	147 (10.3)	12 (4.97)	11.93 (5.72)
060374002	Los Angeles	California	1527	126 (8.3)	10.3 (4.35)	10.61 (5.02)
060374004	Los Angeles	California	1093	46 (4.2)	9.3 (4.6)	10.65 (5.05)
060374008	Los Angeles	California	1074	46 (4.3)	11.45 (5.1)	11.12 (5.23)
130670003	Atlanta	Georgia	1052	58 (5.5)	10.3 (5.82)	10.12 (5.79)
130890002	Atlanta	Georgia	1051	68 (6.5)	10.13 (5.3)	10.23 (5.79)
130892001	Atlanta	Georgia	686	47 (6.9)	10.7 (5.47)	10.98 (5.84)
131210032	Atlanta	Georgia	686	27 (3.9)	10.5 (5.8)	11 (5.85)
360810124	New York	New York	2351	141 (6)	7 (5.1)	7.88 (5.91)
420030008	Pittsburgh	Pennsylvania	593	18 (3)	9 (5.5)	10.26 (6.1)
420030064	Pittsburgh	Pennsylvania	960	53 (5.5)	13.1 (10.8)	11.6 (8.64)
481130069	Dallas	Texas	1250	39 (3.1)	8.9 (5.4)	9.5 (5.14)
482011035	Houston	Texas	1645	72 (4.4)	11.16 (5.1)	11.05 (4.82)
490353006	Salt Lake City	Utah	958	64 (6.7)	6 (4.32)	6.24 (4.76)
490353010	Salt Lake City	Utah	1003	83 (8.3)	7.2 (5.55)	6.49 (5.6)
530530029	Seattle/Tacoma	Washington	1547	106 (6.9)	5.4 (4.1)	5.69 (4.6)

References

Dominici, Francesca, Aidan McDermott, Scott L Zeger, and Jonathan M Samet. 2003. "Airborne Particulate Matter and Mortality: Timescale Effects in Four Us Cities." *American Journal of Epidemiology* 157 (12): 1055–65.