

# Assignment 3








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Data 1

# weather\_izmir

- Data\_id = 42369
- The file contains the weather information of Izmir from 01/01/1994 to 31/12/1997. From given features, the goal is to predict the mean temperature.
- There are no nominal features and missing values , both target and features are all numeric , so I didn't transform the data
- [https://www.openml.org/search?type=data&sort=qualities.NumberOfNumericFeatures&status=active&qualities.NumberOfClasses=lte\\_1&qualities.NumberOfInstances=between\\_1000\\_10000&order=asc&qualities.NumberOfFeatures=between\\_10\\_100&id=42369](https://www.openml.org/search?type=data&sort=qualities.NumberOfNumericFeatures&status=active&qualities.NumberOfClasses=lte_1&qualities.NumberOfInstances=between_1000_10000&order=asc&qualities.NumberOfFeatures=between_10_100&id=42369)

# Data 1 information

	Mean_temperature (target)	numeric	489 distinct values 0 missing attributes
	Max_temperature	numeric	300 distinct values 0 missing attributes
	Min_temperature	numeric	284 distinct values 0 missing attributes
	Dewpoint	numeric	387 distinct values 0 missing attributes
	Precipitation	numeric	26 distinct values 0 missing attributes
	Sea_level_pressure	numeric	90 distinct values 0 missing attributes
	Standard_pressure	numeric	52 distinct values 0 missing attributes

# Model 1

## **Input Layer:**

The model has 9 input nodes, representing the dataset's 10 features, excluding the target variable.

## **Hidden Layer:**

Contains 1 Dense layer with 5 nodes.

Activation function: Sigmoid.

Parameters: 50 (calculated as 9 input features  $\times$  5 nodes + 5 biases).

## **Output Layer:**

Contains 1 Dense layer with 1 node.

Designed for regression tasks.

Parameters: 6 (calculated as 5 hidden nodes  $\times$  1 output node + 1 bias).

# Model 1 summary

Model: "sequential\_21"

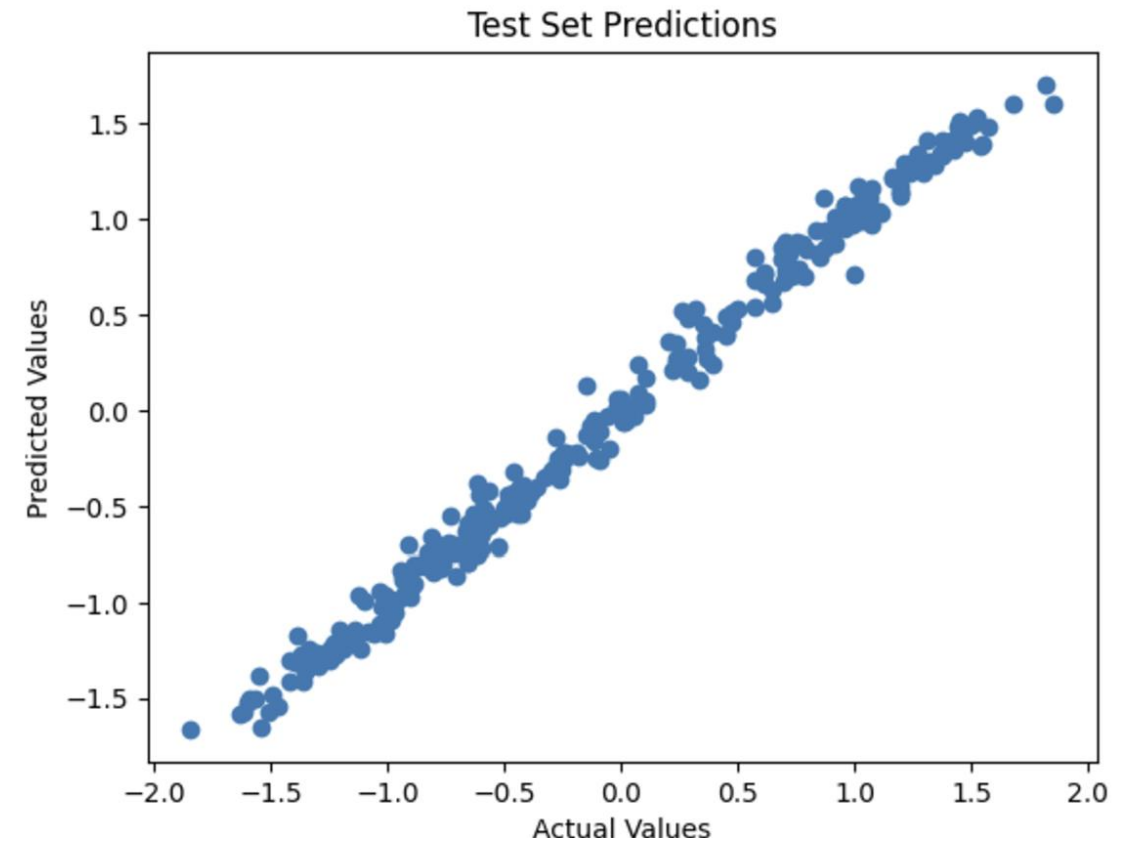
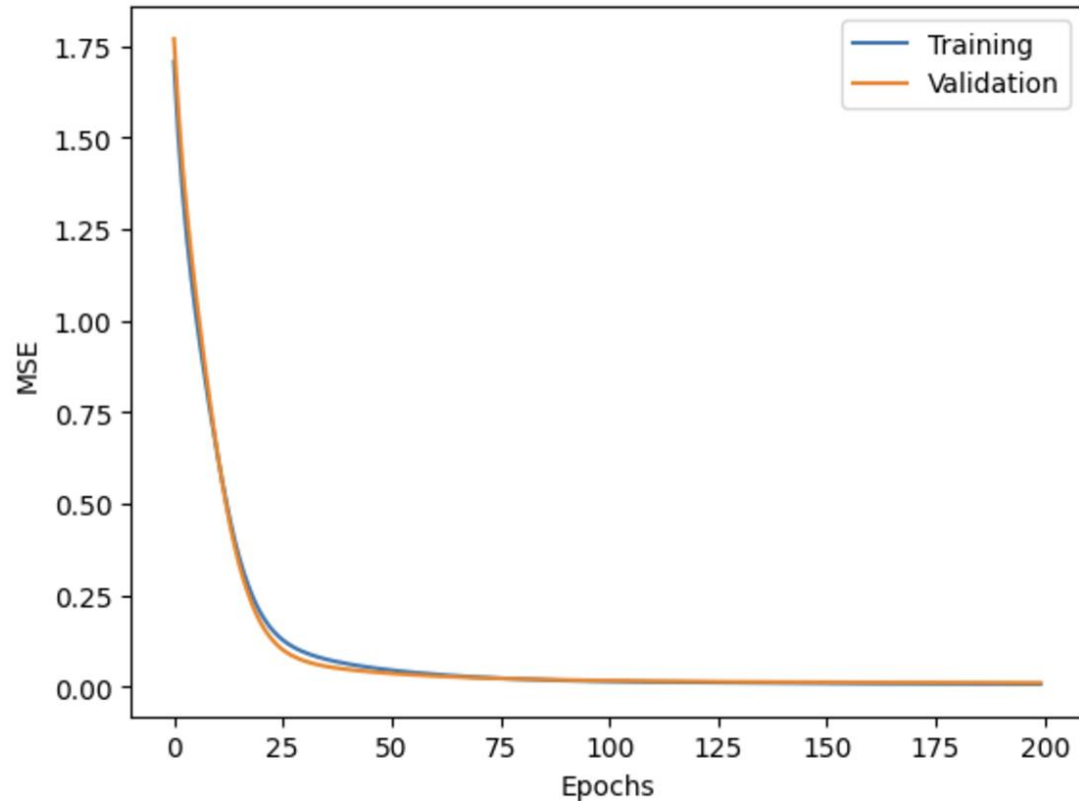
Layer (type)	Output Shape	Param #
dense_40 (Dense)	(None, 5)	50
dense_41 (Dense)	(None, 1)	6

Total params: 56 (224.00 B)

Trainable params: 56 (224.00 B)

Non-trainable params: 0 (0.00 B)

# Model 1 Result



# Model 2

## **Input Layer:**

The model has 9 input nodes, representing the dataset's 10 features, excluding the target variable.

## **Hidden Layer:**

Contains 1 Dense layer with 30 nodes.

Activation function: Sigmoid.

Parameters: 300 (calculated as 9 input features  $\times$  30 nodes + 30 biases).

## **Output Layer:**

Contains 1 Dense layer with 1 node.

Designed for regression tasks.

Parameters: 31 (calculated as 30 hidden nodes  $\times$  1 output node + 1 bias).



# Model 2 summary

Model: "sequential\_20"

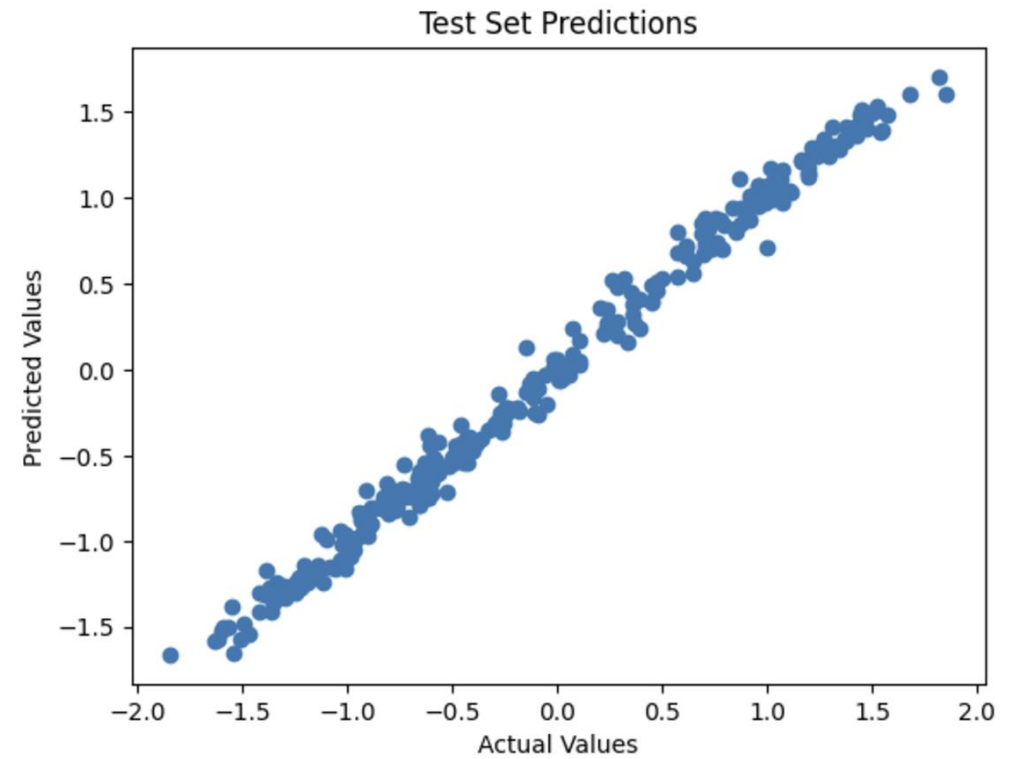
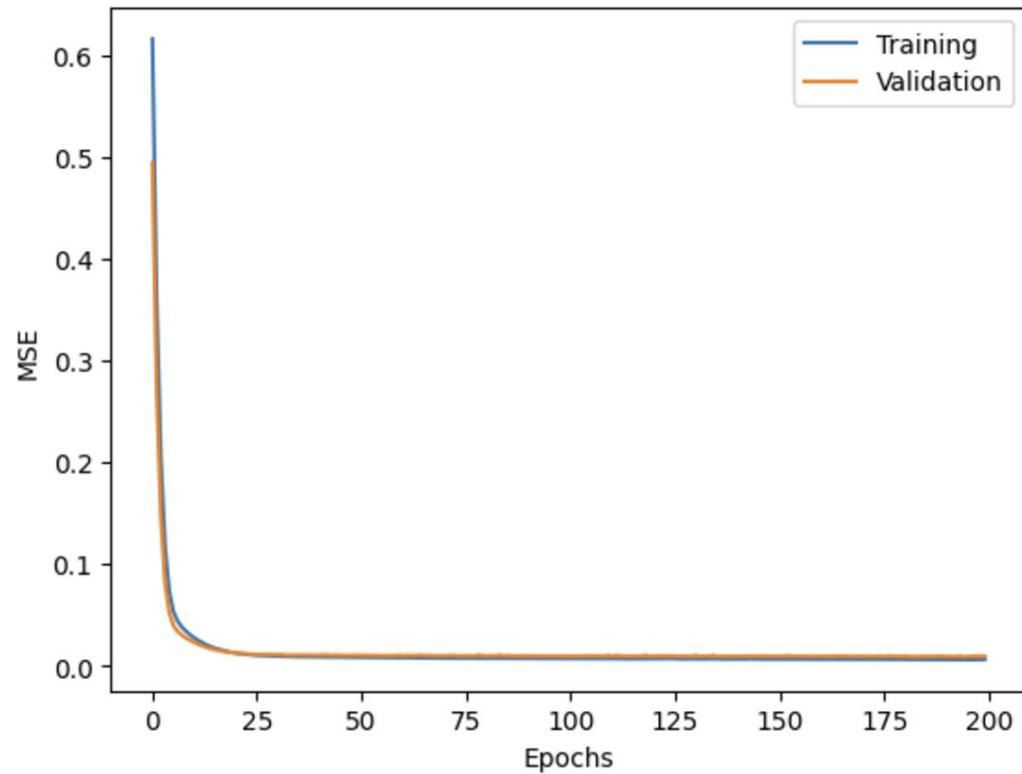
Layer (type)	Output Shape	Param #
dense_38 (Dense)	(None, 30)	300
dense_39 (Dense)	(None, 1)	31

Total params: 331 (1.29 KB)

Trainable params: 331 (1.29 KB)

Non-trainable params: 0 (0.00 B)

# Model 2 Result



# Model 3

## **Input Layer:**

The model has 9 input nodes, representing the dataset's 10 features, excluding the target variable.

## **Hidden Layer:**

Contains 1 Dense layer with 200 nodes.

Activation function: Sigmoid.

Parameters: 2,000 (calculated as 9 input features  $\times$  200 nodes + 200 biases).

## **Output Layer:**

Contains 1 Dense layer with 1 node.

Designed for regression tasks.

Parameters: 201 (calculated as 200 hidden nodes  $\times$  1 output node + 1 bias).

# Model 3 summary

Model: "sequential\_19"

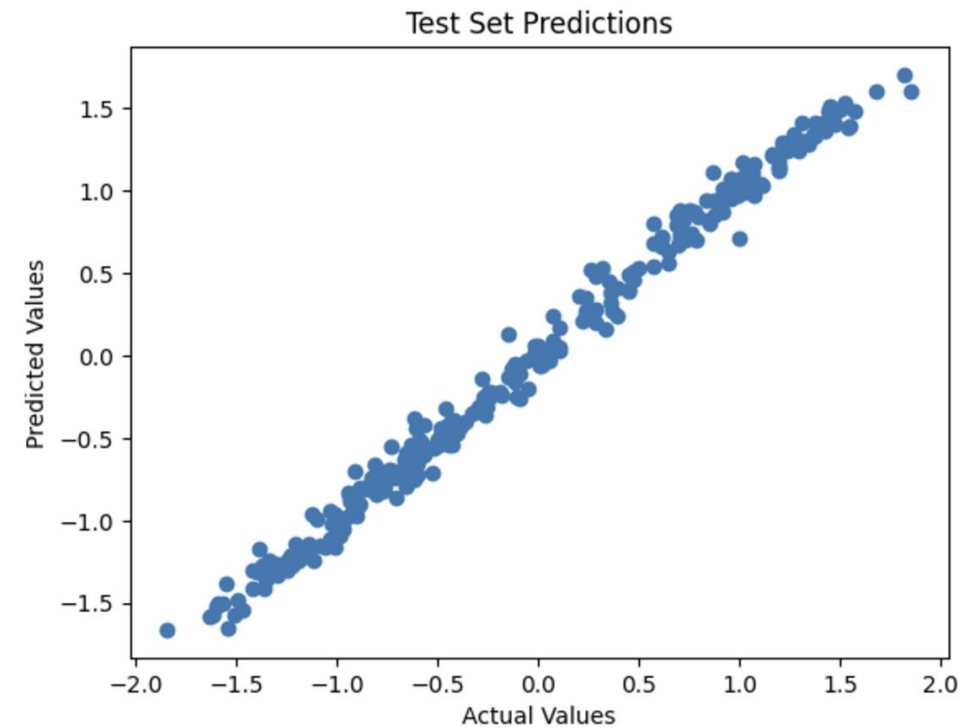
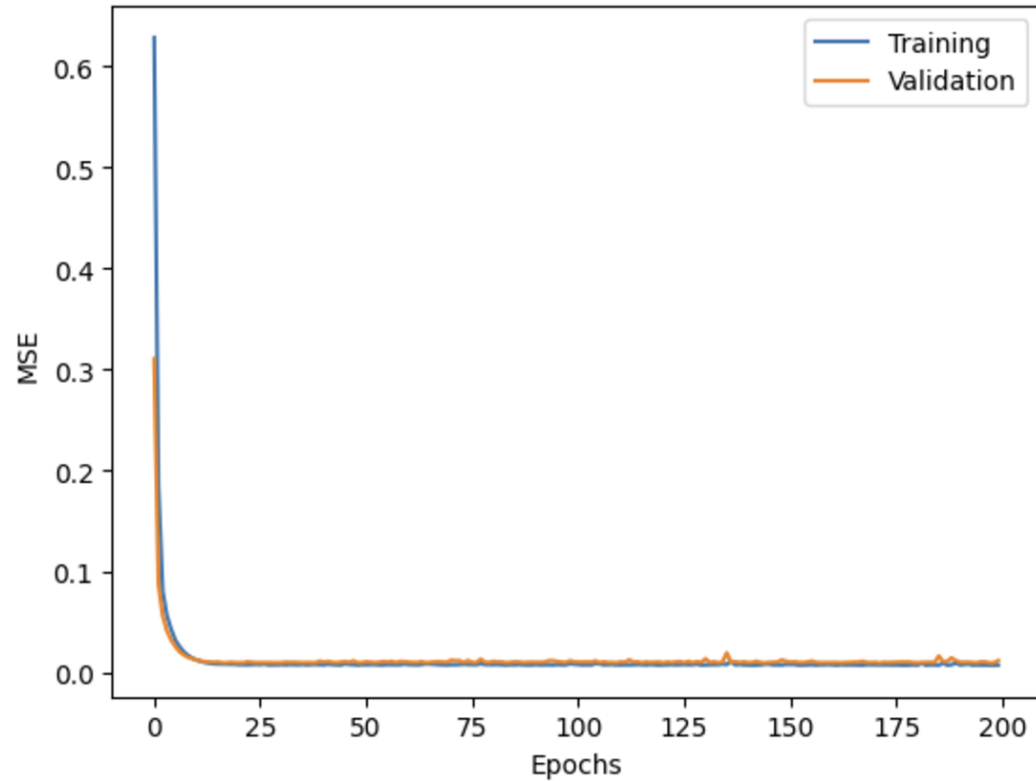
Layer (type)	Output Shape	Param #
dense_36 (Dense)	(None, 200)	2,000
dense_37 (Dense)	(None, 1)	201

Total params: 2,201 (8.60 KB)

Trainable params: 2,201 (8.60 KB)

Non-trainable params: 0 (0.00 B)

# Model 3 Result



## table of test errors

	Model	Test MSE
0	Model 1 (5 nodes)	0.007141
1	Model 2 (30 nodes)	0.006856
2	Model 3 (200 nodes)	0.007815

# Conclusion

- **Training and Validation Errors:**

All three models show a steady decline in both training and validation MSE across epochs.

The validation error closely follows the training error, indicating that the models are not overfitting.

- **Comparison of Test MSE:**

Model 1 (5 nodes): Test MSE = 0.007141

Model 2 (30 nodes): Test MSE = 0.006856

Model 3 (200 nodes): Test MSE = 0.007815

Model 2 achieves the lowest test error, suggesting that 30 nodes in the hidden layer offer the best balance of model complexity and generalization.

- **Model Complexity and Performance:**

Model 1, with only 5 nodes, may be underfitting the data slightly, as its test error is higher than Model 2.

Model 3, with 200 nodes, shows a slight increase in test error compared to Model 2, potentially due to overfitting, as the model is too complex for the dataset.

- **Optimal Model:**

Model 2 strikes the best balance between complexity and performance, achieving the lowest test MSE.

Adding too many nodes (Model 3) does not improve performance and may increase the risk of overfitting.

# Data 2










# wind

- Data\_id = 503
- wind daily average wind speeds for 1961-1978 at 12 synoptic meteorological stations in the Republic of Ireland (Haslett and raftery 1989).
- There are no nominal features and missing values , both target and features are all numeric , so I didn't transform the data
- [https://www.openml.org/search?type=data&sort=runs&status=active&qualities.NumberOfInstances=between\\_1000\\_10000&qualities.NumberOfFeatures=between\\_10\\_100&qualities.NumberOfClasses=lte\\_1&id=503](https://www.openml.org/search?type=data&sort=runs&status=active&qualities.NumberOfInstances=between_1000_10000&qualities.NumberOfFeatures=between_10_100&qualities.NumberOfClasses=lte_1&id=503)

# Data 2 information

15 Features

▼ Expand

Feature Name	Type	Distinct/Missing Values	Ontology
 MAL (target)	numeric	779 distinct values 0 missing attributes	
 year	numeric	18 distinct values 0 missing attributes	
 month	numeric	12 distinct values 0 missing attributes	
 day	numeric	31 distinct values 0 missing attributes	
 RPT	numeric	671 distinct values 0 missing attributes	
 VAL	numeric	607 distinct values 0 missing attributes	
 ROS	numeric	611 distinct values 0 missing attributes	

# Model 1

## **Input Layer:**

The model has 14 input nodes, representing the dataset's 15 features, excluding the target variable.

## **Hidden Layer:**

Contains 1 Dense layer with 5 nodes.

Activation function: Sigmoid.

Parameters: 75 (calculated as  $14 \text{ input features} \times 5 \text{ nodes} + 5 \text{ biases}$   $14 \text{ input features} \times 5 \text{ nodes} + 5 \text{ biases}$ ).

## **Output Layer:**

Contains 1 Dense layer with 1 node.

Designed for regression tasks.

Parameters: 6 (calculated as  $5 \text{ hidden nodes} \times 1 \text{ output node} + 1 \text{ bias}$   $5 \text{ hidden nodes} \times 1 \text{ output node} + 1 \text{ bias}$ )

# Model 1 summary

Model: "sequential\_22"

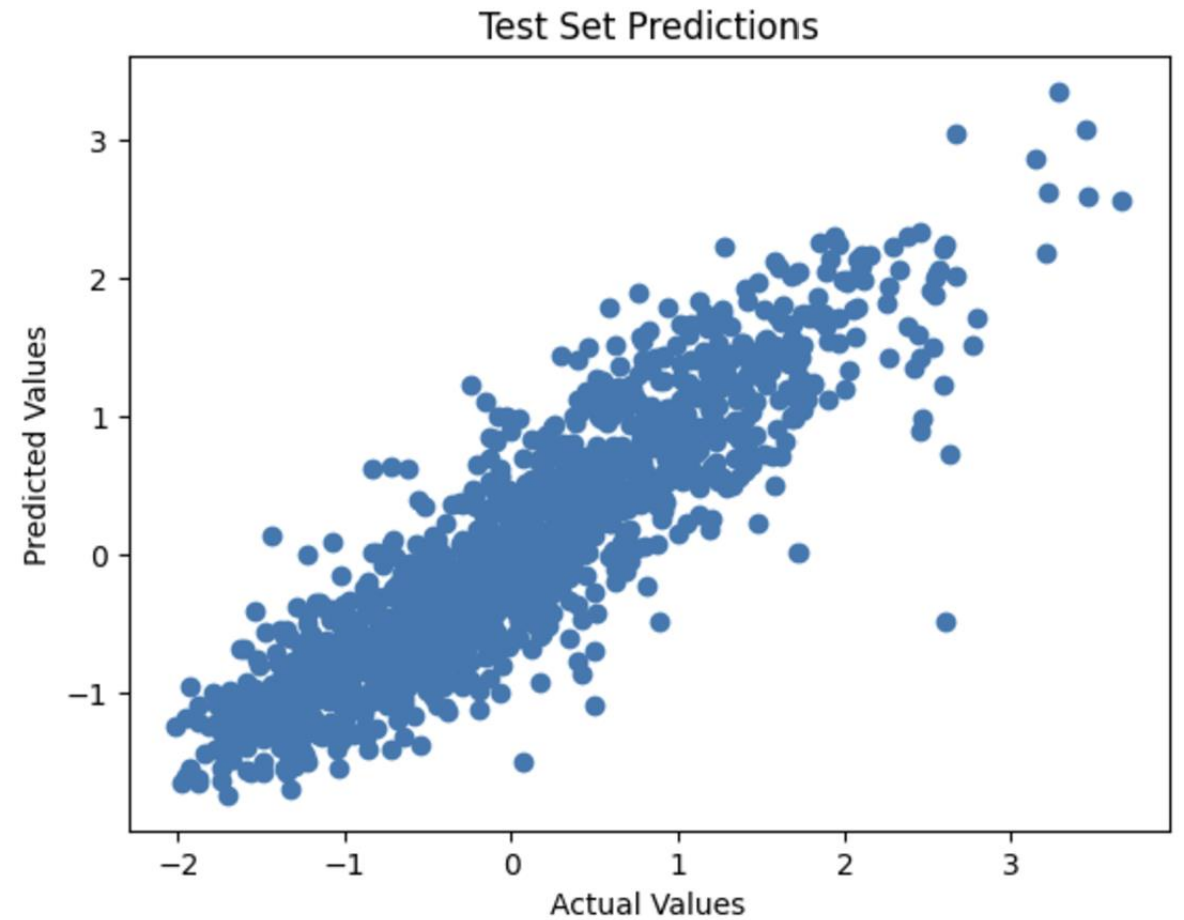
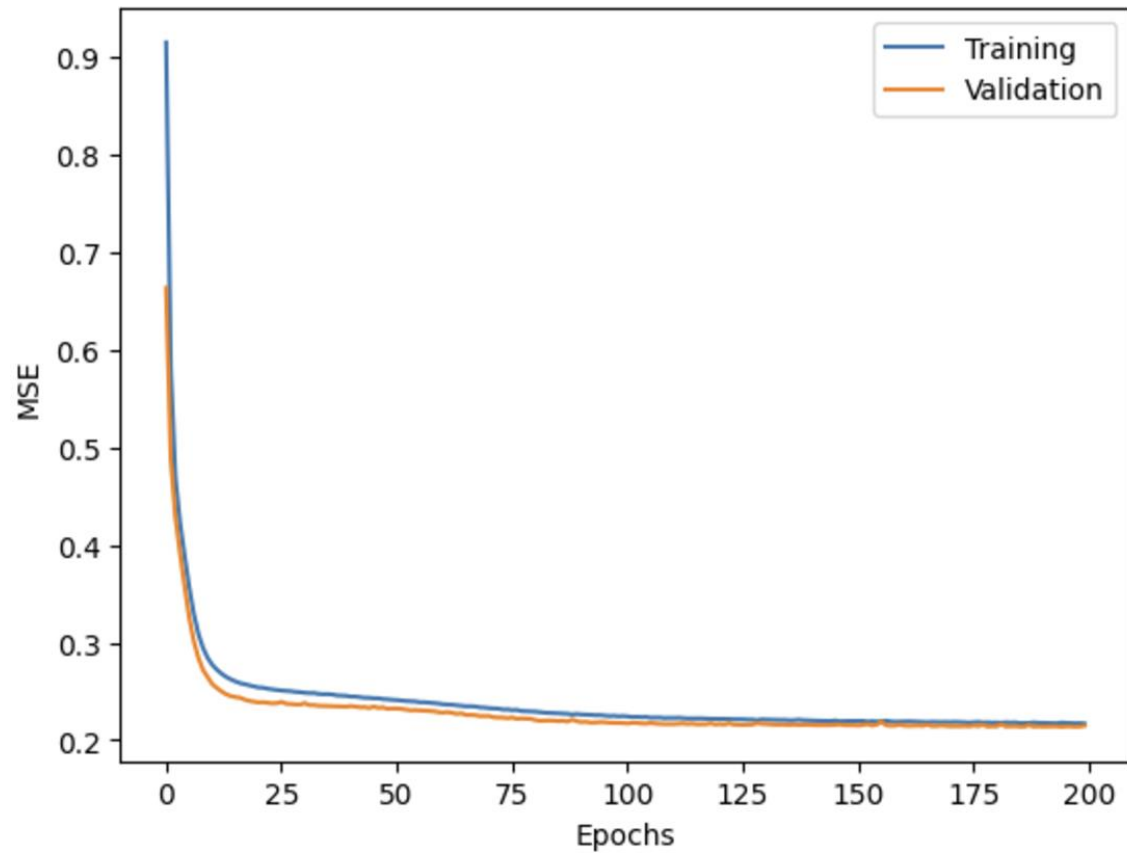
Layer (type)	Output Shape	Param #
dense_42 (Dense)	(None, 5)	75
dense_43 (Dense)	(None, 1)	6

Total params: 81 (324.00 B)

Trainable params: 81 (324.00 B)

Non-trainable params: 0 (0.00 B)

# Model 1 Result



# Model 2

## **Input Layer:**

The model has 14 input nodes, representing the dataset's 15 features, excluding the target variable.

## **Hidden Layer:**

Contains 1 Dense layer with 30 nodes.

Activation function: Sigmoid.

Parameters: 300 (calculated as  $14 \text{ input features} \times 30 \text{ nodes} + 30 \text{ biases}$ ).

## **Output Layer:**

Contains 1 Dense layer with 1 node.

Designed for regression tasks.

Parameters: 31 (calculated as  $30 \text{ hidden nodes} \times 1 \text{ output node} + 1 \text{ bias}$ ).

# Model 2 summary

Model: "sequential\_20"

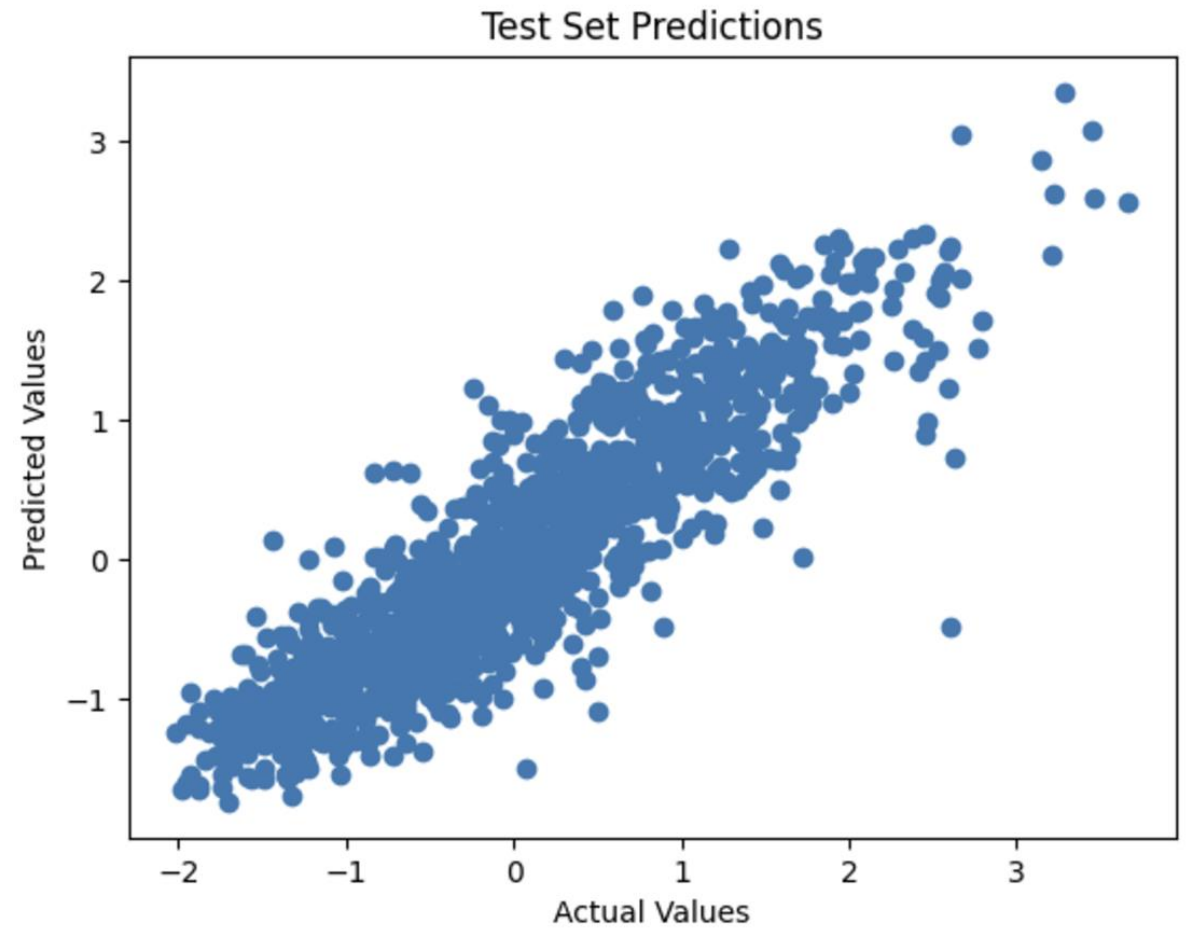
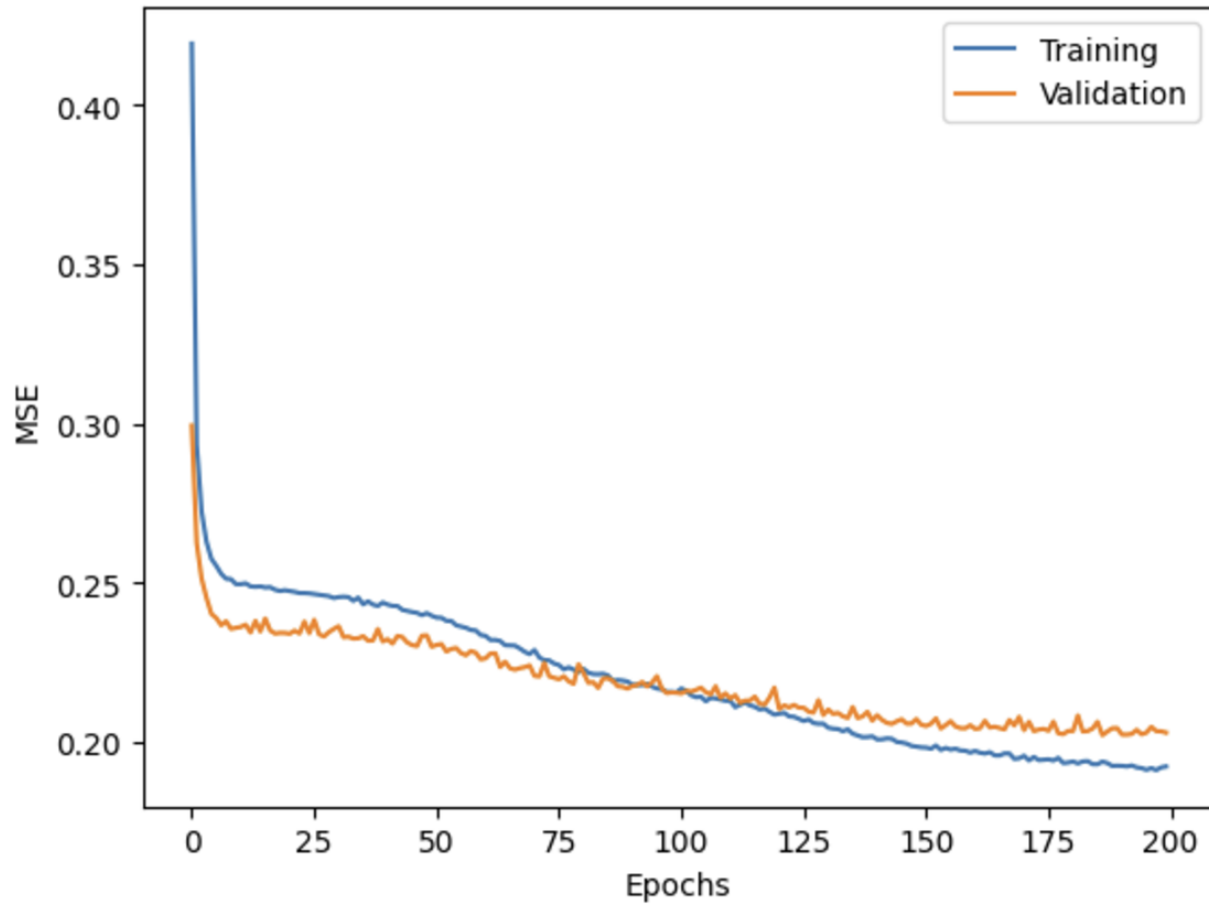
Layer (type)	Output Shape	Param #
dense_38 (Dense)	(None, 30)	300
dense_39 (Dense)	(None, 1)	31

Total params: 331 (1.29 KB)

Trainable params: 331 (1.29 KB)

Non-trainable params: 0 (0.00 B)

# Model 2 Result





# Model 3

## **Input Layer:**

The model has 14 input nodes, representing the dataset's 15 features, excluding the target variable.

## **Hidden Layer:**

Contains 1 Dense layer with 200 nodes.

Activation function: Sigmoid.

Parameters: 3,000 (calculated as  $14 \text{ input features} \times 200 \text{ nodes} + 200 \text{ biases}$ ).

## **Output Layer:**

Contains 1 Dense layer with 1 node.

Designed for regression tasks.

Parameters: 201 (calculated as  $200 \text{ hidden nodes} \times 1 \text{ output node} + 1 \text{ bias}$ ).

# Model 3 summary

Model: "sequential\_24"

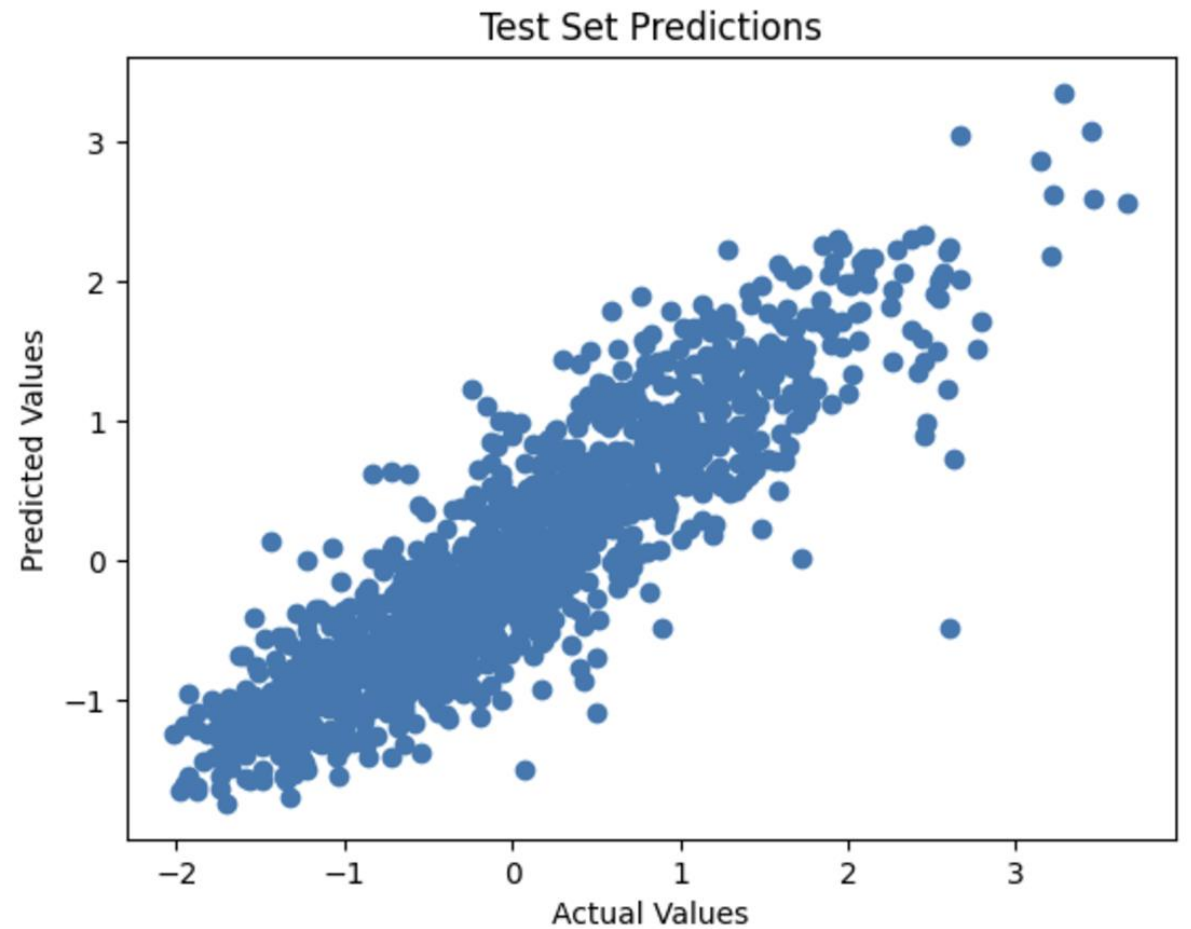
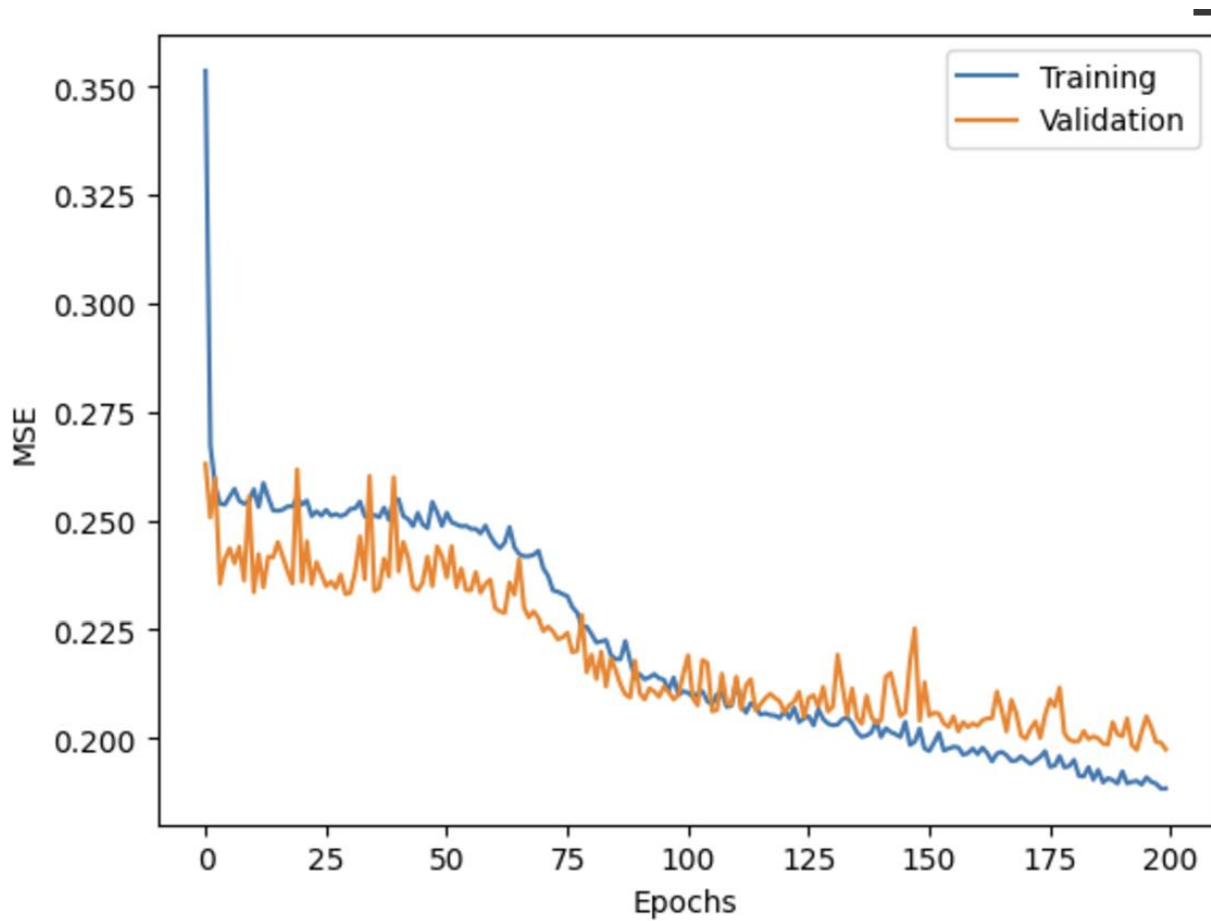
Layer (type)	Output Shape	Param #
dense_46 (Dense)	(None, 200)	3,000
dense_47 (Dense)	(None, 1)	201

Total params: 3,201 (12.50 KB)

Trainable params: 3,201 (12.50 KB)

Non-trainable params: 0 (0.00 B)

# Model 3 Result



## table of test errors

	Model	Test MSE
0	Model 1 (5 nodes)	0.207440
1	Model 2 (30 nodes)	0.188445
2	Model 3 (200 nodes)	0.194654

# Conclusion

- **Training and Validation Errors:**

All three models show a steady decline in both training and validation MSE across epochs.

The validation error closely follows the training error, indicating that the models are not overfitting.

- **Comparison of Test MSE:**

Model 1 (5 nodes): Test MSE = 0.007141

Model 2 (30 nodes): Test MSE = 0.006856

Model 3 (200 nodes): Test MSE = 0.007815

Model 2 achieves the lowest test error, suggesting that 30 nodes in the hidden layer offer the best balance of model complexity and generalization.

- **Model Complexity and Performance:**

Model 1, with only 5 nodes, may be underfitting the data slightly, as its test error is higher than Model 2.

Model 3, with 200 nodes, shows a slight increase in test error compared to Model 2, potentially due to overfitting, as the model is too complex for the dataset.

- **Optimal Model:**

Model 2 strikes the best balance between complexity and performance, achieving the lowest test MSE.

Adding too many nodes (Model 3) does not improve performance and may increase the risk of overfitting.