► The Process Convolution Model (as in the Practicum)...

 \rightarrow written hierarchically

$$\begin{split} \boldsymbol{Z}|\boldsymbol{Y},\tau^2 &\sim \mathsf{N}(\boldsymbol{Y},\tau^2\boldsymbol{I}) \\ \boldsymbol{Y}|\boldsymbol{\Psi}_1,\dots,\boldsymbol{\Psi}_n &\sim \mathsf{GP}(\boldsymbol{0},\boldsymbol{\Sigma}^Y(\boldsymbol{\Psi}_1,\dots,\boldsymbol{\Psi}_n)) \\ \mathsf{Kernel Parameters:} &\begin{cases} \psi_1^1,\dots,\psi_n^1|m_1,v_1,r_1 &\sim \mathsf{GP}(m_1\boldsymbol{1},\boldsymbol{\Sigma}^\psi(v_1,r_1)) \\ \psi_1^2,\dots,\psi_n^2|m_2,v_2,r_2 &\sim \mathsf{GP}(m_2\boldsymbol{1},\boldsymbol{\Sigma}^\psi(v_2,r_2)) \\ \psi_1^3,\dots,\psi_n^3|m_3,v_3,r_3 &\sim \mathsf{GP}(m_3\boldsymbol{1},\boldsymbol{\Sigma}^\psi(v_3,r_3)) \end{cases} \end{split}$$

where $\mathbf{\Sigma}^{Y}(\cdot)$ and $\mathbf{\Sigma}^{\psi}(\cdot)$ are matrix-valued functions of their inputs

 τ^2 . **m**. **v**. **r** \sim something

▶ Integrating out **Y**, the unknown parameters in the model are

$$\underbrace{\boldsymbol{\Psi}_{1}}_{3\times 1},\ldots,\underbrace{\boldsymbol{\Psi}_{n}}_{3\times 1},\tau^{2},\boldsymbol{m},\boldsymbol{v},\boldsymbol{r}$$

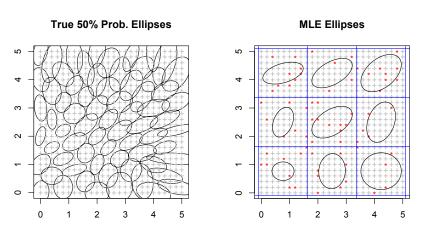
for a total of

$$3n + 1 + 3 + 3 + 3 = 2038$$

(highly dependent!) parameters in the simulated gridded data

Since we can't integrate out the Ψs analytically, MCMC will likely not work → the posterior distribution is highly structured and the dimension of the parameter space is large

ightharpoonup The good news... there appears to be information about the Ψ s in a single realization from the model



- ► Our thoughts:
 - 1. the GP priors on the elements of the Ψ vectors makes the model overly parameterized
 - 2. the Ψ s instead should perhaps be deterministic functions of a few unknown parameters
 - the model itself isn't bad, but more work is needed on prior specification & collapsing a layer of the hierarchy