

A new covariance function for spatio-temporal data analysis with application to atmospheric pollution and sensor networking

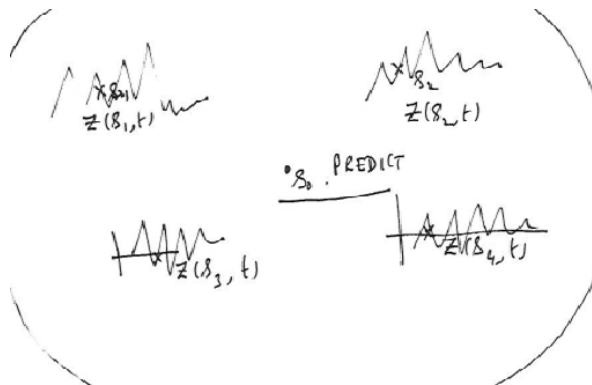
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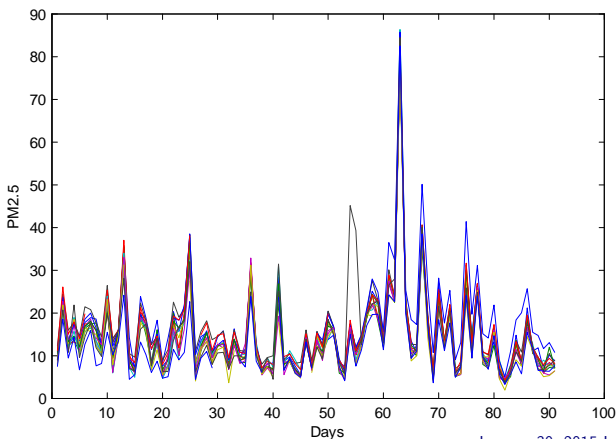
Laboratoire des Signaux et Systemes, Supelec,
Paris, Gif-sur-Yvette

Spatio-temporal observations

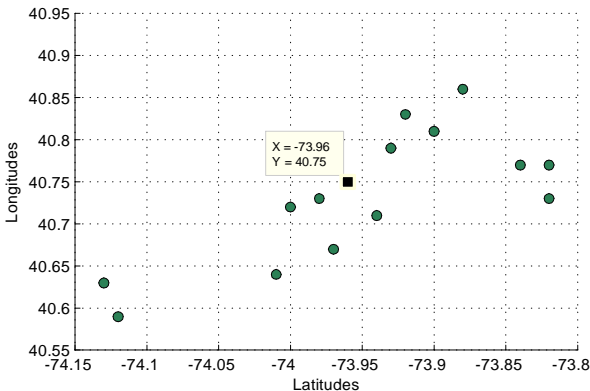


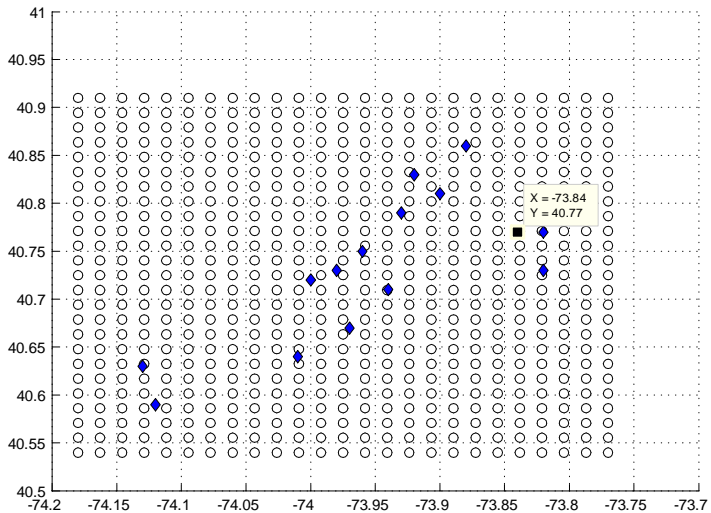
New York City air pollution data, PM2.5 measure is one of six primary air pollutants and is a mixture of fine particles and gaseous compounds such as sulphur dioxide (SO₂) and nitrogen oxides (NO_x)

In 2002, every 3 days and during the first 9 months, 91 equally spaced days, observed at 15 monitoring stations, $X(\mathbf{s}, t)$, see [SM05],
 $\{Z(\mathbf{s}, t) = \Delta_t X(\mathbf{s}, t) : (\mathbf{s}, t) : \mathbf{s} \in \mathbb{R}^2, t \in \mathbb{Z}\}$



Missing data: mean by locations and fixed in time, but 3/4 of data at Location #11 is missing, namely from days 24th to 91st, location \mathbf{s}_0 , $\{Z(\mathbf{s}_0, t); t = 1, 2, 3 \dots n\}$.
Sample $\{Z(\mathbf{s}_i, t); \mathbf{s}_i = 1, 2, \dots, m; t = 1, \dots, n\}$. Bronx, Brooklyn, Manhattan, Queens, Staten Island





We assume that the random process is spatially and temporally second order stationary, homogeneous and isotropic, i.e. $E [Z (\mathbf{s}, t)] = \mu$, $Var [Z (\mathbf{s}, t)] = \sigma_Z^2 < \infty$

$$Cov [Z (\mathbf{s}, t), Z (\mathbf{s} + \mathbf{h}, t + u)] = c (\mathbf{h}, u) = c (\|\mathbf{h}\|, u),$$

We note that $c (\mathbf{h}, 0)$ and $c (\mathbf{0}, u)$ correspond to the purely spatial and purely temporal covariances.

Spatio-temporal **variogram** for $\{Z (\mathbf{h}, t)\}$

$$2\gamma (\mathbf{h}, u) = Var [Z (\mathbf{s} + \mathbf{h}, t + u) - Z (\mathbf{s}, t)].$$

$$2\gamma (\mathbf{h}, u) = 2 [c (\mathbf{0}, 0) - c (\mathbf{h}, u)],$$

for an isotropic process, $\gamma (\mathbf{h}, u) = \gamma (\|\mathbf{h}\|, u)$.

Frequency domain in time and Spatial domain in space, $\{Z(\mathbf{s}_i, t); t = 1, \dots, n\}$

DFT at $\omega_k = \frac{2\pi k}{n}$, $k = 0, \pm 1, \dots, \pm \left[\frac{n}{2}\right]$.

$$J_{\mathbf{s}_i}(\omega_k) = \frac{1}{\sqrt{2\pi n}} \sum_{t=1}^n Z(\mathbf{s}_i, t) e^{-it\omega_k},$$

periodogram

$$I_{\mathbf{s}_i}(\omega_k) = |J_{\mathbf{s}_i}(\omega_k)|^2.$$

$J_{s_i}(\omega_k)$ asymptotically independent and Gaussian

$$E(J_{s_i}(\omega_k)) = 0$$

$$\text{Var}(J_{s_i}(\omega_k)) = E(I_{s_i}(\omega_k)) \simeq g_{s_i}(\omega_k).$$

$$\text{Cov}(J_{s_i}(\omega_k), J_{s_i}(\omega_{k'})) \simeq 0, \quad k \neq k'.$$

Isotropy

$$g_{s_i}(\omega_k) = g(\omega_k), \text{ for all } i$$

Fourier T and Spectral Repr

Introduce white noise

$$J_{s,e}(\omega) = \int e^{i s \cdot \underline{\lambda}} \left[\sqrt{\frac{n}{2\pi}} \right] dZ_e(\underline{\lambda}, \omega).$$

Laplacian

$$\left[\frac{\partial^2}{\partial s_1^2} + \frac{\partial^2}{\partial s_2^2} - |c(\omega)|^2 \right]^v J_s(\omega) = J_{s,e}(\omega).$$

$$\left(-\lambda_1^2 - \lambda_2^2 - |c(\omega)|^2 \right)^v dZ_z(\underline{\lambda}, \omega) = dZ_e(\underline{\lambda}, \omega),$$

Spectral density

$$f_z(\underline{\lambda}, \omega) = \frac{\sigma_e^2}{(2\pi)^2 \left(\lambda_1^2 + \lambda_2^2 + |c(\omega)|^2 \right)^{2\nu}}$$

Covariance

$$\begin{aligned} & \text{Cov}(J_s(\omega), J_{s+h}(\omega)) \\ &= \frac{\sigma_e^2}{2\pi} \left(\frac{\|\mathbf{h}\|}{2|c(\omega)|} \right)^{2\nu-1} \frac{K_{2\nu-1}(|c(\omega)| \|\mathbf{h}\|)}{\Gamma(2\nu)} \end{aligned}$$

$K_{2\nu-1}$: modified Bessel function of the second kind of order $2\nu - 1$.

General, \mathbb{R}^d , $\nu \geq 2$ integer, correlation function is given by

$$\rho(\|\mathbf{h}\|, \omega) = \frac{(\|\mathbf{h}\| |c(\omega)|)^{2\nu - \frac{d}{2}}}{2^{2\nu - \frac{d}{2} - 1} \Gamma(2\nu - \frac{d}{2})} K_{2\nu - \frac{d}{2}}(\|\mathbf{h}\| |c(\omega)|),$$

and

$$C(0, \omega) = \frac{\sigma_e^2}{(2\pi)^{\frac{d}{2}} 2^{\frac{d}{2}} (|c(\omega)|^2)^{2\nu - \frac{d}{2}}} \frac{\Gamma(2\nu - \frac{d}{2})}{\Gamma(2\nu)} = g(\omega).$$

Parameters

$$C(0, \omega) = \frac{\sigma_e^2}{2(2\nu - 1) \left(|c(\omega)|^2 \right)^{2\nu - 1}} = g(\omega).$$

$|c(\omega)|^2$ and common spectral density $g(\omega)$, ARMA, FARMA etc..

Estimation of parameters

Correlation

$$\rho(\|\mathbf{h}\|, \omega) = \frac{C(\|\mathbf{h}\|, \omega)}{C(0, \omega)} = \frac{1}{2^{2\nu-2} \Gamma(2\nu-1)} \times (\|\mathbf{h}\| |c(\omega)|)^{2\nu-1} K_{2\nu-1}(|c(\omega)| \|\mathbf{h}\|).$$

Special case $\nu = 1$,

$$\rho(\|\mathbf{h}\|, \omega) = \|\mathbf{h}\| |c(\omega)| K_1(\|\mathbf{h}\| |c(\omega)|).$$

Spatio-temporal prediction

$$J_{\mathbf{s}_0}(\omega) = \frac{1}{\sqrt{(2\pi n)}} \sum_{t=1}^n Z(\mathbf{s}_0, t) e^{-it\omega},$$

and by inversion, we have

$$Z(\mathbf{s}_0, t) = \sqrt{\frac{n}{2\pi}} \int_{-\pi}^{\pi} J_{\mathbf{s}_0}(\omega) e^{it\omega} d\omega.$$

In other words given $\{J_{\mathbf{s}_0}(\omega), \text{ for all } -\pi \leq \omega \leq \pi\}$, we can uniquely recover the sequence $\{Z(\mathbf{s}_0, t); t = 1, \dots, n\}$.

Given

$$\underline{J}'_m(\omega) = [J_{s_1}(\omega), J_{s_2}(\omega), \dots, J_{s_m}(\omega)].$$

We note

$$\begin{aligned} E[\underline{J}_m(\omega)] &= \mathbf{0} \\ E[\underline{J}_m(\omega) \underline{J}_m^*(\omega)] &= \underline{F}_m(\omega), \end{aligned}$$

$F_m(\omega) = (C(\|\mathbf{s}_i - \mathbf{s}_j\|, \omega); i, j = 1, 2, \dots, m)$, and each element $C(\|\mathbf{s}_i - \mathbf{s}_j\|, \omega)$ is given above.

$$\underline{J}'_{m+1}(\omega) = [J_0(\omega), \underline{J}'_m(\omega)],$$

which has zero mean, and variance covariance matrix

$$\begin{aligned} & E [\underline{J}_{m+1}(\omega) \underline{J}_{m+1}^*(\omega)] \\ = & \begin{bmatrix} C_0(0, \omega) & E(J_0(\omega) \underline{J}'_m(\omega)) \\ E(\underline{J}_m(\omega) J_0^*(\omega)) & E(\underline{J}_m(\omega) \underline{J}'_m(\omega)) \end{bmatrix} \\ = & \begin{bmatrix} C_0(0, \omega) & \underline{G}'_0(\omega) \\ \underline{G}_0(\omega) & \underline{F}_m(\omega) \end{bmatrix}, \end{aligned}$$

$$C_0(0, \omega) = E(J_0(\omega) J_0^*(\omega)) = C(0, \omega),$$

$$\begin{aligned} \underline{G}'_0(\omega) &= E[J_0(\omega) \underline{J}'_m(\omega)] \\ &= [C(\|\mathbf{s}_0 - \mathbf{s}_1\|, \omega), \dots, C(\|\mathbf{s}_0 - \mathbf{s}_m\|, \omega)] \end{aligned}$$

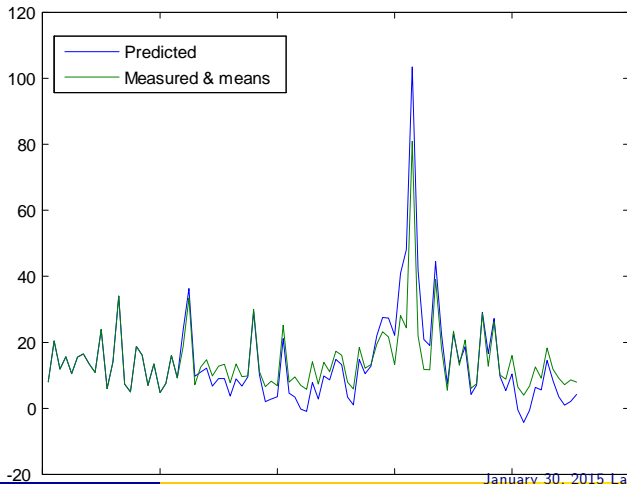
$$E[J_0(\omega) | \underline{J}_m(\omega)] = \underline{G}'_0(\omega) \underline{F}_m^{-1}(\omega) \underline{J}_m(\omega)$$

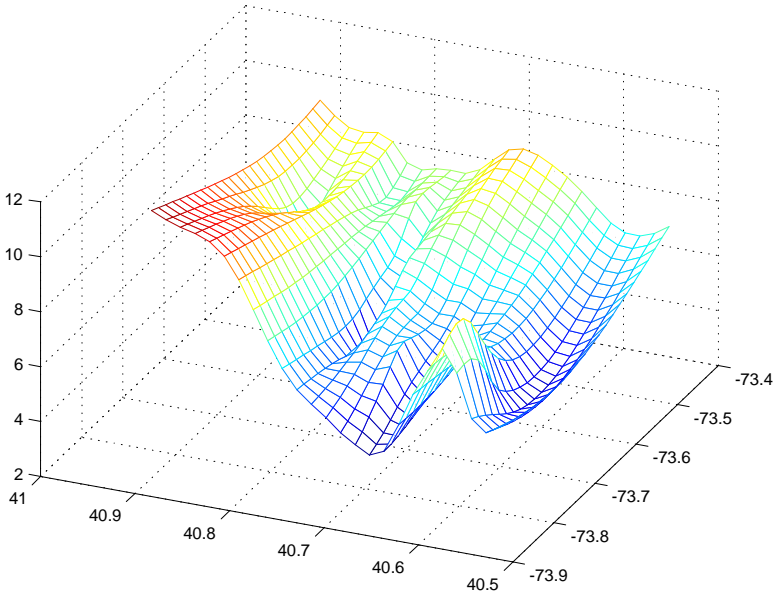
the minimum mean square error

$$\sigma_m^2(\omega) = C(0, \omega) - \underline{G}'_0(\omega) \underline{F}_m^{-1}(\omega) \underline{G}_0(\omega).$$

$$\hat{J}_0(\omega) = \hat{\underline{G}}'_0(\omega) \hat{\underline{F}}_m^{-1}(\omega) J_m(\omega).$$

Air pollution data, common spectral density $g(\omega)$,
ARMA(1,1), parameters estimation by Whittle method,
see [ST13] for details







S. K. SAHU AND K. V. MARDIA.

A bayesian kriged kalman model for short-term forecasting of air pollution levels.

Journal of the Royal Statistical Society: Series C (Applied Statistics) **54**(1), 223–244 (2005).



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A space-time covariance function for spatio-temporal random processes and spatio-temporal prediction (kriging).

ArXiv e-prints (November 2013).