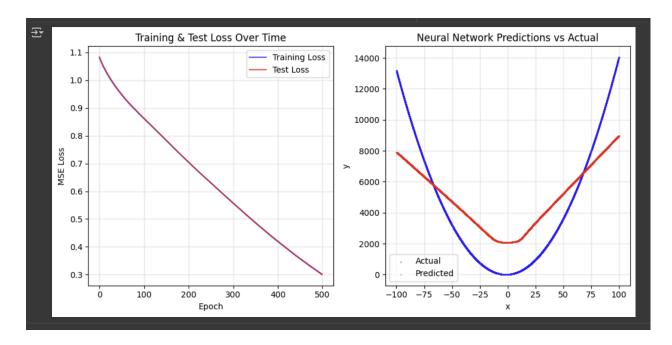
### **MACHINE LEARNING LAB**

## 5th Semester, Academic Year 2025

Date:19-09-2025

Name: Dhruv Thakur	SRN:PES2UG23CS175	Section
		С

# ReLu(baseline)



**₹** 

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PREDICTION RESULTS FOR x = 90.2

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Neural Network Prediction: 8,246.71 Ground Truth (formula): 11,478.02 Absolute Error: 3,231.30 Relative Error: 28.152%

<del>\_\_\_\_</del>

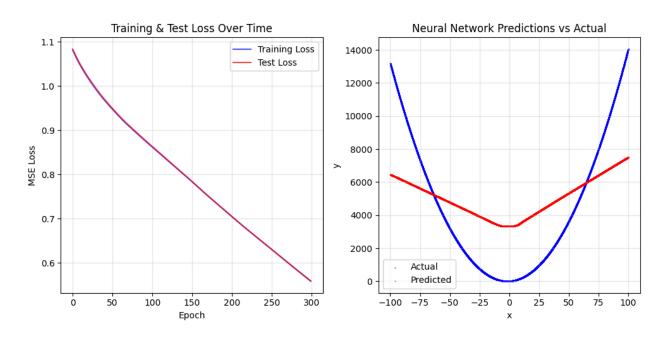
FINAL PERFORMANCE SUMMARY

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Final Training Loss: 0.301565 Final Test Loss: 0.299981 R<sup>2</sup> Score: 0.7013 Total Epochs Run: 500

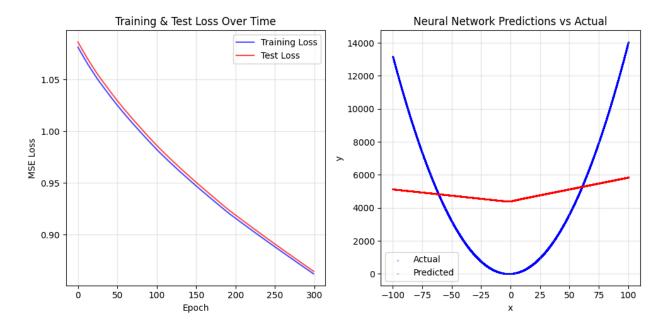
## ReLu 300 epochs

```
print("Training Neural Network with your specific configuration...")
weights, train_losses, test_losses = train_neural_network(
    X_train_scaled, Y_train_scaled, X_test_scaled, Y_test_scaled,
    epochs=300, patience=10
Training Neural Network with your specific configuration...
Starting training...
Architecture: 1 → 96 → 96 → 1
Learning Rate: 0.003
Max Epochs: 300, Early Stopping Patience: 10
Epoch 20: Train Loss = 1.018489, Test Loss = 1.021008
Epoch 40: Train Loss = 0.969659, Test Loss = 0.971934
Epoch 60: Train Loss = 0.929711, Test Loss = 0.931603
Epoch 80: Train Loss = 0.895185, Test Loss = 0.896887
Epoch 100: Train Loss = 0.863207, Test Loss = 0.864600
Epoch 120: Train Loss = 0.831969, Test Loss = 0.833054
Epoch 140: Train Loss = 0.800543, Test Loss = 0.801376
Epoch 160: Train Loss = 0.768616, Test Loss = 0.769140
Epoch 180: Train Loss = 0.736913, Test Loss = 0.737347
Epoch 200: Train Loss = 0.705889, Test Loss = 0.706129
Epoch 220: Train Loss = 0.675576, Test Loss = 0.675683
Epoch 240: Train Loss = 0.646297, Test Loss = 0.646247
Epoch 260: Train Loss = 0.617152, Test Loss = 0.616866
Epoch 280: Train Loss = 0.587738, Test Loss = 0.587296
Epoch 300: Train Loss = 0.558904, Test Loss = 0.558295
```



#### ReLu lr 0.001 300 epoch

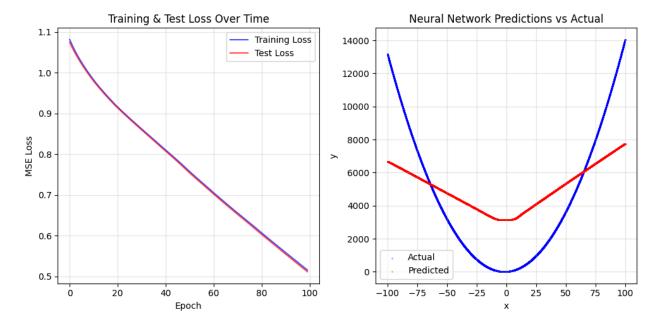
```
→ Training Neural Network with your specific configuration...
    Starting training...
    Architecture: 1 → 96 → 96 → 1
    Learning Rate: 0.001
    Max Epochs: 300, Early Stopping Patience: 10
    Epoch 20: Train Loss = 1.057201, Test Loss = 1.061987
    Epoch 40: Train Loss = 1.035831, Test Loss = 1.040399
    Epoch 60: Train Loss = 1.016770, Test Loss = 1.021118
    Epoch 80: Train Loss = 0.999482, Test Loss = 1.003591
    Epoch 100: Train Loss = 0.983173, Test Loss = 0.987056
    Epoch 120: Train Loss = 0.968329, Test Loss = 0.972043
    Epoch 140: Train Loss = 0.954498, Test Loss = 0.958025
    Epoch 160: Train Loss = 0.941286, Test Loss = 0.944640
    Epoch 180: Train Loss = 0.928504, Test Loss = 0.931674
    Epoch 200: Train Loss = 0.916393, Test Loss = 0.919465
    Epoch 220: Train Loss = 0.905145, Test Loss = 0.908074
    Epoch 240: Train Loss = 0.894138, Test Loss = 0.896934
    Epoch 260: Train Loss = 0.883318, Test Loss = 0.885988
    Epoch 280: Train Loss = 0.872661, Test Loss = 0.875216
    Epoch 300: Train Loss = 0.862119, Test Loss = 0.864564
```



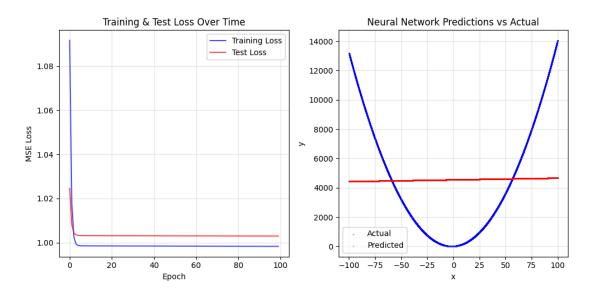
#### **ReLu 0.01 100 epochs**

```
Training Neural Network with your specific configuration...
Starting training...
Architecture: 1 → 96 → 96 → 1
Learning Rate: 0.01
Max Epochs: 100, Early Stopping Patience: 10

Epoch 20: Train Loss = 0.921234, Test Loss = 0.919085
Epoch 40: Train Loss = 0.815125, Test Loss = 0.812463
Epoch 60: Train Loss = 0.710431, Test Loss = 0.707059
Epoch 80: Train Loss = 0.611627, Test Loss = 0.607842
Epoch 100: Train Loss = 0.515813, Test Loss = 0.511717
```



#### Sigmoid Ir 0.01 epochs 100



**₹** 

PREDICTION RESULTS FOR x = 90.2

Neural Network Prediction: 4,659.55 Ground Truth (formula): 11,478.02 Absolute Error: 6,818.47 Relative Error: 59.405%

**₹** 

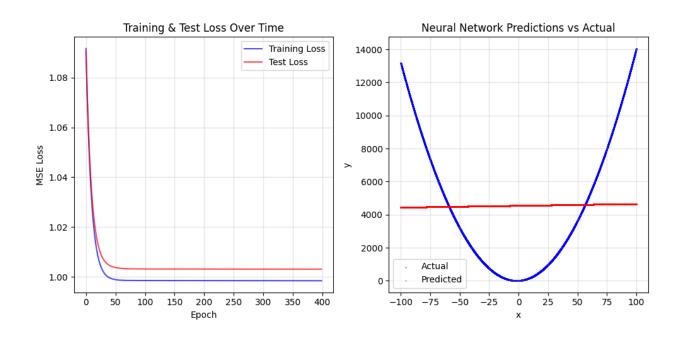
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FINAL PERFORMANCE SUMMARY

Final Training Loss: 0.998295
Final Test Loss: 1.002974
R<sup>2</sup> Score: 0.0013
Total Epochs Run: 100

#### Sigmoid Ir 0.001 epochs 400

```
Training Neural Network with your specific configuration...
Starting training...
Architecture: 1 → 96 → 96 → 1
Learning Rate: 0.001
Max Epochs: 400, Early Stopping Patience: 10
Epoch 20: Train Loss = 1.009672, Test Loss = 1.014539
Epoch 40: Train Loss = 0.999721, Test Loss = 1.004686
Epoch 60: Train Loss = 0.998653, Test Loss = 1.003419
Epoch 80: Train Loss = 0.998534, Test Loss = 1.003211
Epoch 100: Train Loss = 0.998517, Test Loss = 1.003162
Epoch 120: Train Loss = 0.998510, Test Loss = 1.003146
Epoch 140: Train Loss = 0.998505, Test Loss = 1.003139
Epoch 160: Train Loss = 0.998500, Test Loss = 1.003134
Epoch 180: Train Loss = 0.998494, Test Loss = 1.003129
Epoch 200: Train Loss = 0.998489, Test Loss = 1.003125
Epoch 220: Train Loss = 0.998484, Test Loss = 1.003121
Epoch 240: Train Loss = 0.998479, Test Loss = 1.003117
Epoch 260: Train Loss = 0.998474, Test Loss = 1.003113
Epoch 280: Train Loss = 0.998469, Test Loss = 1.003109
Epoch 300: Train Loss = 0.998464, Test Loss = 1.003105
Epoch 320: Train Loss = 0.998459, Test Loss = 1.003102
Epoch 340: Train Loss = 0.998454, Test Loss = 1.003098
Epoch 360: Train Loss = 0.998449, Test Loss = 1.003094
Epoch 380: Train Loss = 0.998444, Test Loss = 1.003090
Epoch 400: Train Loss = 0.998438, Test Loss = 1.003086
```





PREDICTION RESULTS FOR x = 90.2

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Neural Network Prediction: 4,649.68
Ground Truth (formula): 11,478.02
Absolute Error: 6,828.33
Relative Error: 59.491%



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FINAL PERFORMANCE SUMMARY

Final Training Loss: 0.998438
Final Test Loss: 1.003086
R<sup>2</sup> Score: 0.0012
Total Epochs Run: 400

## experiments results:

	A	В	С	D	E	F	G	н
1	Experiment	Learning Rate	No. of Epochs	Optimizer (if use	Activation Function	Final Training Lo	Final Test Loss	R <sup>2</sup> Score
2	ReLU (baseline)	0.003	500	Gradient Descen	ReLU	0.301565	0.299981	0.7013
3	ReLU (300 epocl	0.003	300	Gradient Descen	ReLU	0.558904	0.558295	0.4441
4	ReLU (Ir=0.001,	0.001	300	Gradient Descen	ReLU	0.862119	0.864564	0.1391
5	ReLU (Ir=0.01, 1	0.01	100	Gradient Descen	ReLU	0.515813	0.511717	0.4985
6	Sigmoid (Ir=0.01,	0.01	100	Gradient Descen	Sigmoid	0.998295	1.002974	0.0013
7	Sigmoid (Ir=0.00	0.001	400	Gradient Descen	Sigmoid	0.998438	1.003086	0.0012

#### **Key Insights**

Best Performing Model → ReLU (baseline, Ir=0.003, 500 epochs)

Final Training Loss: 0.3016

Final Test Loss: 0.3000

R<sup>2</sup> Score: 0.7013 (highest of all)

This shows the best fit and generalization.

Reason:

ReLU avoids saturation (unlike Sigmoid).

Adequate epochs (500) allowed the network to converge steadily.

Learning rate of 0.003 was neither too high nor too low, providing a good balance.

Worst Performing Models → Sigmoid activations

Sigmoid (Ir=0.01, 100 epochs):  $R^2 = 0.0013$ 

Sigmoid (Ir=0.001, 400 epochs):  $R^2 = 0.0012$ 

These essentially failed to learn.

Reason:

Sigmoid suffers from the vanishing gradient problem, especially with deeper networks (2 hidden layers here).

Gradients become tiny for large |z|, slowing down or stopping learning.

Even with more epochs (400), the loss stayed ~1.0, showing the model didn't improve.