- 1. Suppose we have covariate data x_1, \ldots, x_n and response data y_1, \ldots, y_n .
 - (i) (3) Write down the simple linear statistical model for y_i , making sure to define each and every term you specify, including all assumptions. Use b_0 and b_1 as notation for the regression coefficients.

$$Y_i = b_0 + b_1 x_i + \varepsilon_i$$

 $\varepsilon_i \text{ ind } N(o_1 o^2)$

(ii) (3) Write out the design matrix X and the response vector y for the simple linear model.

$$X = \begin{bmatrix} 1 & x_1 \\ 1 & x_2 \\ \vdots & \vdots \\ 1 & x_n \end{bmatrix} \qquad y = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix}$$

(iii) (4) Calculate
$$X^TX$$
 and verify that $(X^TX)^{-1} = \frac{1}{-n(\overline{x})^2 + \sum_{i=1}^n x_i^2} \begin{bmatrix} \frac{1}{n} \sum_{i=1}^n x_i^2 & -\overline{x} \\ -\overline{x} & 1 \end{bmatrix}$ (hint: multiply)

$$X^{T}X = \begin{bmatrix} 1 & 1 & \cdots & 1 \\ X_1 & X_2 & \cdots & X_n \end{bmatrix} \begin{bmatrix} 1 & X_1 \\ 1 & X_2 \\ \vdots & \vdots \\ 1 & X_n \end{bmatrix} = \begin{bmatrix} n & n\bar{x} \\ n\bar{x} & \Sigma X_i^2 \end{bmatrix}$$

$$(X^{T}X)(X^{T}X)^{-1} = \frac{1}{-n(\bar{x})^{2} + \Sigma x_{i}^{2}} \begin{bmatrix} n & n\bar{x} \\ n\bar{x} & \Sigma x_{i}^{2} \end{bmatrix} \begin{bmatrix} h \Sigma x_{i}^{2} - \bar{x} \\ -\bar{x} \end{bmatrix}$$

$$= \frac{1}{-n(\bar{x})^{2} + \Sigma x_{i}^{2}} \begin{bmatrix} \Sigma x_{i}^{2} - n(\bar{x})^{2} & -n\bar{x} + n\bar{x} \\ \bar{x} & \Sigma x_{i}^{2} - \bar{x} \Sigma x_{i}^{2} & -n(\bar{x})^{2} + \Sigma x_{i}^{2} \end{bmatrix}$$

$$= \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

- 2. Consider the linear model $\mathbf{Y} = X\mathbf{b} + \boldsymbol{\varepsilon}, \ \varepsilon_i \overset{ind}{\sim} N(0, \sigma^2)$
 - (i) (2) Derive $E(\mathbf{Y})$.

$$E(Y) = E(Xb + E) = Xb + E(E) = Xb$$

(ii) (3) Derive $E(\hat{B}) = E((X^T X)^{-1} X^T Y)$.

$$E(\hat{B}) = E((X^T X)^T X^T Y) = (X^T X)^{-1} X^T E(Y)$$
$$= (X^T X)^{-1} X^T X b = b$$

(iii) (2) Given that $Cov(\varepsilon) = \sigma^2 I$, derive $Cov(\mathbf{Y})$.

$$Cov(Y) = Cov(Xb+E) = Cov(E) = 0^2 I$$

(iv) (3) Derive $Cov(\widehat{\boldsymbol{B}})$.

$$Cov(\hat{B}) = Cov((x^{T}X)^{-1}X^{T}Y) = (X^{T}X)^{T}X^{T}(ov(Y)X(X^{T}X)^{T})$$

$$= (X^{T}X)^{-1}X^{T}o^{2}IX(X^{T}X)^{-1}$$

$$= o^{2}(X^{T}X)^{-1}X^{T}X(X^{T}X)^{-1}$$

$$= o^{2}(X^{T}X)^{-1}$$

- 3. Consider the same model as above, but with the following design matrix: $X = \begin{bmatrix} 1 & 1 \\ 1 & 1 \\ 1 & 0 \end{bmatrix}$
 - (i) (3) Write down the expected values of the four responses and describe the pattern in words.

$$E(Y_1) = bo + b_1$$

$$E(Y_2) = bo + b_1$$

$$E(Y_3) = bo$$

$$E(Y_4) = bo$$

(ii) (2) Calculate (X^TX) and $(X^TX)^{-1}$. (check your work!)

$$X^{T}X = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 1 & 0 & 0 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 1 & 0 \\ 1 & 0 \end{bmatrix} = \begin{bmatrix} 4 & 2 \\ 2 & 2 \end{bmatrix} \begin{bmatrix} 4^{2} & 1 & \frac{1}{2} & \frac{1}{2} \\ 2 & 2 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 1 \end{bmatrix}$$

$$(X^{T}X)^{T} = \frac{1}{8-4} \begin{bmatrix} 2 & -2 \\ -2 & 4 \end{bmatrix} = \begin{bmatrix} \frac{1}{2} & -\frac{1}{2} \\ -\frac{1}{2} & 1 \end{bmatrix}$$

(iii) (2) Calculate the projection matrix for X.

(iv) (3) Given response data y_1, y_2, y_3, y_4 , calculate the fitted values, and explain how they are related to your answer in part (i).

$$\left(\frac{1}{2} \frac{1}$$

(v) (+1) Find a matrix U with orthonormal columns, such that UU^T is the projection matrix for X.