

EPOCH TASK SUBMISSION

Roshan Y Singh

Task 1: Classification Model

Methods Used:

Data Loading and Cleaning:

- Loaded the data from a CSV file using **pandas**.
- Converted the 'Latitude' and 'Longitude' columns to numeric, coercing errors.
- Filtered data to include only records from the state of Karnataka.
- Dropped rows with NaN values in 'Latitude' and 'Longitude'.
- Filtered latitude and longitude to be within valid ranges for Karnataka.

K-Means Clustering:

- **Initialization of Centroids:** Randomly selected initial centroids from the data.
 - **Cluster Assignment:** Assigned each data point to the nearest centroid using Euclidean distance.
 - **Centroid Update:** Recomputed centroids as the mean of data points in each cluster.
 - **Convergence Check:** Repeated assignment and update steps until centroids no longer changed or max iterations reached.
 - **Within-Cluster Sum of Squares (WCSS):** Calculated to evaluate cluster compactness.
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- **Silhouette Score:** Measured how similar a data point is to its own cluster compared to other clusters.
 - **Davies-Bouldin Index:** Evaluated cluster separation, with lower values indicating better clustering.

Evaluation Metrics:

- **Elbow Method:** Plotted WCSS against the number of clusters to identify the optimal k.
- **Silhouette Score:** Plotted to determine the quality of clustering.
- **Davies-Bouldin Index:** Plotted to evaluate cluster compactness and separation.

Visualization:

- Plotted the Elbow method, Silhouette score, and Davies-Bouldin Index to determine the optimal number of clusters.
- Visualized the final clusters and centroids on a scatter plot.

Inferences:

Optimal Number of Clusters:

- The Elbow method, Silhouette score, and Davies-Bouldin Index are crucial for determining the appropriate number of clusters.
- Chose an optimal number of clusters based on the evaluation metrics.
- Also let the number of clusters reflect the number of districts in the state.

Density of Pin Codes:

- These pin codes seemed to well-outline my home state of Karnataka.
- There were a greater number of regions included around the cities, such as Bengaluru and Mysuru, whereas those near the borders with other states and the regions around the Western Ghats were sparse.
- Letting the number of clusters as 4, also lets us in on the four divisions of Karnataka- Bengaluru, Mysuru, Belagavi and Kalaburgi.

Use of Additional Methods:

- Used scales apart from the Elbow plot to choose the optimum number of clusters.
- Also, the centroids seem to fall towards the district centers of particular districts

Reference Links:

- Statquest Video Resources given in the document.
- Scikit-learn Clustering Documentation
- [Silhouette Score - Wikipedia](#)
- [Davies-Bouldin Index - Wikipedia](#)
- Use of AI tool - ChatGPT

Task 2: Classification Model

Code Overview

This project involves two main tasks: Optical Character Recognition (OCR) using a Convolutional Neural Network (CNN) and sentiment analysis using a Logistic Regression model with TF-IDF vectorization. The objective is to recognize text from images, and then analyze the sentiment of the recognized text.

1. OCR using CNN

Steps Involved:

1. Data Cleaning and Preparation:

- Load the alphabet dataset (`alphabets_28x28.csv`) which contains 28x28 pixel images of letters.
- Clean the dataset by removing rows with corrupted data.
- Convert the data into a format suitable for training a CNN.

2. Building the CNN Model:

- Define a CNN with two convolutional layers followed by max-pooling layers, a flattening layer, and two dense layers.
- Compile the model using the Adam optimizer and categorical cross-entropy loss.

3. Training the CNN Model:

- Split the data into training and testing sets.
- Train the model and evaluate its accuracy on the test set.

4. Segmenting and Predicting Letters:

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- Segment input images into smaller images representing individual letters.
 - Use the trained CNN to predict each letter.
 - Reconstruct the recognized text from individual letter predictions.

2. Sentiment Analysis

Steps Involved:

1. Data Preparation:

- Load the sentiment analysis dataset which contains text lines and their associated sentiment labels.
- Preprocess the text by converting it to lowercase and removing punctuation.

2. Training the Sentiment Analysis Model:

- Use TF-IDF vectorization to convert text data into numerical features.
- Train a Logistic Regression model using these features.

3. Evaluating the Sentiment Analysis Model:

- Split the data into training and testing sets.
- Evaluate the model's accuracy on the test set.

4. Predicting Sentiment on OCR Text:

- Use the trained sentiment analysis model to predict the sentiment of the text recognized by the OCR system.

Inferences:

- Observed good accuracies with the CNN model(Around 99%)
- Relatively good accuracies from the Sentiment Analysis Model (83.33%)

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- The Epochs took some time to work and process, and so did the images, however, eventually, the sentences observed were clean and correct
 - The use on TensorFlow module was a keynote
 - The integrated system's overall sentiment analysis accuracy on OCR predictions was approximately 75%, indicating that the pipeline from text recognition to sentiment classification performed reasonably well but had some room for improvement.

Future Improvements:

1. **Improved Segmentation:** Enhance image segmentation techniques to better handle varied character sizes and fonts.
2. **Advanced Preprocessing:** Implement more sophisticated preprocessing steps for both image and text data.
3. **Model Optimization:** Experiment with different CNN architectures and hyperparameters to further improve accuracy.
4. **Additional Metrics:** Evaluate models using additional metrics such as precision, recall, and F1-score to gain more insights into performance.

References:

1. TensorFlow Documentation: [TensorFlow](#)
2. OpenCV Documentation: [OpenCV](#)
3. Scikit-Learn Documentation: [Scikit-Learn](#)
4. Video Resources given in the document, such as StatQuest and NeuralNine videos were very informative.
5. Use of AI Tool - ChatGPT