

Lab 4

STAT 517 - Winter 2023

1. Let $\{X(t) : t \geq 0\}$ be a standard Brownian motion and $M(t) = \max_{0 \leq s \leq t} X(s)$. Show that

$$P(M(t) > a \mid M(t) = X(t)) = e^{-a^2/2t}$$

2. Let $\{X(t) : t \geq 0\}$ be a standard Brownian motion and define

$$V(t) = e^{-\alpha t^2/2} X(\alpha e^{-\alpha t})$$

- (a) Show that $\{V(t) : t \geq 0\}$ is a stationary Gaussian process. This is called the Ornstein-Uhlenbeck process.
 - (b) Show that $\{V(t) : t \geq 0\}$ is a Markov process.
 - (c) Show that the joint distribution of $(V(t_1), V(t_2), \dots, V(t_k))$ for an equally spaced grid with $t_i = t_{i-1} + \Delta$ corresponds to that of a first order autoregressive process in discrete time. Why is this useful when modeling real data?
3. Consider the dataset `mcycle` in the R package `MASS`, which corresponds to a series of measurements of head acceleration in a simulated motorcycle accident, used to test crash helmets

```
library(MASS)
data(mcycle)
plot(mcycle)
```

- (a) Fit Gaussian process models with constant mean function $f(x) = \mu$ and the following covariance matrices to this data:
 - i. $\gamma(x, x') = \sigma^2 e^{-\lambda|x-x'|}$.
 - ii. $\gamma(x, x') = \sigma^2 \left(\nu e^{-\lambda|x-x'|} + \delta_{(x=x')} \right)$ with $\nu = 1/3$.
 - iii. $\gamma(x, x') = \sigma^2 \left(\nu e^{-\lambda|x-x'|} + \delta_{(x=x')} \right)$ with $\nu = 10$.
 - iv. $\gamma(x, x') = \sigma^2 e^{-\lambda(x-x')^2}$.
 - v. $\gamma(x, x') = \sigma^2 \left(\nu e^{-\lambda(x-x')^2} + \delta_{(x=x')} \right)$ with $\nu = 1/3$.

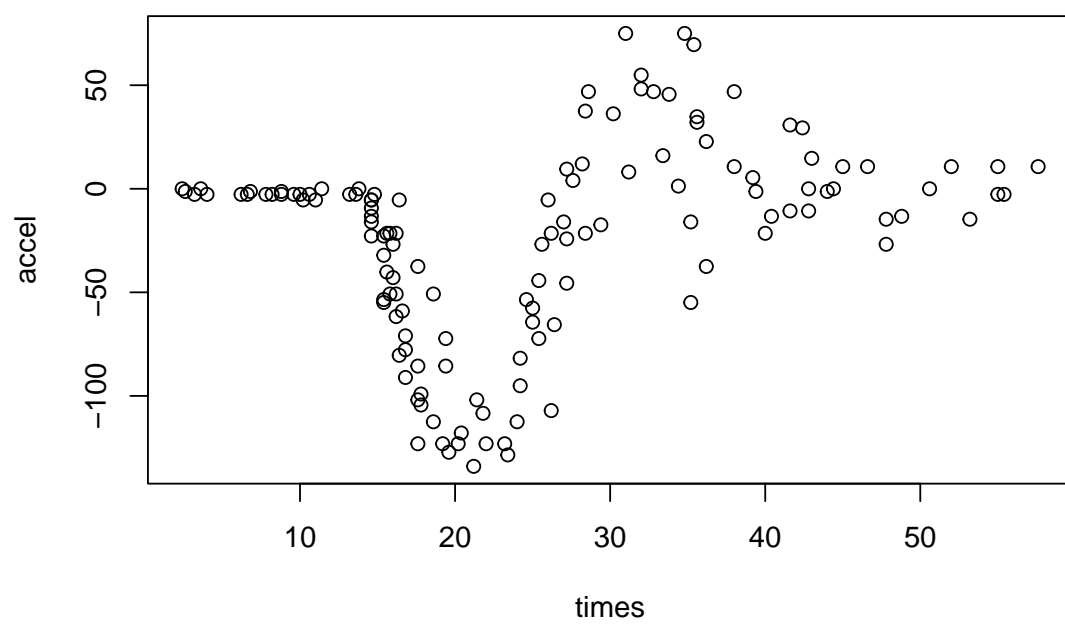


Figure 1: mcycle data.

where

$$\delta_{(x=x')} = \begin{cases} 1 & x = x' \\ 0 & \text{otherwise} \end{cases}$$

In each case, generate a plot of the estimated function in addition to providing point estimators for each of the parameters involved.

- (b) Comment on the different fits and discuss what the effect of the different model parameters is on the resulting estimate