

PLN spatial

Raphaelle Momal

June 3, 2019

A variogram in geostatistics is defined for an intrinsically stationary spatial process Z as follows, for sites s and t separated by a distance d :

$$\begin{aligned}\gamma(d) &= \frac{1}{2} \mathbb{E}[(Z(s) - Z(t))^2] \\ &= \frac{1}{2} (\mathbb{V}(Z(s)) + \mathbb{E}[Z(s)]^2 + \mathbb{V}(Z(t)) + \mathbb{E}[Z(t)]^2 - \mathbb{E}(Z(s)Z(t)))\end{aligned}$$

We wish to write the theoretical variogram formula for the PLN model.

1 Variogram for \mathbf{Y}

We recall the PLN model :

$$\begin{aligned}Z_s &\sim \mathcal{N}(0, \Sigma) \\ Y_{sj} | Z_{sj} &\sim \mathcal{P}(e^{x_s \beta_j + Z_{sj}})\end{aligned}$$

In what follows we want to account for a spatial structure of the data. This means we will need to differentiate between two sites and two species at the same time. Therefore we write $Cov(Z) = [\Gamma]_{stjk}$ where s and t are sites and j and k are species. Two models are under consideration here:

1. $\Gamma_{stjk} = \delta_{st} \sigma_{jk}$
2. $\Gamma_{stjk} = \delta_{st} + \mathbb{1}_{s=t} \sigma_{jk}$

Where δ and σ represent respectively the spatial and inter-species dependence. We assume

- (a) a stationarity in spatial variance, that is that $\delta_{uu} = cst$ for any site u . This means that a behavior can be spatially translated.
- (b) The inter-species dependence only intervenes on a same site, so furthermore two species in two different sites will behave independently.

For clarity sake, we write Γ_{ssjj} as Γ_j , according to the assumption of spatial variance stationarity.

- Est-ce que (b) implique que $\Gamma_{stj} = \Gamma_{st}$?
- En suivant (b), devrait-on écrire le modèle 1 comme

$$\Gamma_{stjk} = \mathbb{1}_{s \neq t} \delta_{st} + \mathbb{1}_{s=t} \delta_{st} \sigma_{jk}$$

1.1 Y characteristics

We know the expectation, variance and covariance expressions for the PLN model. When only one species is considered, we have:

$$\begin{aligned}\mathbb{E}[Y_{sj}] &= e^{x_s \beta_j - \Gamma_j^2/2} = \mu_{sj} \\ \mathbb{V}(Y_{sj}) &= \mu_{sj} + \mu_{sj}^2 (e^{\Gamma_j^2} - 1) \\ \mathbb{E}[Y_{sj} Y_{tj}] &= \mu_{sj} \mu_{tj} e^{\Gamma_{stj}}\end{aligned}$$

1.2 Same species in two sites

For sites s and t at a distance d :

$$\begin{aligned}\gamma_Y(d) &= \frac{1}{2} \left[\mu_s + \mu_s^2 (e^{\Gamma_j^2} - 1) + \mu_t^2 + \mu_t + \mu_t^2 (e^{\Gamma_j^2} - 1) + \mu_t \right] - \mu_s \mu_t e^{\Gamma_{stj}} \\ &= \frac{1}{2} \left[(\mu_s^2 + \mu_t^2) e^{\Gamma_j^2} + \mu_s + \mu_t \right] - \mu_s \mu_t e^{\Gamma_{stj}}\end{aligned}$$

We then define $\tilde{Y}_{sj} = Y_{sj}/\mu_{sj}$. It comes that

$$\begin{aligned}\hat{\gamma}_{\tilde{Y}}(d) &= \frac{1}{2} \mathbb{E} \left[(\tilde{Y}_{sj} - \tilde{Y}_{tj})^2 - \left(\frac{1}{\mu_s} + \frac{1}{\mu_t} \right) \right] \\ &= e^{\Gamma_j^2} - e^{\Gamma_{stj}}\end{aligned}$$

$\hat{\gamma}_{\tilde{Y}}$ resembles the classical semi-variogram for a stationary spatial process except it is a difference of exponentials rather than a difference of direct variance and covariance. **indice en s ou en sj, il faut choisir**

1.3 Two species in two sites

We recall that

$$\mathbb{E}[Y_{sj} Y_{tk}] = \mu_{sj} \mu_{tk} e^{\Gamma_{stjk}}.$$

Similarly to the above paragraph, we get:

$$\mathbb{E}[(Y_{sj} - Y_{tk})^2] = \left[\mu_{sj} + \mu_{tk} + \mu_{sj}^2 e^{\Gamma_{sj}^2} + \mu_{tk}^2 e^{\Gamma_{tk}^2} \right] - \mu_{sj} \mu_{tk} e^{\Gamma_{stjk}}$$

Then,

$$\hat{\gamma}_{\tilde{Y}_{jk}}(d) = \frac{1}{2} \left(e^{\Gamma_{sj}^2} + e^{\Gamma_{tk}^2} \right) - e^{\Gamma_{stjk}}$$

2 Perspectives

- Culture sur INLA, approximation de Laplace et courbure Gaussienne
- Vraisemblance composite, éventuellement utilisée sur les arbres, pour approcher $p(Z|Y)$.