

# Project 1 FYS-STK3155

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## Abstract

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We have defined  $\mathbf{y}$  as a function of a matrix multiplication  $X\boldsymbol{\beta}$  plus an error vector  $\boldsymbol{\epsilon}$ . This means that each element of the vector  $\mathbf{y}$  can be expressed as follows:

$$y_i = \sum_j x_{ij}\beta_j + \epsilon_i$$

If we take the expectation value of this expression we get the following:

$$E[y_i] = E\left[\sum_j x_{ij}\beta_j + \epsilon_i\right]$$
$$E[y_i] = E\left[\sum_j x_{ij}\beta_j\right] + E[\epsilon_i]$$

However, the elements of  $X$  are not stochastic, and neither are the elements of  $\boldsymbol{\beta}$  the first expectation value is simply the sum itself. Furthermore,  $\epsilon$  is explicitly defined as a normal distribution  $N(0, \sigma^2)$ , and will by definition have the expectation value 0. Therefore, we end up with the final expression:

$$E[y_i] = \sum_j x_{ij}\beta_j = \mathbf{X}_{i,*}\boldsymbol{\beta}$$

We can use expectation values to calculate the variance as well:

$$\text{var}[y_i] = E[(y_i - E[y_i])^2]$$