

Supervision Assignments in MLBI

Easter Term 2019

Set 4

Supervisor : Dionysis Manousakas
dm754@cam.ac.uk

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1. (*Bayesian linear and Gaussian process regression.*)

Plot the time series of monthly mean global CO₂ concentrations obtained from the file `co2.txt` (original data obtained from <http://www.esrl.noaa.gov/gmd/ccgg/trends>).

We will apply Bayesian linear and Gaussian process regression to predict the CO₂ concentration $f(t)$ as a function of time t , where $t = \text{Year} + (\text{Month} - 1) = 12$.

- (a) First we model the function using linear regression, that is, using the functional form

$$f(t) = at + b + \epsilon(t)$$

with i.i.d. noise residual $\epsilon(t) \sim N(0, 1)$ and prior $a \sim N(0, 10^2)$, $b \sim N(360, 100^2)$. Compute (e.g. using MATLAB) the posterior mean and covariance over a and b given the CO₂ data.

- (b) Let a_{MAP} , b_{MAP} be the MAP estimate in the question above. The residual is the difference between the observed function values and the predicted mean function values

$$g_{\text{obs}}(t) = f_{\text{obs}}(t) - (a_{\text{MAP}}t + b_{\text{MAP}}),$$

where $f_{\text{obs}}(t)$ is the observed value of the CO₂ concentration at time t . Plot $g_{\text{obs}}(t)$. Do you think these residuals conform to our prior over $\epsilon(t)$? State, with justifications, which characteristics of the residual you think do or do not conform to our prior belief.

- (c) Write a MATLAB function to generate samples drawn from a GP. Specifically, given a covariance kernel function $k(\cdot, \cdot)$ and a vector of input points \mathbf{x} , return a function $f(\mathbf{x})$ evaluated on the input points \mathbf{x} drawn randomly from a GP with the given covariance kernel and with zero mean.
- (d) Test your function by plotting sample functions drawn from the following kernel, for various settings of the hyperparameters

$$k(s, t) = \theta^2 \left(\exp \left(- \frac{2 \sin^2(\pi(s-t)/\tau)}{\sigma^2} \right) + \phi^2 \exp \left(- \frac{(s-t)^2}{2\eta^2} \right) \right) + \zeta^2 \delta_{s=t}$$

Describe the characteristics of the drawn functions, and how the characteristics of the functions depend on the parameters.

- (e) Suppose we were to consider modelling the residual function $g(t)$ using a zero mean GP with the covariance kernel above. Based on the plot of $g(t)$ and your explorations in the preceding part, what do you think will be suitable values for the hyperparameters of k ?

- (f) Extrapolate the CO₂ concentration levels to 2020 using the GP with covariance kernel k of the equation in (d), and your chosen parameter values. Specifically, compute the predictive mean and variance of the residual $g(t)$ for every month between September 2007 and December 2020 given the observed residuals $g_{\text{obs}}(t)$. Plot the means and one standard deviation error bars of the extrapolated CO₂ concentration levels

$$f(t) = a_{\text{MAP}}t + b_{\text{MAP}} + g(t)$$

along with the observed CO₂ levels. Does the behaviour of the extrapolation conform to your expectations? How sensitive are your conclusions to settings of the kernel hyperparameters?

- (g) Why is the above procedure not Bayesian? How would we go about modelling $f(t)$ in a Bayesian framework?