
Assignment 4

Qestion 1: Write a Program

- Calculate Euclidian distance between two 2D points
- Calculate Manhattan (taxicab) distance between two 2D points
- Calculate Chebyshev (chessboard) distance between two 2D points.
- Calculate Minkowski distance with parameter p between two 2D points.

When p=1: Manhattan distance

When p=2: Euclidean distance

When $p \rightarrow \infty$: Chebyshev distance

- Calculate Hamming distance between two binary vectors.
- Calculate cosine distance between two vectors
- Calculate Mahalanobis distance between two vectors.

Qestion 2: Data Transformation

(Min max Transformation)

• Generate sample data

```
np.random.seed(42)
original_data = np.random.normal(loc=50, scale=15, size=100)
```

Manual min-max scaling (0 to 1)

Manually perform min-max scaling on a numpy array or list

Formula: $X_{scaled} = (X - X_{min}) / (X_{max} - X_{min})$

Parameters:

X: numpy array or list to be normalized

Returns:

numpy array with normalized values between 0 and 1

• Custom range min-max scaling (-1 to 1)

Scale data to a custom range [new_min, new_max]

Formula: $X_{\text{scaled}} = (X - X_{\text{min}}) / (X_{\text{max}} - X_{\text{min}}) * (\text{new}_{\text{max}} - \text{new}_{\text{min}}) + \text{new}_{\text{min}}$

Parameters:

X: numpy array or list to be normalized

new_min: minimum value of the new range

new_max: maximum value of the new range

Returns:

numpy array with normalized values between new_min and new_max

Using scikit-learn's MinMaxScaler

• Display results

Min:

Max:

Mean:

Standard Deviation:

• Create a DataFrame for easy comparison

First 5 rows of all datasets

• Visualize the transformations

Question 3:

Given the following predictions and actual values:

```
y_{true} = [1, 0, 1, 1, 0, 0, 1, 0]
```

$$y_pred = [1, 0, 0, 1, 1, 0, 1, 1]$$

Calculate

- **True Positives (TP):** Correctly predicted positives
- **False Positives (FP):** Incorrectly predicted positives
- True Negatives (TN): Correctly predicted negatives
- False Negatives (FN): Incorrectly predicted negatives
- **Accuracy**: Overall correctness (TP + TN) / Total
- **Precision**: How many predicted positives were correct TP / (TP + FP)
- **Recall**: How many actual positives were found TP / (TP + FN)
- **F1 Score**: Harmonic mean of precision and recall
- **Mean Squared Error (MSE) Formula:** MSE = $(1/n) * \Sigma(y_{true} y_{pred})^2$
- Root Mean Square Error (RMSE) Formula: RMSE = sqrt($(1/n) * \Sigma(y_true y_pred)^2$)
- **Mean Absolute Error (MAE) Formula:** MAE = $(1/n) * \Sigma | y_{true} y_{pred} |$
- **Binary Cross-Entropy Loss :** Formula: -[y_true * log(y_pred) + (1 y_true) * log(1 y_pred)]