

# Barcast Model

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## 1 Barcast Model

### 1.1 Data Model

$$\mathbf{W}_t = \mathbf{H}_t T_t + \mathbf{B}_t + \boldsymbol{\eta}_t \quad \boldsymbol{\eta}_t \sim \mathcal{N}(\mathbf{0}, \boldsymbol{\Sigma}_t) \quad \boldsymbol{\Sigma}_t = \begin{pmatrix} \boldsymbol{\Sigma}_{It} & \mathbf{0} \\ \mathbf{0} & \boldsymbol{\Sigma}_{Pt} \end{pmatrix}$$

where  $\mathbf{W}_t = \begin{pmatrix} \mathbf{W}_{It} \\ \mathbf{W}_{Pt} \end{pmatrix}$  is a vector of instrumental observations  $\mathbf{W}_{It}$  and proxy observations  $\mathbf{W}_{Pt}$  at time  $t$ ,  $T_t$  is a latent climate variable,  $\mathbf{H}_t = \begin{pmatrix} \mathbf{H}_{It} \\ \boldsymbol{\beta}_1 \mathbf{H}_{Pt} \end{pmatrix}$ ,  $\mathbf{B}_t = \begin{pmatrix} \mathbf{H}_{It} \\ \mathbf{H}_{Pt} \end{pmatrix} \begin{pmatrix} \mathbf{0}_{N_I} \\ \boldsymbol{\beta}_0 \end{pmatrix}$ ,  $\boldsymbol{\Sigma}_{It} = \tau_I^2 \mathbf{I}_{N_{It}}$ ,  $\boldsymbol{\Sigma}_{Pt} = \tau_P^2 \mathbf{I}_{N_{Pt}}$ ,  $N_I$  is the total number of instrumental observations,  $N_P$  is the total number of proxy observations,  $N_{It}$  is the number of instrumental observations at time  $t$ ,  $N_{Pt}$  is the number of proxy observations at time  $t$ ,  $\mathbf{H}_{It}$  and  $\mathbf{H}_{Pt}$  are selection matrices indicating whether the instrumental or proxy variable was measured at time  $t$ , and  $\boldsymbol{\beta}_0$  and  $\boldsymbol{\beta}_1$  are regression coefficients relating the proxy observations and latent field.

### 1.2 Process Model

$$\mathbf{T} \sim \mathcal{N}(\mathbf{0}, \sigma^2 \mathbf{Q})$$

where  $\mathbf{Q} = \mathbf{D} - \mathbf{W}$  is an intrinsic conditionally autoregressive covariance matrix where  $\mathbf{D}$  is a matrix that has the counts of the number of neighbors of each location on the diagonal and  $\mathbf{W}_{ij}$  is 1 if location  $i$  is a neighbor of location  $j$  and 0 otherwise. The process can be written using a basis representation of the spectral decomposition of  $\mathbf{Q}$  as

$$\mathbf{T} \sim \mathcal{N}(\mathbf{Z}\boldsymbol{\alpha}, \sigma^2 \mathbf{I})$$

where  $\mathbf{Z}$  is the matrix with the eigenvectors of  $\mathbf{Q}$  as columns that represent the basis functions with random coefficients  $\alpha$  given by

$$\alpha \sim \mathcal{N}(\mathbf{0}, \mathbf{\Lambda})$$

where  $\mathbf{\Lambda}$  is a diagonal matrix of the eigenvalues of  $\mathbf{Q}$ .

### 1.3 Parameter Model

$$\begin{aligned}\alpha|\mathbf{Q} &\sim \mathcal{N}(\mathbf{0}, \mathbf{\Lambda},) \\ [\sigma^2] &\propto \frac{1}{\sigma^2} \\ [\tau_I^2] &\propto \frac{1}{\tau_I^2} \\ [\tau_P^2] &\propto \frac{1}{\tau_P^2} \\ \beta_0|\tau_P^2 &\sim \mathcal{N}(\mathbf{0}, \tau_P^2 \mathbf{\Delta}_0^{-1}) \\ \beta_1|\tau_P^2 &\sim \mathcal{N}(\mathbf{0}, \tau_P^2 \mathbf{\Delta}_1^{-1})\end{aligned}$$

## 2 Posterior

$$\begin{aligned}\prod_{t=1}^T [T_t, \beta_0, \beta_1, \alpha, \tau_I^2, \tau_P^2, \sigma^2 | \mathbf{W}_t] &\propto \prod_{t=1}^T [\mathbf{W}_t | T_t, \beta_0, \beta_1, \tau_I^2, \tau_P^2] [\mathbf{T} | \alpha, \sigma^2] \\ &\quad \times [\beta_0 | \tau_P^2] [\beta_1 | \tau_P^2] [\tau_I^2] [\tau_P^2] [\alpha | \mathbf{Q}] [\sigma^2]\end{aligned}$$

### 3 Appendix

#### 3.1 Full Conditional for $\mathbf{T}$

$$\begin{aligned}
[\mathbf{T}|\cdot] &\propto \prod_{t=1}^T [\mathbf{W}_t|T_t, \beta_0, \beta_1, \tau_I^2, \tau_P^2] [\mathbf{T}|\boldsymbol{\alpha}, \sigma^2] \\
&\propto \prod_{t=1}^T \exp \left\{ -\frac{1}{2} (\mathbf{W}_t - (\mathbf{H}_t T_t + \mathbf{B}_t))^T \boldsymbol{\Sigma}_t^{-1} (\mathbf{W}_t - (\mathbf{H}_t T_t + \mathbf{B}_t)) \right\} \\
&\quad \times \exp \left\{ -\frac{1}{2\sigma^2} (\mathbf{T} - \mathbf{Z}\boldsymbol{\alpha})^T (\mathbf{T} - \mathbf{Z}\boldsymbol{\alpha}) \right\} \\
&\propto \exp \left\{ -\frac{1}{2} \sum_{t=1}^T \left[ T_t^2 (\mathbf{H}_t^T \boldsymbol{\Sigma}_t^{-1} \mathbf{H}_t) - T_t (\mathbf{H}_t^T \boldsymbol{\Sigma}_t^{-1} (\mathbf{W}_t - \mathbf{B}_t)) \right] \right\} \\
&\quad \times \exp \left\{ -\frac{1}{2\sigma^2} [\mathbf{T}^T \mathbf{T} - 2\mathbf{T}^T \mathbf{Z}\boldsymbol{\alpha}] \right\} \\
&\propto \exp \left\{ -\frac{1}{2} \left[ \mathbf{T}^T \left( \tilde{\mathbf{A}} + \frac{1}{\sigma^2} \mathbf{I} \right) \mathbf{T} - 2\mathbf{T}^T \left( \tilde{\mathbf{b}} + \frac{\mathbf{Z}\boldsymbol{\alpha}}{\sigma^2} \right) \right] \right\}
\end{aligned}$$

which is  $\text{MVN}(\mathbf{A}^{-1}\mathbf{b}, \mathbf{A}^{-1})$  with

$$\begin{aligned}
\mathbf{A}^{-1} &= \left( \tilde{\mathbf{A}} + \frac{1}{\sigma^2} \mathbf{I} \right)^{-1} \\
\mathbf{b} &= \tilde{\mathbf{b}} + \frac{\mathbf{Z}\boldsymbol{\alpha}}{\sigma^2}.
\end{aligned}$$

where  $\tilde{\mathbf{A}} = \text{diag} \left( \mathbf{H}_1^T \boldsymbol{\Sigma}_1^{-1} \mathbf{H}_1, \dots, \mathbf{H}_T^T \boldsymbol{\Sigma}_T^{-1} \mathbf{H}_T \right)$  and  $\tilde{\mathbf{b}} = \left( \mathbf{H}_1^T \boldsymbol{\Sigma}_1^{-1} (\mathbf{W}_1 - \mathbf{B}_1), \dots, \mathbf{H}_T^T \boldsymbol{\Sigma}_T^{-1} (\mathbf{W}_T - \mathbf{B}_T) \right)^T$

### 3.2 Full conditional for $\beta_0$

$$\begin{aligned}
[\beta_0|\cdot] &\propto \prod_{t=1}^T [\mathbf{W}_t|T_t, \beta_0, \beta_1, \tau_I^2, \tau_P^2] [\beta_0|\tau_P^2] \\
&\propto \prod_{t=1}^T \exp \left\{ -\frac{1}{2} (\mathbf{W}_t - (\mathbf{H}_t T_t + \mathbf{B}_t))^T \boldsymbol{\Sigma}_t^{-1} (\mathbf{W}_t - (\mathbf{H}_t T_t + \mathbf{B}_t)) \right\} \\
&\quad \times \exp \left\{ -\frac{1}{2\tau_P^2} \beta_0^T \boldsymbol{\Delta}_0 \beta_0 \right\} \\
&\propto \exp \left\{ -\frac{1}{2} \left[ \beta_0^T \left( \sum_{t=1}^T \mathbf{H}_{P_t}^T \boldsymbol{\Sigma}_{P_t}^{-1} \mathbf{H}_{P_t} + \boldsymbol{\Delta}_0 \right) \beta_0 - 2\beta_0^T \left( \sum_{t=1}^T \mathbf{H}_{P_t}^T \boldsymbol{\Sigma}_{P_t}^{-1} (\mathbf{W}_{P_t} - \beta_1 \mathbf{H}_{P_t} T_t) \right) \right] \right\}
\end{aligned}$$

which is  $\text{MVN}(\mathbf{A}^{-1}\mathbf{b}, \mathbf{A}^{-1})$  with

$$\begin{aligned}
\mathbf{A}^{-1} &= \left( \sum_{t=1}^T \mathbf{H}_{P_t}^T \boldsymbol{\Sigma}_{P_t}^{-1} \mathbf{H}_{P_t} + \boldsymbol{\Delta}_0 \right)^{-1} \\
\mathbf{b} &= \sum_{t=1}^T \mathbf{H}_{P_t}^T \boldsymbol{\Sigma}_{P_t}^{-1} (\mathbf{W}_{P_t} - \beta_1 \mathbf{H}_{P_t} T_t).
\end{aligned}$$

### 3.3 Full conditional for $\beta_1$

$$\begin{aligned}
[\beta_1 | \cdot] &\propto \prod_{t=1}^T [\mathbf{W}_t | T_t, \beta_0, \beta_1, \tau_I^2, \tau_P^2] [\beta_1 | \tau_P^2] \\
&\propto \prod_{t=1}^T \exp \left\{ -\frac{1}{2} (\mathbf{W}_t - (\mathbf{H}_t T_t + \mathbf{B}_t))^T \boldsymbol{\Sigma}_t^{-1} (\mathbf{W}_t - (\mathbf{H}_t T_t + \mathbf{B}_t)) \right\} \\
&\quad \times \exp \left\{ -\frac{1}{2\tau_P^2} \beta_1^T \boldsymbol{\Delta}_1 \beta_1 \right\} \\
&\propto \exp \left\{ -\frac{1}{2} \beta_1^T \left( \sum_{t=1}^T (\mathbf{H}_{Pt} T_t)^T \boldsymbol{\Sigma}_{Pt}^{-1} (\mathbf{H}_{Pt} T_t) + \boldsymbol{\Delta}_1 \right) \beta_1 \right\} \\
&\quad \times \exp \left\{ -\beta_1^T \left( \sum_{t=1}^T (\mathbf{H}_{Pt} T_t)^T \boldsymbol{\Sigma}_{Pt}^{-1} (\mathbf{W}_{Pt} - \mathbf{H}_{Pt} \beta_0) \right) \right\} \\
&\propto \exp \left\{ -\frac{1}{2\tau_P^2} \beta_1^T \left( \sum_{t=1}^T (\mathbf{H}_{Pt} T_t)^T (\mathbf{H}_{Pt} T_t) + \boldsymbol{\Delta}_1 \right) \beta_1 \right\} \\
&\quad \times \exp \left\{ -\frac{1}{\tau_P^2} \beta_1^T \left( \sum_{t=1}^T (\mathbf{H}_{Pt} T_t)^T (\mathbf{W}_{Pt} - \mathbf{H}_{Pt} \beta_0) \right) \right\}
\end{aligned}$$

which is  $\text{MVN}(\mathbf{A}^{-1} \mathbf{b}, \mathbf{A}^{-1})$  with

$$\begin{aligned}
\mathbf{A}^{-1} &= \left( \frac{1}{\tau_P^2} \sum_{t=1}^T (\mathbf{H}_{Pt} T_t)^T (\mathbf{H}_{Pt} T_t) + \boldsymbol{\Delta}_1 \right)^{-1} \\
\mathbf{b} &= \frac{1}{\tau_P^2} \sum_{t=1}^T (\mathbf{H}_{Pt} T_t)^T (\mathbf{W}_{Pt} - \mathbf{H}_{Pt} \beta_0).
\end{aligned}$$

### 3.4 Full conditional for $\tau_I^2$

$$\begin{aligned}
[\tau_I^2 | \cdot] &\propto \prod_{t=1}^T [\mathbf{W}_t | T_t, \beta_0, \beta_1, \tau_I^2, \tau_P^2] [\tau_I^2] \\
&\propto \prod_{t=1}^T |\Sigma_t|^{-\frac{1}{2}} \exp \left\{ -\frac{1}{2} (\mathbf{W}_t - (\mathbf{H}_t T_t + \mathbf{B}_t))^T \Sigma_t^{-1} (\mathbf{W}_t - (\mathbf{H}_t T_t + \mathbf{B}_t)) \right\} \frac{1}{\tau_I^2} \\
&\propto (\tau_I^2)^{-\frac{1}{2} \sum_{t=1}^T N_{It} - 1} \exp \left\{ -\frac{1}{2} \sum_{t=1}^T \frac{(W_{It} - (H_{It} T_t + B_{It}))^2}{\tau_I^2} \right\} \\
&\propto (\tau_I^2)^{-\frac{1}{2} \sum_{t=1}^T N_{It} - 1} \exp \left\{ -\frac{1}{\tau_I^2} \left( \frac{1}{2} \sum_{t=1}^T (W_{It} - H_{It} T_t)^2 \right) \right\}
\end{aligned}$$

which is  $\text{IG}\left(\frac{1}{2} \sum_{t=1}^T N_{It}, \frac{1}{2} \sum_{t=1}^T (W_{It} - H_{It} T_t)^2\right)$

### 3.5 Full conditional for $\tau_P^2$

$$\begin{aligned}
[\tau_P^2 | \cdot] &\propto \prod_{t=1}^T [\mathbf{W}_t | T_t, \beta_0, \beta_1, \tau_I^2, \tau_P^2] [\tau_P^2] \\
&\propto \prod_{t=1}^T |\Sigma_t|^{-\frac{1}{2}} \exp \left\{ -\frac{1}{2} (\mathbf{W}_t - (\mathbf{H}_t T_t + \mathbf{B}_t))^T \Sigma_t^{-1} (\mathbf{W}_t - (\mathbf{H}_t T_t + \mathbf{B}_t)) \right\} \frac{1}{\tau_P^2} \\
&\propto (\tau_P^2)^{-\frac{1}{2} \sum_{t=1}^T N_{Pt} - 1} \\
&\quad \times \exp \left\{ -\frac{1}{\tau_P^2} \left( \frac{1}{2} \sum_{t=1}^T (\mathbf{W}_{Pt} - (\beta_1 \mathbf{H}_{Pt} T_t + \mathbf{H}_{Pt} \beta_0))^T (\mathbf{W}_{Pt} - (\beta_1 \mathbf{H}_{Pt} T_t + \mathbf{H}_{Pt} \beta_0)) \right) \right\}
\end{aligned}$$

which is  $\text{IG}\left(\frac{1}{2} \sum_{t=1}^T N_{Pt}, \frac{1}{2} \sum_{t=1}^T (\mathbf{W}_{Pt} - (\beta_1 \mathbf{H}_{Pt} T_t + \mathbf{H}_{Pt} \beta_0))^T (\mathbf{W}_{Pt} - (\beta_1 \mathbf{H}_{Pt} T_t + \mathbf{H}_{Pt} \beta_0))\right)$ .

### 3.6 Full conditional for $\alpha$

$$\begin{aligned}
[\alpha | \cdot] &\propto [\mathbf{T} | \alpha, \sigma^2] [\alpha | \mathbf{Q}] \\
&\propto \exp \left\{ -\frac{1}{2\sigma^2} (\mathbf{T} - \mathbf{Z}\alpha)^T (\mathbf{T} - \mathbf{Z}\alpha) \right\} \exp \left\{ -\frac{1}{2} \alpha^T \mathbf{\Lambda}^{-1} \alpha \right\} \\
&\propto \exp \left\{ -\frac{1}{2} \left[ \alpha^T \left( \frac{\mathbf{Z}^T \mathbf{Z}}{\sigma^2} + \mathbf{\Lambda}^{-1} \right) \alpha - 2\alpha^T \frac{\mathbf{Z}^T \mathbf{T}}{\sigma^2} \right] \right\}
\end{aligned}$$

which is  $N(\mathbf{A}^{-1}\mathbf{b}, \mathbf{A}^{-1})$  with

$$\mathbf{A}^{-1} = \left( \frac{\mathbf{Z}^T \mathbf{Z}}{\sigma^2} + \mathbf{\Delta}^{-1} \right)^{-1}$$

$$\mathbf{b} = \frac{\mathbf{Z}^T \mathbf{T}}{\sigma^2}.$$

### 3.7 Full conditional for $\sigma^2$

$$\begin{aligned} [\sigma^2 | \cdot] &\propto [\mathbf{T} | \boldsymbol{\alpha}, \sigma^2] [\sigma^2] \\ &\propto (\sigma^2)^{-\frac{T}{2}} \exp \left\{ -\frac{1}{2\sigma^2} (\mathbf{T} - \mathbf{Z}\boldsymbol{\alpha})^T (\mathbf{T} - \mathbf{Z}\boldsymbol{\alpha}) \right\} \frac{1}{\sigma^2} \\ &\propto (\sigma^2)^{-\frac{T}{2}-1} \exp \left\{ -\frac{1}{\sigma^2} \frac{(\mathbf{T} - \mathbf{Z}\boldsymbol{\alpha})^T (\mathbf{T} - \mathbf{Z}\boldsymbol{\alpha})}{2} \right\} \end{aligned}$$

which is  $\text{IG}\left(\frac{T}{2}, \frac{(\mathbf{T} - \mathbf{Z}\boldsymbol{\alpha})^T (\mathbf{T} - \mathbf{Z}\boldsymbol{\alpha})}{2}\right)$

## 4 Extensions to the model

- Predictive Process for  $\Sigma_\epsilon$  in the spatial model
- Include the number of trees used to create a chronology in the model
- ???

## 5 References

- Martin Tingley's BARCAST papers
- Sayer's discussion on Wishart priors <http://www.math.wustl.edu/~sawyer/hmhandouts/Wishart.pdf>