

CDCI 567-Machine Learning Assignment-1

1 Question-1

1.1 MLE

1.1.1 Beta($\alpha, 1$)

x_1, x_2, \dots, x_n are samples of $Beta(\alpha, 1)$ distribution.

In this case, pdf of $x(i), i = 1, \dots, n$ is

$$f(x_i|\alpha, 1) = \frac{x^{\alpha-1}}{\beta(\alpha, 1)} = \frac{x^{\alpha-1}}{\int_0^1 t^{\alpha-1} dt} = \alpha x^{\alpha-1}$$

Likelihood function and log likelihood function:

$$L(\alpha, 1|x) = \prod_{i=1}^n \alpha x^{\alpha-1}$$

$$l(\alpha, 1|x) = \log(\alpha^n) + \log\left(\prod_{i=1}^n x_i^{\alpha-1}\right) = n\log(\alpha) + \sum_{i=1}^n (\alpha - 1)\log(x_i)$$

Derivative of $l(\alpha, 1|x)$:

$$\frac{\partial l(\alpha, 1|x)}{\partial \alpha} = \frac{n}{\alpha} + \sum_{i=1}^n \log(x_i) = 0$$

$$\frac{\partial^2 l(\alpha, 1|x)}{\partial \alpha^2} = -\frac{n}{\alpha^2} < 0$$

$$\hat{\alpha} = -\frac{n}{\sum_{i=1}^n \log(x_i)}$$

1.1.2 Normal($\theta, \text{diag}(\theta)$)

x_1, x_2, \dots, x_n are samples of $Normal(\theta, \text{diag}(\theta))$ distribution. In this case, pdf of $x_i, i = 1, \dots, n$ is

$$f(x_i|\theta, \text{diag}(\theta)) = \frac{1}{\sqrt{2\pi \text{diag}(\theta)}} \exp\left(-\frac{(x_i - \theta)}{2\theta}\right)$$

Likelihood function and log likelihood function:

$$L(\theta, \text{diag}(\theta)|x_i) = (2\pi)^{-\frac{d}{2}} |\text{diag}(\theta)|^{-\frac{1}{2}} \exp\left(-\frac{1}{2}(x - \theta)^T \text{diag}(\theta)^{-1}(x - \theta)\right)$$

$$|diag(\theta)| = \theta^d$$

$$l(\theta, diag(\theta)|x_i) = -\frac{d}{2}\log(2\pi\theta) - \sum_{i=1}^n \frac{1}{2} \frac{(x_i - \theta_i)^T (x_i - \theta_i)}{\theta_i}$$

Derivative of $l(\theta, diag(\theta)|x_i)$:

$$\frac{\partial l(\theta, diag(\theta)|x_i)}{\partial \theta} = -\frac{dn}{2\theta} - \frac{n}{2} + \frac{1}{2\theta^2} \sum_{i=1}^n x_i^2 = 0$$

$$\hat{\theta} = -\frac{d}{2} + \frac{1}{2} \sqrt{d + \frac{4}{n} \sum_{i=1}^n x_i^2}$$

1.2 Kernel Density Estimation

Random variables X_1, \dots, X_n are i.i.d. sampled from $f(x)$ function.

$$E_{X_1, \dots, X_n}[\hat{f}(x)] = E\left[\frac{1}{nh} \sum_{i=1}^n K\left(\frac{x-t}{h}\right)\right] = \frac{1}{nh} \sum_{i=1}^n E\left[K\left(\frac{x-t}{h}\right)\right] = \frac{1}{nh} n E\left[K\left(\frac{x-t}{h}\right)\right]$$

$$E_{X_1, \dots, X_n}[\hat{f}(x)] = \frac{1}{h} \int K\left(\frac{x-t}{h}\right) dt$$

$$f(x-hz) = f(x) + f'(x)(x-hz-x) + \frac{f''(x)}{2!}(x-hz-x)^2 + \dots + \frac{f^k(x)}{k!}(x-hz-x)^k + h_k(x-hz)(x-hz-x)^k$$