```
In [87]: !pip install numpy==2.0.2
!pip install pandas==2.2.2
!pip install tensorflow_cpu==2.18.0
!pip install scikit-learn
```

```
Requirement already satisfied: numpy==2.0.2 in /opt/conda/lib/python3.11/site-pac
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Requirement already satisfied: tensorflow-io-gcs-filesystem>=0.23.1 in /opt/cond
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a/lib/python3.11/site-packages (from tensorflow_cpu==2.18.0) (0.37.1)
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Requirement already satisfied: threadpoolctl>=3.1.0 in /opt/conda/lib/python3.11/
site-packages (from scikit-learn) (3.5.0)
 import numpy as np
 import pandas as pd
```

```
In [67]: #Importing necessary libraries
  import numpy as np
  import pandas as pd
  from sklearn.model_selection import train_test_split
  from sklearn.preprocessing import StandardScaler
  from tensorflow.keras.models import Sequential
  from tensorflow.keras.layers import Dense, Input
  from sklearn.metrics import mean_squared_error
```

```
In [88]: # Reading the data
url = 'https://cocl.us/concrete_data'
data = pd.read_csv(url)
```

In [96]: # To get the count of rows and columns of the dataset
 data.shape

Out[96]: (1030, 9)

In [97]: # To display top 5 rows of the dataset
 data.head()

Out[97]:

:		Cement	Blast Furnace Slag	Fly Ash	Water	Superplasticizer	Coarse Aggregate	Fine Aggregate	Age	Streng
	0	540.0	0.0	0.0	162.0	2.5	1040.0	676.0	28	79.
	1	540.0	0.0	0.0	162.0	2.5	1055.0	676.0	28	61.8
	2	332.5	142.5	0.0	228.0	0.0	932.0	594.0	270	40.7
	3	332.5	142.5	0.0	228.0	0.0	932.0	594.0	365	41.0
	4	198.6	132.4	0.0	192.0	0.0	978.4	825.5	360	44.
	4									

In [98]: # To get the details of datatype for each column and memory required
 data.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1030 entries, 0 to 1029
Data columns (total 9 columns):

#	Column	Non-Null Count	Dtype
0	Cement	1030 non-null	float64
1	Blast Furnace Slag	1030 non-null	float64
2	Fly Ash	1030 non-null	float64
3	Water	1030 non-null	float64
4	Superplasticizer	1030 non-null	float64
5	Coarse Aggregate	1030 non-null	float64
6	Fine Aggregate	1030 non-null	float64
7	Age	1030 non-null	int64
8	Strength	1030 non-null	float64

dtypes: float64(8), int64(1)

memory usage: 72.6 KB

In [99]: # To get overall statistics of the dataset
data.describe()

Out[99]:		Cement	Blast Furnace Slag	Fly Ash	Water	Superplasticizer	Coarse Aggregate
	count	1030.000000	1030.000000	1030.000000	1030.000000	1030.000000	1030.000000
	mean	281.167864	73.895825	54.188350	181.567282	6.204660	972.91893
	std	104.506364	86.279342	63.997004	21.354219	5.973841	77.753954
	min	102.000000	0.000000	0.000000	121.800000	0.000000	801.000000
	25%	192.375000	0.000000	0.000000	164.900000	0.000000	932.000000
	50%	272.900000	22.000000	0.000000	185.000000	6.400000	968.000000
	75%	350.000000	142.950000	118.300000	192.000000	10.200000	1029.400000
	max	540.000000	359.400000	200.100000	247.000000	32.200000	1145.000000
	4						•
In [100	# To display null values in each column data.isnull().sum()						
Out[100	Cement 0 Blast Furnace Slag 0 Fly Ash 0 Water 0 Superplasticizer 0 Coarse Aggregate 0 Fine Aggregate 0 Age 0 Strength 0 dtype: int64						
In [101	<pre>#To display the columns of the dataset concrete_data_columns = data.columns data.columns</pre>						
Out[101			regate', 'Fi	e Slag', 'Fl ne Aggregate	-	er', 'Superplas trength'],	ticizer',
In [102	<pre>#Define the predictors and the Target predictors = data[concrete_data_columns[concrete_data_columns != 'Strength']] target = data['Strength'] print(predictors.head())</pre>						ength']]

```
Cement Blast Furnace Slag Fly Ash Water Superplasticizer \
        0
            540.0
                                  0.0
                                           0.0 162.0
        1
            540.0
                                  0.0
                                           0.0 162.0
                                                                    2.5
        2
           332.5
                                142.5
                                          0.0 228.0
                                                                    0.0
        3
           332.5
                                142.5
                                          0.0 228.0
                                                                    0.0
                                132.4
                                          0.0 192.0
        4
            198.6
                                                                    0.0
           Coarse Aggregate Fine Aggregate Age
        0
                     1040.0
                                      676.0
        1
                     1055.0
                                      676.0
        2
                                      594.0 270
                      932.0
        3
                      932.0
                                      594.0 365
                                      825.5 360
        4
                      978.4
In [103...
         print(target.head())
        0
             79.99
        1
             61.89
        2
             40.27
        3
             41.05
             44.30
        Name: Strength, dtype: float64
 In [ ]: n_cols = predictors_norm.shape[1] # Number of predictors
In [104...
          # Step A: Without Normalization and Epoch = 50, Evaluate 50 Times
          # Define the regression model with 1 hidden layer
          def regression_model():
              model = Sequential()
              model.add(Input(shape=(n_cols,))) # Specify the input shape
              model.add(Dense(10, activation='relu')) # Hidden Layer with 10 nodes and Re
              model.add(Dense(1)) # Output Layer
              model.compile(optimizer='adam', loss='mean_squared_error') # Compile the mo
              return model
          # Initialize an empty list to store the mean squared errors
          mse list = []
          # Loop to train and evaluate the model 50 times
          for i in range(50):
              # Split the data into training and test sets
              X_train, X_test, y_train, y_test = train_test_split(predictors, target, test
              # Build the model
              model = regression_model()
              # Train the model on the training data
              model.fit(X_train, y_train, epochs=50, verbose=0)
              # Evaluate the model on the test data
              predictions = model.predict(X test)
              mse = mean_squared_error(y_test, predictions)
              mse_list.append(mse)
          mean mse A = np.mean(mse list)
          std_mse_A = np.std(mse_list)
          print(f"Mean of Mean Squared Errors (Step A - No Normalization, 50 epochs): {mea
          print(f"Standard Deviation of Mean Squared Errors (Step A - No Normalization, 50
```

		_
10/10	0s	14ms/step
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10/10	0s	14ms/step
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20, 20	0s	6ms/step
10/10	0s	5ms/step
10/10	0s	6ms/step
20, 20	0s	6ms/step
,	0s	5ms/step
-0, -0	0s	6ms/step
10/10	0s	6ms/step

Mean of Mean Squared Errors (Step A - No Normalization, 50 epochs): 303.630609049

Standard Deviation of Mean Squared Errors (Step A - No Normalization, 50 epochs): 233.7104445832209

```
In [105... # Step B: With Normalization and Epoch = 50, Evaluate 50 Times, Compare with Ste
    # Normalize the predictors
    predictors_norm = (predictors - predictors.mean()) / predictors.std()
```

```
# Initialize an empty list to store the mean squared errors
mse_list = []
# Loop to train and evaluate the model 50 times
for i in range(50):
   # Split the data into training and test sets
   X_train, X_test, y_train, y_test = train_test_split(predictors_norm, target,
   # Build the model
   model = regression_model()
   # Train the model on the training data
   model.fit(X_train, y_train, epochs=50, verbose=0)
   # Evaluate the model on the test data
   predictions = model.predict(X_test)
    mse = mean_squared_error(y_test, predictions)
   mse_list.append(mse)
mean_mse_B = np.mean(mse_list)
std_mse_B = np.std(mse_list)
print(f"Mean of Mean Squared Errors (Step B - With Normalization, 50 epochs): {m
print(f"Standard Deviation of Mean Squared Errors (Step B - With Normalization,
print(f"Difference in Mean Squared Errors between Step A and Step B: {mean_mse_A
```

10/10	0s	5ms/step
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10/10	0s	6ms/step
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	0s	6ms/step
10/10		6ms/step
10/10	0s	5ms/step
10/10	0s	7ms/step
10/10	0s	5ms/step
10/10	0s	5ms/step
10/10	0s	6ms/step
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10/10	0s	6ms/step
10/10	0s	6ms/step

Mean of Mean Squared Errors (Step B - With Normalization, 50 epochs): 370.2345658

Standard Deviation of Mean Squared Errors (Step B - With Normalization, 50 epoch s): 107.75069223008683

Difference in Mean Squared Errors between Step A and Step B: -66.60395680067569

```
In [106... # Step C: With Normalization and Epoch = 100, Compare with Step B
# Initialize an empty list to store the mean squared errors
mse_list = []
```

```
# Loop to train and evaluate the model 50 times
for i in range(50):
   # Split the data into training and test sets
   X_train, X_test, y_train, y_test = train_test_split(predictors_norm, target,
   # Build the model
   model = regression_model()
   # Train the model on the training data
   model.fit(X_train, y_train, epochs=100, verbose=0)
   # Evaluate the model on the test data
   predictions = model.predict(X_test)
   mse = mean_squared_error(y_test, predictions)
   mse_list.append(mse)
mean_mse_C = np.mean(mse_list)
std_mse_C = np.std(mse_list)
print(f"Mean of Mean Squared Errors (Step C - With Normalization, 100 epochs): {
print(f"Standard Deviation of Mean Squared Errors (Step C - With Normalization,
print(f"Difference in Mean Squared Errors between Step B and Step C: {mean_mse_B
```

10/10	0s	5ms/step
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10/10	0s	5ms/step
10/10	0s	6ms/step
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10/10	0s	8ms/step
10/10	0s	7ms/step
10/10	0s	6ms/step
10/10	0s	6ms/step
10/10	0s	5ms/step
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10/10	0s	6ms/step
10/10	0s	7ms/step
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Mean of Mean Squared Errors (Step C - With Normalization, 100 epochs): 165.212853 79928682

Standard Deviation of Mean Squared Errors (Step C - With Normalization, 100 epoch s): 18.51426082827909

Difference in Mean Squared Errors between Step B and Step C: 205.02171205062515

```
In [107... # Define the regression model with three hidden layers
def regression_model_increased_layers():
    model = Sequential()
    model.add(Input(shape=(predictors_norm.shape[1],))) # Specify the input sha
```

```
model.add(Dense(10, activation='relu')) # First hidden Layer with 10 nodes
    model.add(Dense(10, activation='relu')) # Second hidden Layer with 10 nodes
    model.add(Dense(10, activation='relu')) # Third hidden Layer with 10 nodes
    model.add(Dense(1)) # Output Layer
    model.compile(optimizer='adam', loss='mean_squared_error') # Compile the mo
    return model
# Initialize an empty list to store the mean squared errors
mse_list = []
# Loop to train and evaluate the model 50 times
for i in range(50):
   # Split the data into training and test sets
   X_train, X_test, y_train, y_test = train_test_split(predictors_norm, target,
   # Build the model
   model = regression_model_increased_layers()
   # Train the model on the training data
   model.fit(X_train, y_train, epochs=50, verbose=0)
   # Evaluate the model on the test data
   predictions = model.predict(X_test)
   mse = mean_squared_error(y_test, predictions)
    mse_list.append(mse)
mean_mse_D = np.mean(mse_list)
std_mse_D = np.std(mse_list)
print(f"Mean of Mean Squared Errors (Step D - Increased Layers, With Normalizati
print(f"Standard Deviation of Mean Squared Errors (Step D - Increased Layers, Wi
print(f"Difference in Mean Squared Errors between Step B and Step D: {mean_mse_B
```

10/10	0s	7ms/step
10/10	0s	9ms/step
10/10	0s	8ms/step
10/10	0s	8ms/step
10/10	0s	7ms/step
10/10	0s	7ms/step
10/10	0s	8ms/step
10/10	0s	8ms/step
10/10	0s	7ms/step
10/10	0s	8ms/step
10/10	0s	8ms/step
10/10	0s	6ms/step
10/10	0s	7ms/step
10/10	0s	7ms/step
10/10	0s	8ms/step
10/10	0s	9ms/step
10/10	0s	8ms/step
10/10	0s	8ms/step
10/10	0s	9ms/step
10/10	0s	9ms/step
10/10	0s	7ms/step
10/10	0s	7ms/step
10/10	0s	7ms/step
10/10	0s	8ms/step
10/10	0s	7ms/step
10/10	0s	7ms/step
10/10	0s	9ms/step
10/10	0s	7ms/step
10/10	0s	7ms/step
10/10	0s	8ms/step
10/10	0s	8ms/step
10/10	0s	10ms/step
10/10		9ms/step
10/10	0s	11ms/step
10/10	0s	8ms/step
,	0s	9ms/step
10/10	0s	8ms/step
,	0s	8ms/step
10/10	0s	9ms/step
10/10	0s	8ms/step
	0s	8ms/step
-0, -0	0s	8ms/step
10, 10		7ms/step
20/ 20		7ms/step
,		7ms/step
	0s	7ms/step
10/10 ——————————————————————————————————		, ,

Mean of Mean Squared Errors (Step D - Increased Layers, With Normalization, 50 ep ochs): 125.87031561843013

Standard Deviation of Mean Squared Errors (Step D - Increased Layers, With Normal ization, $50 \ \text{epochs}$): 16.765422916874382

Difference in Mean Squared Errors between Step B and Step D: 244.36425023148183