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Online Supplement

**Spatio-Temporal Models with Space-Time Interaction  
and Their Applications to Air Pollution Data**

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1. Additional figures

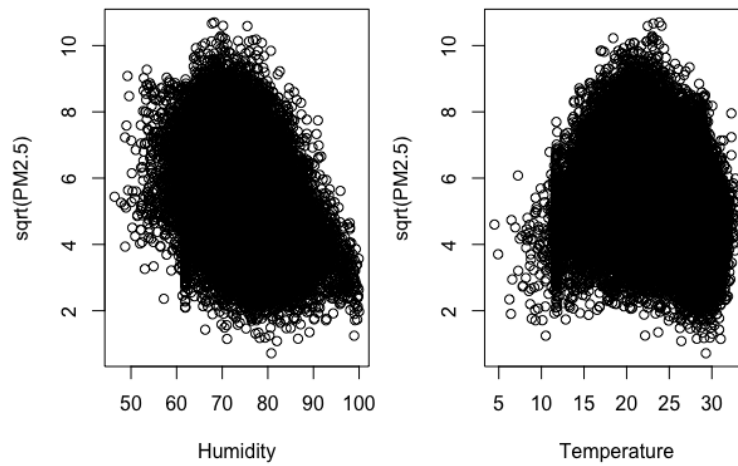


Figure 1: Scatter plots of the square roots of the  $PM_{2.5}$  observations, with respect to relative humidity (left) and temperature (right)

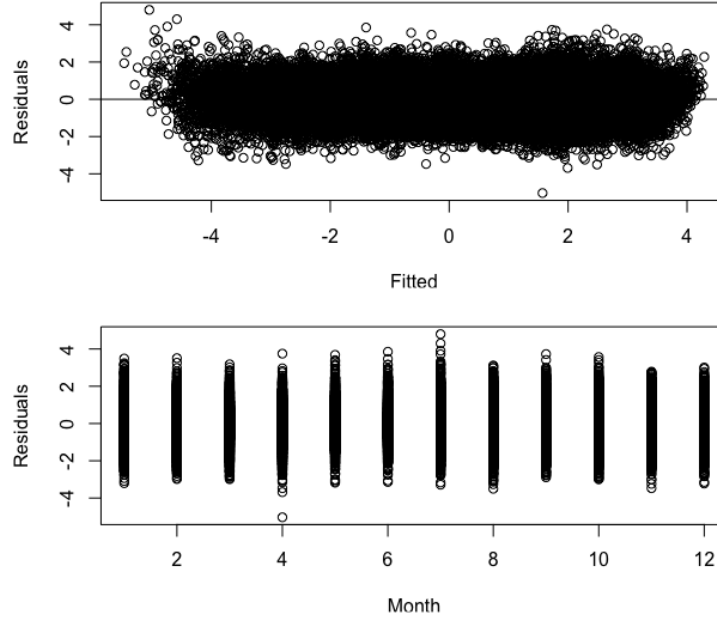


Figure 2: (Top) Standardized residuals are plotted against fitted values; (Bottom) Standardized residuals are plotted corresponding to different months

## 2. Proof of Theorem 1

Note that the set-up of our problem is similar to a generalized least squares (GLS) problem, where  $Y = X\theta + \varepsilon$ , such that  $\varepsilon \sim N(0, \sigma^2\Omega)$ . Following our previous notations,  $\Omega = (\Sigma_v + D)$ , where  $D$  is a diagonal matrix with diagonal elements equal to some  $\tau_j^2$ .

Now, for proving the required result, we define three different estimators of  $\theta$ . Below,  $\hat{\theta}$  is the estimator we are considering

in this study,  $\hat{\theta}_G$  denotes the usual GLS estimator, and  $\hat{\theta}_F$  is a feasible GLS estimator.

$$\hat{\theta} = (X'\hat{\Omega}^{-1/2}W\hat{\Omega}^{-1/2}X)^{-1}(X'\hat{\Omega}^{-1/2}W\hat{\Omega}^{-1/2}Y)$$

$$\hat{\theta}_G = (X'\Omega^{-1}X)^{-1}(X'\Omega^{-1}Y)$$

$$\hat{\theta}_F = (X'\hat{\Omega}^{-1}X)^{-1}(X'\hat{\Omega}^{-1}Y)$$

In the above,  $W$  is the weight matrix as defined in Section 3.2 of the main paper and  $\hat{\Omega}$  is our estimate of the covariance matrix. For convenience, we use  $N = nT$  hereafter. Following Baltagi [2011, Chapter 9], we know that  $\sqrt{N}(\hat{\theta}_G - \theta)$  and  $\sqrt{N}(\hat{\theta}_F - \theta)$  have the same asymptotic distribution  $N(0, \sigma^2 Q^{-1})$ , where  $Q = \lim(X'\Omega^{-1}X/N)$  as  $N \rightarrow \infty$ , if  $X'(\hat{\Omega}^{-1} - \Omega^{-1})X/N \xrightarrow{P} 0$  and  $X'(\hat{\Omega}^{-1} - \Omega^{-1})\varepsilon/N \xrightarrow{P} 0$ . Further, a sufficient condition for this to hold is that  $\hat{\Omega}$  is a consistent estimator for  $\Omega$  and that  $X$  has a satisfactory limiting behavior.

Let us now assume that the estimate  $\hat{\tau}_j^2$  is consistent for  $\tau_j^2$ , for all  $j$ . That would automatically ensure the consistency of  $\hat{\Omega}$  and thereby we can conclude that  $\hat{\theta}_F$  and  $\hat{\theta}_G$  have same asymptotic distribution. Further, note that  $X'\hat{\Omega}^{-1/2}W\hat{\Omega}^{-1/2}X - X'\hat{\Omega}^{-1}X = X'\hat{\Omega}^{-1/2}(W - I)\hat{\Omega}^{-1/2}X$ . Taking any appropriate norm (2-norm, for example) on both sides, we can argue that

$$\left\| X'\hat{\Omega}^{-1/2}W\hat{\Omega}^{-1/2}X - X'\hat{\Omega}^{-1}X \right\| \rightarrow 0$$

as  $N \rightarrow \infty$ , in view of the fact that  $\|W - I\| = 2/\log N$ , and that  $\hat{\Omega}$  is a consistent estimator for  $\Omega$ , the population covariance matrix. In a similar fashion, we can show that

$$\left\| X'\hat{\Omega}^{-1/2}W\hat{\Omega}^{-1/2}\varepsilon - X'\hat{\Omega}^{-1}\varepsilon \right\| \rightarrow 0$$

as  $N \rightarrow \infty$ , and thus we can conclude that  $\sqrt{N}(\hat{\theta} - \theta)$  and  $\sqrt{N}(\hat{\theta}_F - \theta)$  have the same asymptotic distribution.

Clearly, all we need to prove is that  $\hat{\tau}_j^2$  is a consistent estimator for  $\tau_j^2$  for all  $j$ . To this end, recall that  $\hat{\tau}_j^2$  is the maximum likelihood estimator (MLE) of  $\tau_j^2$  for the problem  $\hat{\varepsilon}_j \sim N(0, (\Sigma_v^{(j)} + \tau_j^2 I))$ , where  $\hat{\varepsilon}_j$  is the vector of scaled residuals corresponding to the  $j$ th season and  $\Sigma_v^{(j)}$  is the submatrix of  $\Sigma_v$  corresponding to the same. It is known that MLE is a consistent estimator for such problems. Let  $n_j$  be the length of  $\varepsilon_j$ . Since  $T \rightarrow \infty$ , it is clear that the number of observations per season will also approach infinity, and thus  $n_j \rightarrow \infty$ . Hence,  $\hat{\tau}_j^2$  is consistent for  $\tau_j^2$  and that ends our proof for the asymptotic normality of  $\hat{\theta}$ . The consistency result follows automatically from the above.

## References

Badi H. Baltagi. *Econometrics*. Springer Texts in Business and Economics. Springer, Heidelberg, fifth edition, 2011. URL <https://doi.org/10.1007/978-3-642-20059-5>.

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