

## Assignment #2

Our calculations were made in Python, the code is attached in a notebook file.

### Question 1 – Linear Models and Bootstrapping

a. *Linear regression model:*

The estimated value for the slope and intercept that we've found are as follows:

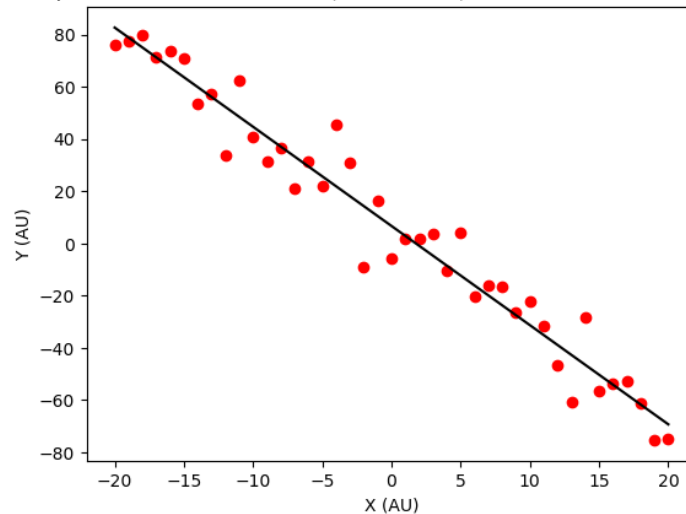
The min MSE for this beta vector is: 96.10310512036297

The slope and intercept value with the lowest MSE:  $a=-3.8$ ,  $b=6.7$

The best fit linear regression is:  $y=-3.8*x+6.7$

Figure 1:

Scatter plot of the measured data (vectors X, Y) and the best fit linear regression

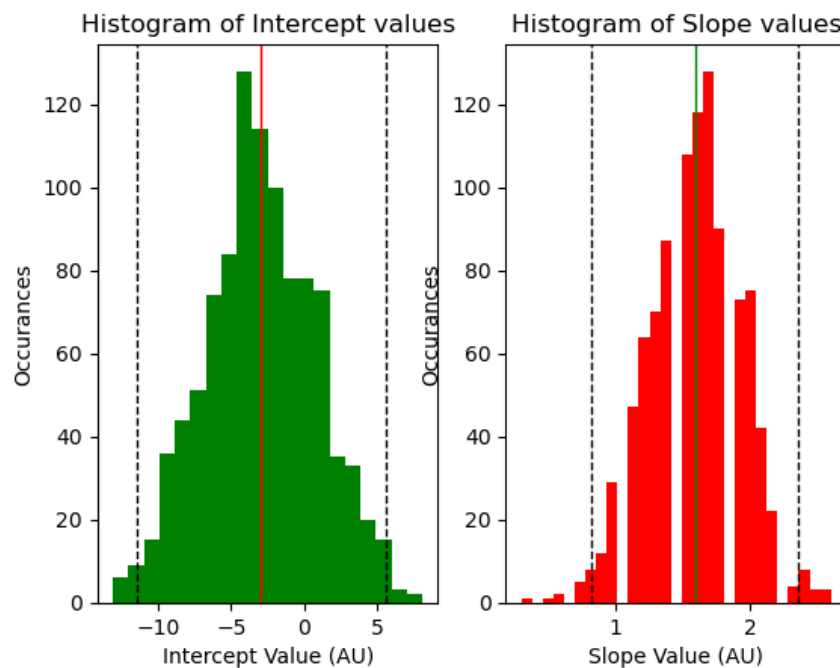


Estimating the STD of the noise distribution: In our case, the noise is defined as the difference between the predicted and measured values. So, we've calculated a distribution of those differences and its STD (here we're dealing with a sample and not a population, therefore the  $df=n-1$ ).

Estimated Noise STD: 9.92499935184865

b. *Bootstrapping:*

Figure 2: Histograms of the best slope and intercept estimates for 1000 generated samples



The average value for the estimated slopes is 1.58 , and -2.75 for the intercepts, as is denoted by the solid line in [Figure 2](#).

Answer for bonus: the dotted lines in [Figure 2](#) represent the 97.5% CI for each distribution.

The boundaries of the CI for the intercepts are: -11.417141795340003 5.55714179534001

The boundaries of the CI for the slopes are: 0.8238823920782311 2.3609176079217695

### Question 2 – GLM

The set of parameters  $\theta : [(17, 6, -19, 48), (1, 4, -7, 1), (3, 0.02, 0.1, -1)]$  the LL values are presented in a corresponding order.

- a. To find the best set of parameters  $\theta$ , we need to calculate the log likelihood of each set and chose the set with the highest LL value.

The calculated Log-Likelihood values are: [-5043703341.508967, 453.0694004659954, 411.3052242236605]

The theta vector with the max LL is: (1, 4, -7, 1)

- b. This regularization factor "punishes" high  $\theta$  values. After adding the regularization factor

The calculated new Log-Likelihood values are: [-5043704836.508967, 419.5694004659954, 406.3000242236605]

The theta vector with the max LL is: (1, 4, -7, 1)

as we can see, the LL values changed but the  $\theta$  set with the highest one remains the same – the answer didn't change.

- c. To set up the Poisson regression, we've used a *Log* link function (the default of the Poisson family in *statsmodels* GLM) and a noise distribution from the *Poisson-family*.

```

Generalized Linear Model Regression Results
=====
Dep. Variable:                y      No. Observations:          122
Model:                      GLM      Df Residuals:              118
Model Family:              Poisson   Df Model:                  3
Link Function:              Log      Scale:                    1.0000
Method:                    IRLS      Log-Likelihood:           -294.21
Date:                      Tue, 13 Dec 2022    Deviance:                 118.31
Time:                      18:44:51    Pearson chi2:             114.
No. Iterations:            4          Pseudo R-squ. (CS):       0.01275
Covariance Type:          nonrobust
=====
               coef      std err          z      P>|z|      [0.025      0.975]
-----
const         2.1681      0.259      8.370      0.000      1.660      2.676
x1            2.0390      3.443      0.592      0.554     -4.709      8.787
x2           -3.3974      3.920     -0.867      0.386     -11.080      4.285
x3            1.6671      1.732      0.963      0.336     -1.728      5.062
=====
The Theta vector is: [ 2.16805029  2.03900614 -3.39743477  1.66713681]

```

- d. The first value is the intercept. From the rest, as we're told that the X values are neurons:  $X_1$ ,  $X_2$ ,  $X_3$  influencing our neuron's Y FR. Therefore, we can infer that negative  $\theta$  values represent inhibitory inputs while positive ones represent excitatory inputs.
- $X_1$  is excitatory,  $X_2$  is inhibitory and  $X_3$  is excitatory as well. Furthermore, neuron  $X_2$  has the highest influence on Y's FR, as it has the highest absolute coefficient value.