

Exercise 5 (python code + text):

Consider a regression problem where both the independent and dependent quantities are scalars and are related via the following linear model

$$y = \theta_o \cdot x + \eta$$

where η follows the zero mean normal distribution with variance σ^2 and $\theta_o = 2$ (thus, the actual model is $y = 2 \cdot x + \eta$).

(a) Generate $d = 50$ data set as follows:

- Generate a set D_1 of $N = 30$ data pairs (y_i', x_i) , where $y' = 2 \cdot x$.
- Add zero mean and $\sigma^2 = 64$ variance Gaussian noise to the y_i' 's, resulting to y_i' s.

- The **observed** data pairs are (y_i, x_i) , $i = 1, \dots, 30$, which constitute the data set D_1 .

Repeat the above procedure $d=50$ times in order to generate 50 different data sets.

- Compute the LS linear **estimates** of θ_o based on D_1, D_2, \dots, D_d (thus, $\hat{\theta}_1, \hat{\theta}_2, \dots, \hat{\theta}_d$ numbers/estimates will result).
- Consider now the random variable $\hat{\theta}$ that models $\hat{\theta}_1, \hat{\theta}_2, \dots, \hat{\theta}_d$ (that is, $\hat{\theta}_1, \hat{\theta}_2, \dots, \hat{\theta}_d$ can be viewed as instances of the random variable $\hat{\theta}$)¹ and
 - compute the $MSE = E[(\hat{\theta} - \theta_o)^2]$ and
 - depict graphically the values $\hat{\theta}_1, \hat{\theta}_2, \dots, \hat{\theta}_d$ and comment on how they are spread around θ_o .

Hint: For (c) approximate MSE as $MSE = \frac{1}{d} \sum_{i=1}^d (\hat{\theta}_i - \theta_o)^2$.