Gaussian Processes

Robin Aldridge-Sutton

07/05/2022

```
# Functions to sample from and predict values of a Gaussian process.
source("GP funcs.R")
```

A Gaussian process (GP) is stochastic process (a distribution over functions) such that for any finite set of input values the function values have a multivariate Gaussian distribution.

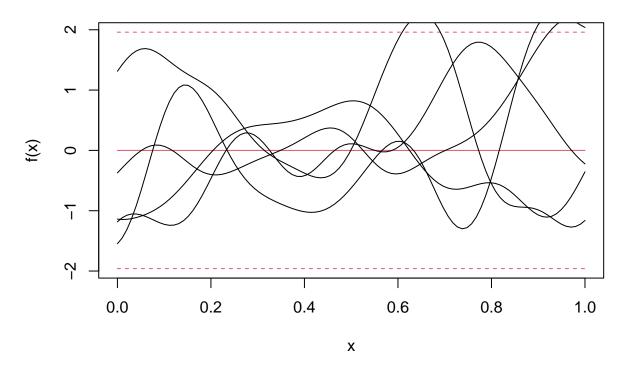
A GP can be used as a functional prior, e.g.

$$f(\mathbf{x}) \sim N(\mu(\mathbf{x}) = \mathbf{0}, \Sigma = K(\mathbf{x}, \mathbf{x})),$$
$$K(\mathbf{x}, \mathbf{x}')_{i,j} = \sigma_f^2 \exp\left(\frac{(x_i - x_j')^2}{2l^2}\right).$$

For some reason too many samples with too large of a length scale gives an error in the Cholesky factorization of the covariance matrix. Function sd makes no difference.

```
# Plot samples from a GP
plot_samp(
    n_samp = 100, # Number of samples per realisation
    n_real = 5, # Number of realisations
    l = 0.1, # Length scale
    sigma_f = 1 # Function standard deviation
)
```

Samples from a Gaussian process



The posterior is then the conditional distribution of functions given observed function values at a set of input values. This has a simple analytical form.

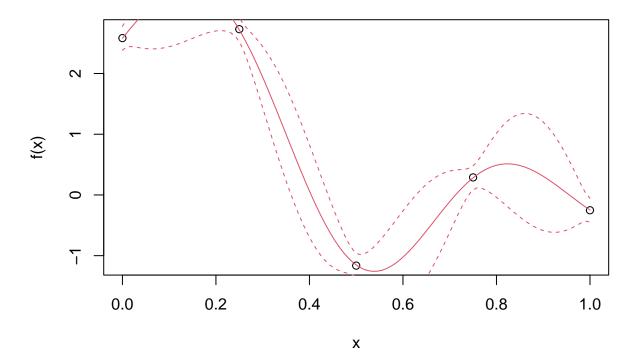
$$f(\mathbf{x})|f(\mathbf{x}') \sim N(\mu(\mathbf{x}), \Sigma),$$

$$\mu(\mathbf{x}) = K(\mathbf{x}', \mathbf{x})K(\mathbf{x}, \mathbf{x})^{-1}f(\mathbf{x}'),$$

$$\Sigma = K(\mathbf{x}', \mathbf{x}') - K(\mathbf{x}', \mathbf{x})K(\mathbf{x}, \mathbf{x})^{-1}K(\mathbf{x}, \mathbf{x}').$$

```
plot_GP(
    n_samp = 5, # Number of points to sample
    n_pred = 100, # Number of data points to predict
    l = 0.2, # Length scale
    sigma_f = 2, # Function standard deviation
    # sigma_n = 0.1 # Noise standard deviation
    sigma_n = 0.1 # Noise standard deviation
)
```

Gaussian process regression



$$\mathbf{y} = f(\mathbf{x}) + \epsilon,$$

$$\epsilon \sim N(0, \sigma_n^2 I_d)$$

$$\mathbf{x} \in \mathbb{R}^d$$

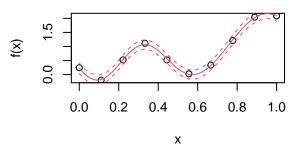
```
par(mfrow = c(2, 2))

for (i in 1:4)
  plot_GP(
    n_samp = 10, # Number of points to sample
    n_pred = 100, # Number of data points to predict
    1 = 0.2, # Length scale
    sigma_f = 2, # Function standard deviation
    sigma_n = 0.1 # Noise standard deviation
)
```

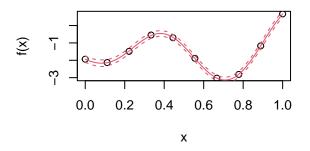
Gaussian process regression

X 0 0 0.0 0.2 0.4 0.6 0.8 1.0 x

Gaussian process regression



Gaussian process regression



Gaussian process regression

