HW1

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[11]: #Problem 1
      "a) x1 has 5 features, x2 has 4 features"
[11]: 'a) x1 has 5 features, x2 has 4 features'
 [3]: #b)
      import numpy as np
      # Define the vectors
      x1 = np.array([2, 3, 4, 8, 9])
      x2 = np.array([2, -3, -4, 89])
      # Calculate norms for x1
      L1\_norm\_x1 = np.sum(np.abs(x1))
      L2\_norm\_x1 = np.sqrt(np.sum(x1**2))
      L_{inf_norm_x1} = np.max(np.abs(x1))
      # Calculate norms for x2
      L1\_norm\_x2 = np.sum(np.abs(x2))
      L2\_norm\_x2 = np.sqrt(np.sum(x2**2))
      L_inf_norm_x2 = np.max(np.abs(x2))
      # Print results
      print("Vector x1:")
      print(f"L1 Norm: {L1_norm_x1}")
      print(f"L2 Norm: {L2_norm_x1}")
      print(f"LâĹd Norm: {L_inf_norm_x1}")
      print("\nVector x2:")
      print(f"L1 Norm: {L1_norm_x2}")
      print(f"L2 Norm: {L2_norm_x2}")
      print(f"LâĹd Norm: {L_inf_norm_x2}")
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Vector x1:

L1 Norm: 26 L2 Norm: 13.19090595827292 LâĹđ Norm: 9 Vector x2: L1 Norm: 98 L2 Norm: 89.16277250063504 LâĹđ Norm: 89 [4]: #Problem 2 input_height, input_width, input_channels = 1024, 1024, 3 output_height, output_width, output_channels = 64, 64, 3 # Calculate input vector length input_vector_length = input_height * input_width * input_channels # Calculate output vector length output_vector_length = output_height * output_width * output_channels # Matrix W dimensions and vector b size W_elements = output_vector_length * input_vector_length b_elements = output_vector_length # Print results print("Input Vector Length:", input_vector_length) print("Output Vector Length:", output_vector_length) print("Number of elements in Matrix W:", W_elements) print("Number of elements in Vector b:", b_elements) Input Vector Length: 3145728 Output Vector Length: 12288 Number of elements in Matrix W: 38654705664 Number of elements in Vector b: 12288 [5]: #Problem 3 import numpy as np # Define the matrix W W = np.array([[1, -1], [2, 0]])# Calculate norms L1_norm = np.max(np.sum(np.abs(W), axis=0)) # Maximum column sum L2_norm = np.linalg.norm(W, 2) # Spectral norm (2-norm) L_inf_norm = np.max(np.sum(np.abs(W), axis=1)) # Maximum row sum

Frobenius norm

Frobenius_norm = np.sqrt(np.sum(W**2))

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# Print results
     print(f"L1 Norm: {L1_norm}")
     print(f"L2 Norm: {L2_norm}")
     print(f"LâĹd Norm: {L_inf_norm}")
     print(f"Frobenius Norm: {Frobenius_norm}")
    L1 Norm: 3
    L2 Norm: 2.2882456112707374
    LâĹđ Norm: 2
    Frobenius Norm: 2.449489742783178
[6]: #Problem 4
    from sklearn.datasets import load_iris
     from sklearn.model_selection import train_test_split
     from sklearn.linear_model import LogisticRegression
     from sklearn.metrics import accuracy_score
     # Load the Iris dataset
     iris = load_iris()
     X = iris.data # Features (4 columns: sepal length, sepal width, petal length,
     \rightarrowpetal width)
     y = iris.target # Target (class labels)
     # Choose 3 out of the 4 features (e.g., 0, 1, 2)
     X_selected = X[:, :3] # Selecting the first 3 features
     # Split the data into training and testing sets
     X_train, X_test, y_train, y_test = train_test_split(X_selected, y, test_size=0.
     \rightarrow2, random_state=42)
     # Initialize and train a Logistic Regression model
     model = LogisticRegression(max_iter=200) # Increase max_iter for convergence
     model.fit(X_train, y_train)
     # Make predictions
     y_pred = model.predict(X_test)
     # Calculate accuracy
     accuracy = accuracy_score(y_test, y_pred)
     # Print the accuracy
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Accuracy using 3 features: 100.00%

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[8]: #Problem 5 import numpy as np
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print(f"Accuracy using 3 features: {accuracy * 100:.2f}%")

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import tensorflow as tf
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense, Input
from sklearn.model_selection import train_test_split
from sklearn.metrics import mean_squared_error
# Generate data: integers from 1 to 100 and their 7th root
X = np.arange(1, 101).reshape(-1, 1) # Input data (1 to 100)
y = X**(1/7) # Target data (7th root)
# Split data into training and testing sets
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,__
 →random_state=42)
# Define the neural network model
model = Sequential([
    Input(shape=(1,)),
                                       # Input layer explicitly defined
    Dense(10, activation='relu'),
                                      # Hidden layer with 10 neurons
    Dense(10, activation='relu'),
                                      # Another hidden layer
    Dense(1)
                                        # Output layer
])
# Compile the model
model.compile(optimizer='adam', loss='mse', metrics=['mae'])
# Train the model
history = model.fit(X_train, y_train, epochs=500, verbose=0)
# Evaluate the model on the test data
y_pred = model.predict(X_test)
mse = mean_squared_error(y_test, y_pred)
print(f"Mean Squared Error on Test Data: {mse:.4f}")
# Test the model with a few examples
test_values = np.array([[10], [50], [100]])
predictions = model.predict(test_values)
for i, val in enumerate(test_values.flatten()):
    print(f"7th root of {val}: True = {val**(1/7):.4f}, Predicted =
 \hookrightarrow {predictions[i][0]:.4f}")
Mean Squared Error on Test Data: 0.0863
7th root of 10: True = 1.3895, Predicted = 0.9867
7th root of 50: True = 1.7487, Predicted = 1.5464
7th root of 100: True = 1.9307, Predicted = 2.2461
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[10]: #Problem 6
      import numpy as np
      import pandas as pd
      from sklearn.datasets import load_iris
      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
      # Load the Iris dataset
      iris = load_iris()
      data = pd.DataFrame(iris.data, columns=iris.feature_names)
      data['target'] = iris.target
      # Extract input (sepal length) and output (petal length)
      X = data['sepal length (cm)'].values.reshape(-1, 1) # Input: Sepal length
      y = data['petal length (cm)'].values # Output: Petal length
      # Split the data into training and testing sets
      X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,
      →random state=42)
      # Scale the input features
      scaler = StandardScaler()
      X_train = scaler.fit_transform(X_train)
      X_test = scaler.transform(X_test)
      # Build the neural network model
      model = Sequential([
          Input(shape=(1,)),
                                                # Input layer explicitly defined
          Dense(10, activation='relu'),
                                                # Hidden layer with 10 neurons
          Dense(10, activation='relu'),
                                                # Hidden layer with 10 neurons
          Dense(1)
                                                 # Output layer with 1 neuron_
      \hookrightarrow (regression)
      ])
      # Compile the model
      model.compile(optimizer='adam', loss='mse', metrics=['mae'])
      # Train the model
      history = model.fit(X_train, y_train, epochs=300, verbose=0)
      # Evaluate the model
      y_pred = model.predict(X_test)
      mse = mean_squared_error(y_test, y_pred)
      print(f"Mean Squared Error on Test Data: {mse:.4f}")
```