### 1. Prove the results about the smoothness of the members of the Matérn family.

We use the theorem that if

$$\frac{d^{2\nu}}{d\tau^{2\nu}}\rho(\tau)$$

exists and is finite at  $\tau = 0$ , then the random field having  $\rho(\tau)$  as its correlation function is  $\nu$  times differentiable at 0.

Without loss of generality, let  $\phi = 1$ . The Matérn correlation function is given by

$$\rho(\tau) = \frac{\tau^{\nu}}{2^{\nu-1}\Gamma(\nu)} K_{\nu}(\tau), \qquad \tau \ge 0, \nu > 0.$$

For small  $\tau$  and  $\nu > 0$ ,  $K_{\nu}(\tau) \approx \Gamma(\nu) 2^{\nu-1} \tau^{-\nu}$ . Also,  $\frac{d}{d\tau} \tau^{\nu} K_{\nu}(\tau) = -\tau^{\nu} K_{\nu-1}(\tau)$  and  $K_{\nu}(\tau) = K_{-\nu}(\tau)$ . We will be taking  $\tau \to 0$ , so we use the approximation for  $K_{\nu}(\tau)$ . This leads to the derivative,

$$\frac{d}{d\tau}\rho(\tau) = -\frac{\tau^{\nu}}{2^{\nu-1}\Gamma(\nu)}K_{\nu-1}(\tau) 
= \begin{cases}
-\frac{\tau^{\nu}}{2^{\nu-1}\Gamma(\nu)}K_{\nu-1}(\tau), & \nu - 1 \ge 0 \\
-\frac{\tau^{\nu}}{2^{\nu-1}\Gamma(\nu)}K_{1-\nu}(\tau), & \nu - 1 < 0
\end{cases} 
\approx \begin{cases}
-\frac{\tau^{\nu}}{2^{\nu-1}\Gamma(\nu)}\Gamma(\nu - 1)2^{\nu-2}\tau^{-\nu+1}, & \nu - 1 \ge 0 \\
-\frac{\tau^{\nu}}{2^{\nu-1}\Gamma(\nu)}\Gamma(1 - \nu)2^{-\nu}\tau^{\nu-1}, & \nu - 1 < 0
\end{cases} 
\approx \begin{cases}
-\tau G_1(\nu), & \nu - 1 \ge 0 \\
-\tau^{2\nu-1}G_2(\nu), & \nu - 1 < 0
\end{cases} .$$

Therefore,

$$\rho'(0) \begin{cases} = 0, & \nu \ge 1 \\ \in (-\infty, 0), & 1/2 \le \nu < 1. \\ = -\infty, & 0 < \nu < 1/2 \end{cases}$$

The second derivative is given by

$$\rho''(\tau) = \begin{cases} \frac{-\tau^{\nu-1} K_{\nu-1}(\tau) + \tau^{\nu} K_{\nu-2}(\tau)}{2^{\nu-1} \Gamma(\nu)}, & \nu \ge 2\\ \frac{-\tau^{\nu-1} K_{\nu-1}(\tau) + \tau^{\nu} K_{2-\nu}(\tau)}{2^{\nu-1} \Gamma(\nu)}, & 1 \le \nu < 2,\\ \frac{-\tau^{\nu-1} K_{1-\nu}(\tau) + \tau^{\nu} K_{2-\nu}(\tau)}{2^{\nu-1} \Gamma(\nu)}, & 0 < \nu < 1 \end{cases}$$

and is evaluated at  $\tau = 0$  to

$$\rho''(0) \begin{cases} = 0, & \nu \ge 2 \\ \in (-\infty, 0), & 1 \le \nu < 2 \\ = -\infty, & 0 < \nu < 1 \end{cases}$$

We have that the second derivative is finite when  $\nu \geq 1$ , leading to a random field that is one time mean square differentiable. To show this generalizes to  $\nu \geq d$ , we need to keep taking derivatives of  $\rho(\tau)$ . I suspect that on each even derivative, the orders of certain Bessel functions are negated when those orders are less than  $\nu$ . When this happens, the approximation will lead to a term having  $\tau$  raised to a negative exponent causing the derivative to evaluate to  $-\infty$ .

#### 2. Use the spectral representation to show that the product of two valid correlation functions is a valid correlation function.

A valid correlation function is the characteristic function of some random variable,

$$\rho(\boldsymbol{\tau}) = E\left[e^{i\boldsymbol{\tau}^{\top}\mathbf{X}}\right].$$

Suppose we have two valid correlation functions  $\rho_1(\tau)$  and  $\rho_2(\tau)$  associated with independent random variables  $\mathbf{X}_1$  and  $\mathbf{X}_2$ , respectively. Then the product is written

$$\rho(\boldsymbol{\tau}) = \rho_1(\boldsymbol{\tau})\rho_2(\boldsymbol{\tau}) = E\left[e^{i\boldsymbol{\tau}^{\top}\mathbf{X}_1}\right]E\left[e^{i\boldsymbol{\tau}^{\top}\mathbf{X}_2}\right]$$
$$= E\left[e^{i\boldsymbol{\tau}^{\top}\mathbf{X}_1}e^{i\boldsymbol{\tau}^{\top}\mathbf{X}_2}\right]$$
$$= E\left[e^{i\boldsymbol{\tau}^{\top}(\mathbf{X}_1 + \mathbf{X}_2)}\right],$$

so  $\rho$  is the characteristic function of  $\mathbf{X}_1 + \mathbf{X}_2$  and thus the product of two valid correlation functions is a valid correlation function. Note, our assumption of independence for  $\mathbf{X}_1$  and  $\mathbf{X}_2$  presents no issues.  $\mathbf{X}_1$  and  $\mathbf{X}_2$  may be dependent, but then we could simply define new independent random variables  $\mathbf{Y}_1$  and  $\mathbf{Y}_2$  with the same marginal distributions as  $\mathbf{X}_1$  and  $\mathbf{X}_2$ , resulting in the same correlation functions in either case.

3. The spectral density of a correlation in the Matérn family has tails whose thickness depends on the smoothness parameter. Conjecture: the smoothness of the corresponding random field depends on the number of moments of the spectral density. What can you say about this conjecture?

For correlation function

$$\rho(\tau) \propto (a\tau)^{\nu} K_{\nu}(a\tau), \qquad \tau \ge 0, \nu > 0, a = 1/\phi > 0,$$

we have the corresponding spectral density

$$f(x) \propto \frac{1}{(1 + (x/a)^2)^{\nu + n/2}},$$

where n is the dimension  $\tau$  (and x). This density has a form comparable to the t-distribution. Using integration by parts, we calculate the kth moment as

$$E(X^k) \propto \int x^k (1 + (x/a)^2)^{-(\nu + n/2)} dx$$

$$= -\frac{a^2}{2\nu + n - 2} \frac{x^{k-1}}{(1 + (x/a)^2)^{(2\nu + n - 2)/2}} \Big|_{-\infty}^{\infty} + \int \frac{(k-1)a^2}{2\nu + n - 2} \frac{x^{k-2}}{(1 + (x/a)^2)^{-(2\nu + n - 2)/2}} dx.$$

The first term (and hence the second term also) will be finite when  $2\nu + n - 2 \ge k - 1$ , or  $\nu \ge (k - n + 1)/2$ . In one dimension, n = 1, we see that when  $\nu \ge k/2$  the kth moment exists. This may be related to the theorem used in the first problem, that we need to have 2d-differentiable correlation function to have a d-differentiable random field. Here, we need the 2dth moment to exist, i.e.  $\nu \ge d$ , to have smoothness.

## 4. Use the K-L representation to approximate the exponential correlation for range parameter equal to 1. Plot the approximation for several orders and compare to the actual correlation.

The Karhunen-Loeve representation for a random process is given by

$$X(s) = \sum_{j=1}^{\infty} \sqrt{\lambda_j} \psi_j(s) Z_j$$

where  $Z_j$  are mean zero processes with  $E(Z_jZ_k)=\delta_{jk}$  and  $\psi_j$  is a basis of orthogonal functions such that

$$\int \psi_j(s) \overline{\psi_k(s)} ds = \delta_{jk}$$

and

$$\int C(s, s')\psi_j(s)ds = \lambda_j\psi_j(s').$$

For an exponential correlation function with  $\phi = 1$  on an interval [-L, L], we have

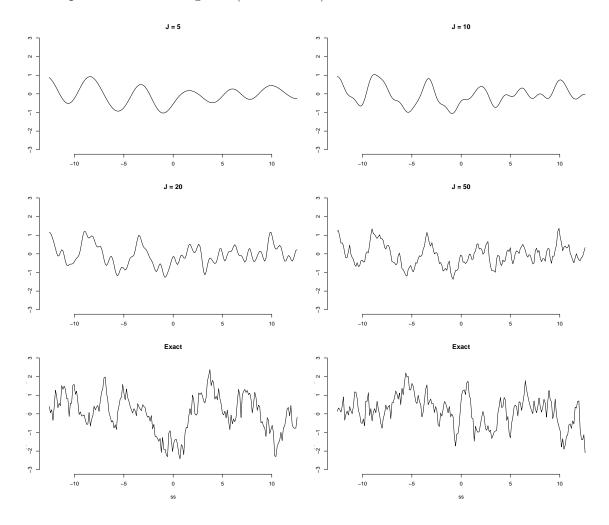
$$\lambda_i = \begin{cases} 2/(1 + v_j^2), & i = 2j + 1\\ 2/(1 + w_k^2), & i = 2k \end{cases}$$

$$\psi_i(s) = \begin{cases} \frac{\cos(v_j s)}{\sqrt{L + \sin(2v_j L)/(2v_j)}}, & i = 2j + 1\\ \frac{\sin(w_k s)}{\sqrt{L - \sin(2w_k L)/(2w_k)}}, & i = 2k \end{cases}$$

where  $v_i$  and  $w_k$  are the non-negative solutions to

$$1 - v \tan(vL) = 0$$
$$w + \tan(wL) = 0$$

sorted in increasing order. We end the approximation at J=5,10,20,50 summands. We let  $Z_j \sim N(0,1), \ j=1,\ldots,J$  and  $L=2\pi$  and obtain realizations of X(s) on a grid  $s\in [-4\pi,4\pi]$ . We also obtain realizations with geoR's grf() function which produces realizations from a Gaussian process. One draw for each J is shown (top four) and two Gaussian process draws are given (bottom two).



### 5. Repeat for the approximation given on Page 13 of the fifth set of slides.

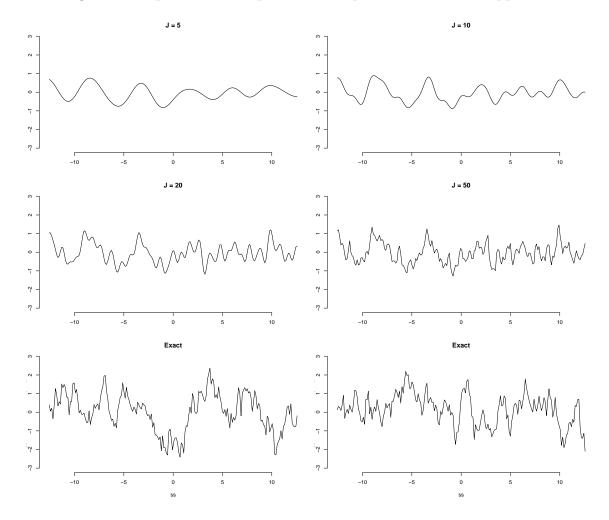
The approximation of interest is given by

$$\lambda_j \approx f(j\pi/(2L)),$$
  $\psi_j(s) \approx ce^{ij\pi s/(2L)}$ 

where f(k) is the spectrum at k (and we set c=1). Since the exponential correlation is equivalent to the Matérn with  $\nu=1/2$ , the spectral density, for  $\phi=1$ , is given by

$$f(k) = \frac{1}{(1+k^2)^{(n+1)/2}}$$

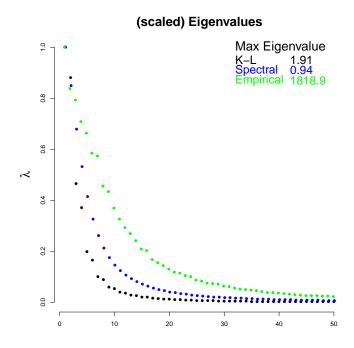
The next figure is comparable to the previous, except with this second approximation.



The same seed for generating the random draws was used in both problems 4 and 5. As we can see, the approximations are effectively identical. Both approaches reveal their inadequacies when J is too low, but at just J = 50, there is little distinct when obtaining realizations from the definition.

# 6. Generate 100 realizations of a univariate Gaussian process with exponential correlation with range parameter 1. Compare the empirically estimated eigenvalues and eigenfunctions to the ones given by the K-L and the approximation on Page 12.

Empirical estimates for the eigenvalues and eigenfunctions of 100 realizations were obtained via SVD. Here, we used a cutoff at J=50 for the approximations. We see a substantial difference between all three eigenvalues and eigenfunctions. The eigenvalues are given in the first plot. These are scaled so the first (and largest) eigenvalue is 1 (for plotting purposes).



The last plot shows the eigenvectors associated with the first six eigenvalues. The are plotted against  $s \in [-4\pi, 4\pi]$ . The periodicity is much more evident in the approximations (perhaps due to construction). As in the first plot, black is K-L representation, blue is the approximation using the spectral density, and green is the empirical estimates from SVD.

