

- Given a random field,  $\alpha(x, \omega)$ , we can write

$$\alpha(x, \omega) = \bar{\alpha}(x) + \sum_{i=1}^{\infty} \sqrt{\lambda_i} \phi_i(x) \Theta_i(\omega)$$

• Note  $\bar{\alpha}$ ,  $\lambda_i$ ,  $\phi_i$  are deterministic

•  $\Theta_i(\omega)$  is stochastic, i.e., if we assume  $\alpha$  is a Gaussian random field,  $\Theta_i \sim N(0, 1)$

-  $\lambda_i + \phi_i$  are eigenpairs corresponding  $C(x, x')$

Ex) Absolute exponential

$$C(x, x') = \sigma^2 \exp\left(-\frac{|x - x'|}{L}\right), \quad x \in [-1, 1]$$

• This works well when dealing with non differentiable random field

•  $L$  is the correlation length,  $\sigma^2$  is general Variance

- Note, the eigenvalues + eigenfunctions

$$\lambda_n = \sigma^2 \frac{2L}{1 + L^2 v_n^2}$$

where  $v_n$  solve

System of transcendental equations

$$\phi_n = \begin{cases} \frac{\sin(v_n x)}{(1 - \sin(2v_n)/2v_n)^{1/2}}, & n \text{ is even} \\ \frac{\cos(v_n x)}{(1 + \frac{\sin(2v_n)}{2v_n})^{1/2}}, & n \text{ is odd} \end{cases}$$

Note: If  $L \rightarrow \infty$ ,  $C(x, x') = \sigma^2$

$\Rightarrow$  "Full correlation"

$\Rightarrow$  you get one eigenvalue  $> 0$

- If  $L \rightarrow 0$ , no correlation

• To analyze, we set  $\sigma^2 = \frac{1}{2L(1 - \exp(-1/L))}$

$$\Rightarrow C(x, x') = \frac{1}{2L(1 - \exp(-1/L))} \cdot \exp\left(-\frac{|x-x'|}{L}\right)$$

$$\Rightarrow \lim_{L \rightarrow 0} C(x, x') = \delta(x - x')$$

- The eigenvalue problem becomes

$$\phi_n(x) = \lambda_n \phi_n(x) \Rightarrow \lambda_n = 1$$

\* So no information is communicated through  $C(x, x')$

Ex) Squared exponential (radial basis function)

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$$C(x, x') = \sigma^2 \exp\left(-\frac{(x-x')^2}{2L^2}\right), \quad x \in \mathbb{R}$$

• eigenvalues come from solving eigenvectors that are Hermitean Polynomials

$$\phi_n(x) = H_n(\sqrt{2\sigma} x)$$

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~ Consider the truncated random field

$$\alpha(x, \omega) \approx \bar{\alpha}(x) + \underbrace{\sum_{i=1}^N \sqrt{\lambda_i} \phi_i(x) \theta_i(\omega)}_{\alpha_N(x, \omega)}$$

- Note that  $\{\phi_1, \dots, \phi_N\} \xrightarrow{N \rightarrow \infty} \alpha$

- Note that 
$$E[||\alpha_n||^2]^{N \rightarrow \infty} = \sum_{i=1}^{\infty} \lambda_i$$

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- This definition was not so clear. Let's instead write the following:

- Assume  $\alpha(x, \omega)$  is zero mean (or assume we are working with  $\alpha_c(x, \omega)$ )
- By definition, orthogonality of  $\phi_i$  gives us

$$\begin{aligned} E[||\alpha||^2] &= \int_{\Omega} \left[ \int_D |\alpha(x, \omega)|^2 dx \right] \underbrace{dP(\omega)}_{\text{Probability measure}} \\ &= \int_{\Omega} \left[ \int_D \left| \sum_{i=1}^{\infty} \sqrt{\lambda_i} \phi_i \Theta_i \right|^2 \right] dP(\omega) \\ &= \int_{\Omega} \left[ \int_D \sum_{i=1}^{\infty} \lambda_i |\phi_i|^2 |\Theta_i|^2 \right] dP(\omega) \end{aligned}$$

(- Recall that  $E[\Theta_i \Theta_j] = \delta_{ij}$  &  $\phi_i$  are orthonormal)

$$\begin{aligned} &= \sum_{i=1}^{\infty} \lambda_i. \quad \text{we then assume } \exists \epsilon > 0 \text{ s.t. for } N < \infty \\ D_N &= \frac{E[||\alpha^N - \alpha||^2]}{E[||\alpha||^2]} = \frac{\sum_{i=N+1}^{\infty} \lambda_i}{\sum_{i=1}^{\infty} \lambda_i} \leq \epsilon^2 \end{aligned}$$

where  $N$  dictates our tolerance

- For sufficiently large  $N$ ,  $\alpha_N \Rightarrow \alpha - \bar{\alpha}$ , based on  $\lambda_i$  decay.

(Note: If  $\bar{\alpha} = 0$ ,  
 $\alpha_N \Rightarrow \alpha$ )

- If we assume  $\alpha(x, \omega)$  is a Gaussian Random field

then uncorrelated  $\Rightarrow$  independent

$$\Rightarrow \theta_i(\omega) \sim N(0, 1)$$

- Throughout, we have assumed a  $C(x, x')$  for  $\alpha(x, \omega)$ .

• The true  $C(x, x')$  is typically unknown

• One alternative is to construct an estimate of  $C(x, x')$  from samples of  $\alpha(x, \omega)$

$$\tilde{C}_{est} = \frac{1}{N-1} \sum (x - \bar{x})(x - \bar{x})^T$$