## A GP model for shear fields

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#### Basic premise

We model the 2D lens potential  $\psi$  as a Gaussian Process (GP) model,

$$\psi \sim GP(0, \Sigma),$$

where the GP covariance  $\Sigma(\vec{x}, \vec{y})$  is a function of two 2D position vectors to denote the covariance of the realizations of the lens potential at sky positions  $\vec{x}$  and  $\vec{y}$ .

#### Observable lensing quantities

- The lensing convergence and shear are 2nd derivatives of the lens potential.
- Because the derivative of a GP is also a GP, the vector of  $\mathbf{s} \equiv (\kappa, \gamma_1, \gamma_2)$  is GP distributed.
- The covariance of **s** is given by 4th derivatives of the covariance of  $\Sigma$  of  $\psi$ . That is,

$$Cov(\psi_{,ij}(\vec{x}), \psi_{,k\ell}(\vec{y})) = \Sigma_{,x_ix_iy_ky_\ell}(\vec{x}, \vec{y}).$$

#### GP covariance parameterizations

Generally, we will write the GP covariance as the product of a precision parameter  $\lambda$  and a 'kernel' k,

$$\Sigma(\vec{x}, \vec{y}) \equiv \lambda^{-1} k(r^2(\vec{x}, \vec{y}, D)),$$

where,

$$r^2 \equiv (\vec{x} - \vec{y})^T D(\vec{x} - \vec{y})$$

for some metric  $D^{-1}$ .

#### GP kernels in george

The george python package has a number of GP kernels already implemented. Two common interesting ones are:

1. Squared exponential kernel,

$$k(r^2) = \exp(-\frac{1}{2}r^2)$$

2. Matern kernels with parameter  $\nu$  (only  $\nu = (3/2, 5/2)$  implemented in george),

$$k_{\nu}(r^2) = \frac{1}{\Gamma(\nu)2^{\nu-1}} (2\nu r^2)^{\nu/2} K_{\nu}(\sqrt{2\nu r^2}),$$

where  $K_{\nu}$  is the modified Bessel function of the 2nd kind.

#### GP covariance derivatives

## General relations (1): derivatives of $r^2$

From before,  $r^2 \equiv (\vec{x}-\vec{y})^T D(\vec{x}-\vec{y}).$  Then,  $\frac{\partial r^2}{\partial x_i} = [D(\vec{x}-\vec{y})]_i \equiv X_i$   $\frac{\partial r^2}{\partial u_i} = -X_i$ 

$$\operatorname{Hess}_{ij}(r^2) = D_{ij} \left( \begin{array}{cc} 1 & -1 \\ -1 & 1 \end{array} \right)$$

## General relations (2a): derivatives of $\Sigma$

The 1st, 2nd, and 4th derivatives of  $\Sigma$  that we need are,

$$\begin{split} \Sigma_{,x_i} &= \lambda^{-1} k^{'} r_{,x_i}^2 \\ \Sigma_{,x_ix_j} &= \lambda^{-1} \left[ k^{''} r_{,x_j}^2 r_{,x_i}^2 + k^{'} r_{,x_ix_j}^2 \right] \\ \Sigma_{,x_ix_jy_ky_\ell} &= \lambda^{-1} [k^{''''} r_{,x_i}^2 r_{,y_k}^2 r_{,y_\ell}^2 + k^{'''} \left( r_{,x_ix_j}^2 r_{,y_k}^2 r_{,y_\ell}^2 + 5 \text{perm.} \right) + k^{''} \left( r_{,x_ix_j}^2 r_{,y_ky_\ell}^2 + 2 \text{perm.} \right) ] \end{split}$$

where primes on k denote derivatives with respect to  $r^2$ .

#### General relations (2b): derivatives of $\Sigma$ (simplified)

$$\Sigma_{,(x,y)_{i}} = \pm \lambda^{-1} k^{'} X_{i}$$
 
$$\Sigma_{,x_{i}x_{j}} = \lambda^{-1} \left[ k^{''} X_{i} X_{j} + k^{'} D_{ij} \right]$$
 
$$\Sigma_{,x_{i}x_{j}} y_{k} y_{\ell} = \lambda^{-1} [k^{''''} X_{i} X_{j} X_{k} X_{\ell} + k^{'''} \left( D_{ij} X_{k} X_{\ell} + 5 \text{perm.} \right) + k^{''} \left( D_{ij} D_{k\ell} + 2 \text{perm.} \right) \right]$$

#### Observations about GP covariances of convergence and shear

- The derivatives of  $\Sigma$  have terms that depend on  $X_i \equiv [D(\vec{x} \vec{y})]_i$ , which *increases* in magnitude with increasing separation on the sky. So, distant points can be highly correlated.
- In our nominal model,  $D_{ij} \propto -\ln(\rho)$  where  $\rho$  is the correlation parameter for the GP.
- At small sky separations  $X_i \approx 0$  and the covariance of the convergence or shear  $\sim D_{ij}^2 \Sigma$ . So, the variance of the convergence and shear is enhanced by a factor  $[\ln(\rho)]^2$  over that of the lens potential.

## Derivatives of specific GP kernels

## Kernel derivatives: Squared exponential

$$k(r^2) \equiv \exp(-\frac{1}{2}r^2)$$

Because there's just the exponential term, the nth derivatives are trivial,

$$k^{(n)} = \left(-\frac{1}{2}\right)^n k$$

### Final GP covariance derivatives: Squared exponential

$$\Sigma_{,(x,y)_i} = \mp \frac{1}{2} X_i \Sigma$$

$$\Sigma_{,x_i x_j} = \left[ \frac{1}{4} X_i X_j - \frac{1}{2} D_{ij} \right] \Sigma$$

$$\Sigma_{,x_i x_j y_k y_\ell} = \left[ \frac{1}{16} X_i X_j X_k X_\ell - \frac{1}{8} \left( D_{ij} X_k X_\ell + \text{5perm.} \right) + \frac{1}{4} \left( D_{ij} D_{k\ell} + \text{2perm.} \right) \right] \Sigma$$

### Kernel derivatives: Matern (1)

$$k_{\nu}(r^2) = \frac{1}{\Gamma(\nu)2^{\nu-1}} (2\nu r^2)^{\nu/2} K_{\nu}(\sqrt{2\nu r^2}),$$

The *n*th derivatives of the modified Bessel function of the 2nd kind can be represented by the Meijer G function. But, george only implements the cases  $\nu = 3/2, 5/2$  and the kernel has a simpler form for these cases,

$$k_{\nu}(r^2) = p_{\nu}(r) \exp\left(-\sqrt{2\nu r^2}\right)$$

where  $p_{\nu}$  is a polynomial in  $\sqrt{r^2}$ .

We'll consider the 1st-4th derivatives of this expression next.

# Lensing parameters

#### Convergence and shear from lens potential

$$\kappa = \frac{1}{2} (\psi_{,11} + \psi_{,22})$$

$$\gamma_1 = \frac{1}{2} (\psi_{,11} - \psi_{,22})$$

$$\gamma_2 = \frac{1}{2} (\psi_{,12} + \psi_{,21})$$

# Convergence and shear GP covariances

$$Cov(\kappa) = \frac{1}{4} \left( \Sigma_{,1111} + \Sigma_{,1122} + \Sigma_{,2211} + \Sigma_{,2222} \right)$$

$$Cov(\gamma_1) = \frac{1}{4} \left( \Sigma_{,1111} - \Sigma_{,1122} - \Sigma_{,2211} + \Sigma_{,2222} \right)$$

$$Cov(\gamma_2) = \frac{1}{4} \left( \Sigma_{,1212} + \Sigma_{,1221} + \Sigma_{,2112} + \Sigma_{,2121} \right)$$