

ASSIGNMENT 1  
NON-LINEAR REGRESSION ANALYSIS

1. Suppose  $\{x_1, \dots, x_n\}$  is a set of  $n$  real numbers. Find  $a$  such that  $\sum_{i=1}^n (x_i - a)^2$  is minimized.
2. Suppose  $\{x_1, \dots, x_n\}$  is a set of  $n$  real numbers. Find  $a$  such that  $\sum_{i=1}^n |x_i - a|$  is minimized.
3. Suppose  $\{(x_1, y_1), \dots, (x_n, y_n)\}$  is a set of  $n$  pairs. Find  $a$  such that  $\sum_{i=1}^n |y_i - ax_i|$  is minimized.
4. Consider the following simple linear regression model:

$$y_i = a^0 + b^0 x_i + \epsilon_i; \quad i = 1, \dots, n.$$

Here  $\epsilon_i$ 's are i.i.d. normal random variables with mean 0 and variance 1. (a) Find  $\hat{a}$  and  $\hat{b}$ , the least squares estimators of  $a^0$  and  $b^0$ . (b) As  $n \rightarrow \infty$ , where these quantities  $\hat{a}$  and  $\hat{b}$  will converge? Under what conditions on  $x_i$ 's  $\hat{a}$  and  $\hat{b}$  will converge to  $a^0$  and  $b^0$  respectively.

5. Consider the same problem as in Problem 4. But it is known that  $0 \leq a \leq 1$  and  $0 \leq b \leq 1$ . Answer (a) and (b).
6. Consider the following simple linear regression model

$$y_i = b^0 x_i + \epsilon_i; \quad i = 1, \dots, n.$$

Suppose  $\epsilon_i$ 's are i.i.d. normal random variables with mean 0 and variance 1. Find the least absolute deviation estimator of  $b^0$ . **Find a sufficient condition so that it is consistent, i.e. as  $n \rightarrow \infty$ ,  $\hat{b}$  converges to  $b^0$ .**

7. Consider the following simple linear regression model:

$$y_i = a^0 + b^0 x_i + \epsilon_i; \quad i = 1, \dots, n.$$

Here  $\epsilon_i$ 's are i.i.d. random variables with mean 0 and variance 1, it is known that  $0 \leq a \leq 1$  and  $0 \leq b \leq 1$ . Let us assume you have enough computer power in your disposal. Can you suggest a method to compute the least absolute deviation estimators of  $a^0$  and  $b^0$ .

8. Consider the following simple linear regression model:

$$y_i = a^0 + b^0 x_i + \epsilon_i; \quad i = 1, \dots, n.$$

Here  $\epsilon_i$ 's are i.i.d. random variables with mean 0 and variance 1. It is known that  $a^0 = 2b^0$ , find the least absolute deviation estimator of  $a$ .

9. Suppose we have the following observations:

$$y_i = \alpha + \beta i + \epsilon_i; \quad i = 1, \dots, 50.$$

- (a) Suppose  $E(\epsilon_i) = 0$  and  $\text{Var}(\epsilon_i) = i^2$ , find the least squares estimators of  $\alpha$  and  $\beta$ .
- (b) Under the same conditions as in (a), find the generalized least squares estimators of  $\alpha$  and  $\beta$ . Will you prefer the least squares estimators or the generalized least squares estimators and why?
- (c) Suppose you know that  $\alpha + \beta = 2$ , how do you find the least squares estimators of  $\alpha$  and  $\beta$  under the assumptions that  $E(\epsilon_i) = 0$  and  $\text{Var}(\epsilon_i) = 1$ .

10. Consider the following linear model (here  $\omega$  is a known constant  $\in (0, \pi)$ )

$$y_t = A \cos(\omega t) + B \sin(\omega t) + \epsilon_t; \quad t = 1, \dots, N$$

- (a) Suppose,  $E(\epsilon_t) = 0$  and  $\text{Var}(\epsilon_t) = \sigma^2$ , find the least squares estimators of  $A$  and  $B$ .
- (b) Find an estimator of  $\sigma^2$ .
- (c) Under the assumptions that for large  $N$ ,  $\frac{1}{N} \sum_{t=1}^N \cos^2(\omega t) \approx \frac{1}{2}$ ,  $\frac{1}{N} \sum_{t=1}^N \sin^2(\omega t) \approx \frac{1}{2}$ ,  $\frac{1}{N} \sum_{t=1}^N \sin(\omega t) \cos(\omega t) \approx 0$ , find the approximate values of the least squares estimators of  $A$  and  $B$  for large  $N$ .

11. Consider the following linear model

$$\mathbf{Y} = \mathbf{X}\boldsymbol{\beta} + \boldsymbol{\epsilon}.$$

Here  $\mathbf{Y}$  is a  $2n \times 1$  vector,  $\mathbf{X}$  is a  $2n \times 3$  matrix,  $\boldsymbol{\beta}$  is a  $3 \times 1$  unknown parameter vector,  $\boldsymbol{\epsilon}^T = (\epsilon_1, \dots, \epsilon_{2n})$  and  $\epsilon_i$ 's are *i.i.d.*  $N(\mathbf{0}, \sigma^2)$ . The matrix  $\mathbf{X}$  is as follows;

$$\mathbf{X}^T = \begin{bmatrix} 1 & 1 & 1 & 1 & \dots & 1 & 1 & 1 \\ 2 & -2 & 2 & -2 & \dots & 2 & -2 & 2 & -2 \\ 3 & 3 & -3 & -3 & \dots & 3 & 3 & -3 & -3 \end{bmatrix}$$

- (a) Find the least squares estimates of  $\boldsymbol{\beta}$  and  $\sigma^2$ .
- (b) Find the distributions of the least squares estimates of  $\boldsymbol{\beta}$  and  $\sigma^2$ .
- (c) Construct a  $100(1-\alpha)\%$  confidence set of  $\boldsymbol{\beta}$ .
- (d) Show that the least squares estimate of  $\boldsymbol{\beta}$  is a consistent estimate of  $\boldsymbol{\beta}$ .
- (e) Find the asymptotic distribution of the least squares estimate of  $\boldsymbol{\beta}$ .

12. Consider the following regression model:

$$y_t = f_t(a_0, a_1, b_0, b_1) + \epsilon_t, \quad t = 1, 2, \dots, n,$$

here  $\epsilon_t$ 's are independent and identically distributed random variables with mean zero and variance 1 and

$$f_t(a_0, a_1, b_0, b_1) = \begin{cases} a_0 + a_1 t & \text{if } t < t_0 \\ b_0 + b_1 t & \text{if } t \geq t_0 \end{cases}$$

Here  $1 < t_0 < n$ , is an integer

- (a) If  $t_0$  is known find the least squares estimators of  $a_0$ ,  $a_1$ ,  $b_0$  and  $b_1$ .
- (b) If  $t_0$  is known, and it is known that  $a_0 + a_1 t_0 = b_0 + b_1 t_0$ , find the least squares estimators of  $a_0$ ,  $a_1$ ,  $b_0$  and  $b_1$ .
- (c) If  $t_0$  is unknown find the least squares estimators of  $a_0$ ,  $a_1$ ,  $b_0$ ,  $b_1$  and  $t_0$ .
- (d) If  $t_0$  is unknown, but it is known that  $a_0 + a_1 t_0 = b_0 + b_1 t_0$ , find the least squares estimators of  $a_0$ ,  $a_1$ ,  $b_0$ ,  $b_1$  and  $t_0$ .
- (e) I want to construct 95% confidence interval of  $a_0/b_0$  how will you do it?

ASSIGNMENT 2  
NON-LINEAR REGRESSION ANALYSIS

1. Consider the following non-linear regression model;

$$y(t) = \frac{\alpha_1 + \alpha_2 t}{\beta_1 + \beta_2 t} + e(t); \quad t = 1, \dots, n.$$

Here  $e(t)$ 's are independent identically distributed normal random variables with mean zero and finite variance  $\sigma^2$ .

- (a) If it is known that  $\beta_1 = 1$  and  $\beta_2 = 2$ , find the maximum likelihood estimators of  $\alpha$  and  $\alpha_2$
- (b) Provide an algorithm to find the maximum likelihood estimators of  $\alpha_1$ ,  $\alpha_2$ ,  $\beta_1$ , and  $\beta_2$  if all are unknown.
- (c) Find the maximum likelihood estimators of  $\sigma^2$ .
- (d) Do you think the maximum likelihood estimators will be consistent estimators of the unknown parameters or not?

2. Consider the following non-linear regression model:

$$y_t = \frac{1}{\theta_0 + \theta_1 t} + \epsilon_t; \quad t = 1, \dots, n.$$

Here  $\theta_0$  and  $\theta_1$  are unknown parameters and  $\epsilon_t$ 's are *i.i.d.* random variables with mean zero and variance  $\sigma^2$ .

- (a) Provide the Newton-Raphson algorithm to compute the least squares estimate of  $\theta_0$  and  $\theta_1$ .
- (b) Suppose  $\theta_0 = 1$  (known) and we want to estimate  $\theta_1$  only. We have only two observations 0.51 and 0.31 at  $t = 1$  and at  $t = 2$  respectively. We are using the Newton-Raphson algorithm to compute the least squares estimate of  $\theta_1$ . If at the  $i$ -th iterate the estimated value is 1.0, what will be the value at the  $(i+1)$ -th iterate. Explain it graphically also.
- (c) If both  $\theta_0$  and  $\theta_1$  are unknown, are the least squares estimates consistent? Justify your answer.
- (d) If  $\theta_1 = 0$ , find a consistent estimate of  $\theta_0$  (not necessarily least squares estimate) and show that it is consistent.
- (e) If  $\theta_0 = 0$ , find a consistent estimate of  $\theta_1$  (not necessarily least squares estimate) and show that it is consistent.

3. Consider the following non-linear regression model;

$$y_t = \cos(\omega_0 t) + \sin(\omega_0 t) + e_t; \quad t = 1, \dots, N$$

Here  $e_t$ 's are i.i.d. normal random variables with mean zero and variance 1. Find the least squares estimator of  $\omega_0$  using Newton-Raphson and also Gauss-Newton method.

4. Consider the following linear model

$$\mathbf{Y} = \mathbf{X}\boldsymbol{\beta} + \boldsymbol{\epsilon}.$$

Here  $\mathbf{Y} = (1, \dots, 1)$  is a  $2n \times 1$  vector,  $\mathbf{X}$  is a  $2n \times 2$  matrix,  $\boldsymbol{\beta} = (\beta_1, \beta_2)$  is a  $2 \times 1$  unknown parameter vector,  $\boldsymbol{\epsilon}^T = (\epsilon_1, \dots, \epsilon_{2n})$  and  $\epsilon_i$ 's are i.i.d.  $N(0, \sigma^2)$  random variables. The matrix  $\mathbf{X}$  is as follows;

$$\mathbf{X}^T = \begin{bmatrix} 1 & 1 & \dots & 1 & 1 \\ 2 & 2^2 & \dots & 2^{2n-1} & 2^{2n} \end{bmatrix}$$

- (a) Find explicitly the maximum likelihood estimates of  $\boldsymbol{\beta}$  and  $\sigma^2$ .
- (b) Find the distributions of the maximum likelihood estimates of  $\boldsymbol{\beta}$  and  $\sigma^2$ .
- (c) Construct a 95% confidence set of  $\boldsymbol{\beta}$ .
- (d) Show that the maximum likelihood estimate of  $\boldsymbol{\beta}$  is a consistent estimate of  $\boldsymbol{\beta}$ .
- (e) Find the asymptotic distribution of the maximum likelihood estimate of  $\boldsymbol{\beta}$ .
- (f) If we know  $\beta_1 = \beta_2 = \beta$ , find the maximum likelihood estimate of  $\beta$ .
- (g) Will it be consistent?

5. Suppose we have the following observations:

$$y_i = \alpha + \beta i + \epsilon_i; \quad i = 1, \dots, 50.$$

- (a) Suppose,  $E(\epsilon_i) = 5$  and  $\text{Var}(\epsilon_i) = 1$ , find the least squares estimators of  $\alpha$  and  $\beta$ .
- (b) Suppose  $E(\epsilon_i) = 0$  and  $\text{Var}(\epsilon_i) = i^2$ , find the least squares estimators of  $\alpha$  and  $\beta$ .
- (c) Under the same conditions as in (b), find the generalized least squares estimators of  $\alpha$  and  $\beta$ . Will you prefer the least squares estimators or the **generalized least squares estimators** and why?
- (d) Suppose you know that  $\alpha + \beta = 2$ , how do you find the least squares estimators of  $\alpha$  and  $\beta$  under the assumptions that  $E(\epsilon_i) = 0$  and  $\text{Var}(\epsilon_i) = 1$ .

6. Consider the following linear model (here  $\omega$  is a known constant  $\in (0, \pi)$ )

$$y_t = A \cos(\omega t) + B \sin(\omega t) + \epsilon_t; \quad t = 1, \dots, N$$

- (a) Suppose,  $E(\epsilon_t) = 0$  and  $\text{Var}(\epsilon_t) = \sigma^2$ , find the least squares estimators of  $A$  and  $B$ .
- (b) Find an estimator of  $\sigma^2$ .
- (c) Under the assumptions that for large  $N$ ,  $\frac{1}{N} \sum_{t=1}^N \cos^2(\omega t) \approx \frac{1}{2}$ ,  $\frac{1}{N} \sum_{t=1}^N \sin^2(\omega t) \approx \frac{1}{2}$ ,  $\frac{1}{N} \sum_{t=1}^N \sin(\omega t) \cos(\omega t) \approx 0$ , find the approximate values of the least squares estimators of  $A$  and  $B$  for large  $N$ .

7. Consider the following non-linear regression model;

$$y(x_t) = e^{Ax_t} + \epsilon_t; \quad t = 1, \dots, N$$

Explain how do you find the least squares estimators of A under the assumptions that  $E(\epsilon_t) = 0$  and  $\text{Var}(\epsilon_t) = 1$ .

ASSIGNMENT 3  
NON-LINEAR REGRESSION ANALYSIS

1. Consider the following non-linear regression model;

$$y(x_t) = e^{Ax_t} + \epsilon_t; \quad t = 1, \dots, N$$

Explain how do you find the least squares estimators of  $A$  under the assumptions that  $E(\epsilon_t) = 0$  and  $\text{Var}(\epsilon_t) = 1$ .

2. Consider the following non-linear regression model:

$$y(t) = \frac{2}{1 + \theta x_t} + e(t); \quad t = 1, \dots, N.$$

Here  $e(t)$ 's are i.i.d. random variables with the following probability density function

$$f(x) = c|x|^2 e^{-|x|^3}; \quad -\infty < x < \infty, c > 0.$$

- (a) Find  $c$ , so that  $f(x)$  becomes a proper probability density function.
  - (b) Based on the observed sample  $\{y(1), \dots, y(n)\}$ , find the likelihood function of  $\theta$ .
  - (c) Provide an explicit method to compute the maximum likelihood estimator of  $\theta$ , starting from an initial guess value of  $\theta$ , say  $\theta^{(0)}$ .
  - (d) Construct 95% confidence interval of  $e^\theta$ .
3. Consider the following general non-linear regression model

$$y_t = Af_t(\theta) + \epsilon_t; \quad t = 1, \dots, n.$$

Here  $A$  and  $\theta$  are scalars, and  $f_t(\theta)$  is a twice differentiable function for all  $t$ . The error random variables  $\epsilon_t$  are *i.i.d.* with the probability density function  $g(x)$ , where

$$g(x) = \frac{\lambda}{2} e^{-\lambda|x|}, \quad -\infty < x < \infty.$$

- (a) If  $\theta$  is known, find a method to compute the maximum likelihood estimate of  $A$ .
  - (b) If both are unknown find a method to compute the maximum likelihood estimates of  $A$  and  $\theta$ .
  - (c) In case of (b), how to construct 95% confidence intervals for  $A$  and  $\theta$ ?
4. Consider the following non-linear regression model:

$$y(t) = \frac{\alpha t}{1 + \beta t} + \epsilon_t; \quad t = 1, \dots, N.$$

Here  $\epsilon_t$ 's are normally distributed with mean 0 and variance  $t^2$ .

- (a) Find the least squares estimator of  $\alpha$  when  $\beta$  is known.
- (b) Show that when both  $\alpha$  and  $\beta$  are unknown the least squares estimators of  $\alpha$  and  $\beta$  can be obtained by solving a one dimensional optimization problem.
- (c) Suggest an suitable algorithm to compute the least squares estimators of  $\alpha$  and  $\beta$ . on  $\beta$ .

ASSIGNMENT 4  
NON-LINEAR REGRESSION ANALYSIS

In all the following questions we are considering the non-linear/ linear regression model of the form:

$$y_t = f_t(\theta^0) + \epsilon_t; \quad t = 1, \dots, n.$$

Here  $f_t(\theta)$  is continuous in  $\theta \in \Theta \subset R^p$ ,  $\theta^0$  is an interior point of  $\Theta$ ,  $\epsilon_t$ 's are i.i.d. random variables with mean 0 and finite variance. Further

$$D_n(\theta_1, \theta_2) = \frac{1}{n} \sum_{t=1}^n (f_t(\theta_1) - f_t(\theta_2))^2 \xrightarrow{\text{uniformly}} D(\theta_1, \theta_2) = 0 \text{ iff } \theta_1 = \theta_2.$$

1. Consider the following non-linear regression model;

$$y(x_t) = e^{A^0 t} + \epsilon_t; \quad t = 1, \dots, N$$

Here  $A^0 \in (-1, 0)$ ,  $\epsilon_t$ 's are i.i.d. random variables as above. Does this model satisfy the above conditions?

2. Consider the following non-linear regression model:

$$y(t) = \frac{2}{1 + \theta^0 t} + e(t); \quad t = 1, \dots, N.$$

Here  $e(t)$ 's are i.i.d. random variables with the following probability density function

$$f(x) = c|x|^2 e^{-|x|^3}; \quad -\infty < x < \infty, c > 0.$$

Here  $\theta^0 \in (0, 100)$ ,  $\epsilon_t$ 's are i.i.d. random variables as above. Does this model satisfy the above conditions?

3. Consider the following general non-linear regression model

$$y_t = A^0 \cos(\theta^0 t) + B^0 \sin(\theta^0 t) + \epsilon_t; \quad t = 1, \dots, n.$$

Here  $A^0, B^0 \in (-10, 10)$  and  $\theta^0 \in (0, \pi)$ .  $\epsilon_t$ 's are i.i.d. random variables as above. Does this model satisfy the above conditions?

4. Consider the following linear regression model:

$$y(t) = \alpha^0 + \beta^0 t + \epsilon_t; \quad t = 1, \dots, N.$$

Here  $\alpha^0, \beta^0 \in (-10, 10)$ ,  $\epsilon_t$ 's are i.i.d. normally distributed with mean 0 and variance 1. Does this model satisfy the above conditions? Is it possible to obtain consistent estimators of  $\alpha^0$  and  $\beta^0$  in this case?