## 2. Gaussian Processes

The file KernelCode.R distributed with the exam contains code to construct a kernlab function for the Matern covariance function with  $\nu = 3/2$ :

$$k(\mathbf{x}, \mathbf{x}') = \sigma_f^2 \left( 1 + \frac{\sqrt{3}r}{\ell} \right) \exp\left( -\frac{\sqrt{3}r}{\ell} \right).$$

where  $r = |\mathbf{x} - \mathbf{x}'|$ .

- (a) Let  $f \sim GP(0, k(\mathbf{x}, \mathbf{x}'))$  a priori and let  $\sigma_f^2 = 1$  and  $\ell = 0.5$ . Plot k(0, z) as a function of z. You can use the grid  $\mathbf{zGrid} = \mathbf{seq}(0.01, 1, \mathbf{by=0.01})$  for the plotting. Carefully interpret this plot in a way that shows that you understand what it shows. Connect your discussion to the smoothness of f. Finally, repeat this exercise with  $\sigma_f^2 = 0.5$  and discuss the effect this change has on the distribution of f (2 p.)
- (b) The file lidar.RData contains two variables logratio and distance. Load the variables into memory with the R command load("lidar.RData"). Compute the posterior distribution of f in the model

$$logratio = f(distance) + \varepsilon, \quad \varepsilon \sim N(0, 0.05^2).$$

You should do this for both length scales  $\ell = 1$  and  $\ell = 5$ . Set  $\sigma_f = 1$ . Your answer should be in the form of a scatter plot of the data overlayed with curves for

- i. the posterior mean of f
- ii. 95% probability intervals for f
- iii. 95% prediction intervals for a new data point y

Use the gausspr function in the kernlab package for i), but not for ii) and iii).

Discuss the differences in results from using the two length scales.

Do you think a GP with a Matern( $\nu = 3/2$ ) kernel is a good model for this data? If not, what could be the problem with this model? Can you think of a better model? Discuss.

[Hint:  $Cov(f) = K(X_{\star}, X_{\star}) - K(X_{\star}, X) \left[K(X, X) + \sigma_n^2 I\right]^{-1} K(X, X_{\star})$  and remember that %\*% does matrix multiplication and solve computes inverses in R] (2 p)

(c) (No need to do any computations here). Discuss how a Bayesian would handle the case where the kernel hyperparameters are unknown. What if the noise variance is unknown? (1 p)