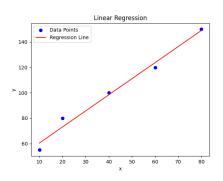
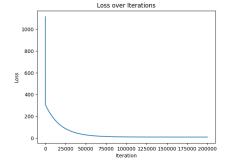
## Labwork 2: Linear Regression

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### 1 Implementation

- First, we implement functions to calculate the single loss value, 2 partial derivatives over  $w_0$  and  $w_1$ , and a function to calculate the loss of all data points.
- Read the csv file and extract the value of x and y as lists of floats
- Perform gradient descent to optimize the weights  $w_0$  and  $w_1$ . It iteratively updates  $w_0$  and  $w_1$  using the gradients (df0 and df1) and the learning rate lr
- After training, the optimized weights w0 and w1 are used to plot the regression line based on the learned weights and a plot shows the loss over iterations
- Here is the result when the gradient descent is run for 200,000 iterations with a learning rate of 0.0001 when the initial value for  $w_0$  and  $w_1$  is 1 and 1 respectively



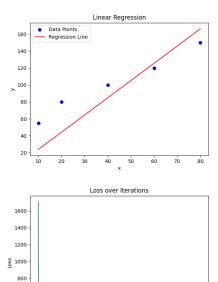


- After updating, the loss is around  $9.46\,$ 

# 2 The effect of different learning rates for convergence

### 2.1 Learning rate is too small

- When the learning rate is too small, it requires many updates before reaching the minimum point



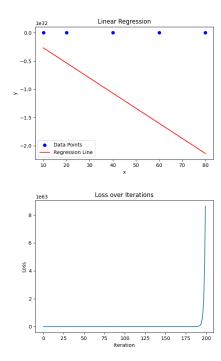
- After updating for  $200,\!000$  times, the loss is around 277.74

### 2.2 Learning rate is too large

600

- When the learning rate is too large, it could cause drastic updates, which lead to divergent behaviors

25000 50000 75000 100000 125000 150000 175000 200000



- The algorithm overshoot the minimum and lead to divergence