ASL Alphabet Image Recognition – Code Demonstration

Project Summary

This project classifies American Sign Language (ASL) letters (A–Z plus del, space, and nothing) using image data. A pre-trained **VGG16 model** is used for feature extraction, and custom dense layers are added to specialize the network for ASL classification.

Step-by-Step Code Explanation

1. Setup & Environment Configuration

- Kaggle Authentication: Copy the kaggle.json API key to ~/.kaggle/.
- Download Dataset: Use kaggle datasets download -d grassknoted/asl-alphabet.
- Unzip the contents into a directory called asl-alphabet.

2. Dependencies

- TensorFlow/Keras (model building)
- OpenCV & PIL (image loading/processing)
- NumPy, Pandas (data handling)
- Plotly, Matplotlib (visualization)
- imutils, sklearn, tqdm, glob

🗱 3. Configuration & Seeding

- Hyperparameters are stored in a class CFG.
- batch_size = 64
- img_height = 64
- img_width = 64
- epochs = 10
- num classes = 29

4. Label Creation & Metadata Building

- Uses string.ascii_uppercase + ['del', 'nothing', 'space'] to define 29 class labels.
- Each label's directory is scanned for image paths and stored in a metadata DataFrame.

★ 5. Dataset Splitting

- Uses train_test_split (stratified) to split metadata into:
 - o 70% Training
 - o 15% Validation
 - 15% Testing

6. Image Data Generators

- Uses ImageDataGenerator(rescale=1./255) for normalization.
- Image generators load and batch the image files from disk efficiently.

🧠 7. Model Creation - VGG16 Transfer Learning

- Base Model: VGG16 (excluding top classifier).
- Custom Top Layers:
 - o Flatten
 - o Dense(256, relu) → Dropout
 - \circ Dense(512, relu) \rightarrow Dropout
 - Dense(29, softmax)
- The base VGG16 layers are frozen to retain learned features.

🏋 8. Model Compilation & Training

- Loss Function: Categorical Crossentropy
- Optimizer: Adam
- Metrics: Accuracy
- Uses ModelCheckpoint to save the best model based on validation accuracy.

9. Evaluation

- Loads the saved model weights from disk.
- Uses the test data to generate predictions.
- Evaluates performance using a confusion matrix.

10. Feature Extraction & Visualization

- A secondary model is created to extract 512-dimensional features from the last dense layer.
- Uses **t-SNE** to reduce feature vectors to 2D.
- Plots 2D feature space with labels to visualize class separability.

Demonstration Plan

Use this plan as a structure for your live or recorded demo:

- 1. **Introduction**: Briefly explain ASL and the goal of the project.
- 2. **Environment Setup**: Walk through mounting Google Drive, loading Kaggle API key, and installing dependencies.
- 3. **Preprocessing**: Show how metadata is created from directory structure.
- 4. Model Building:
 - o Load VGG16
 - Add dense layers
 - Display model.summary()
- 5. **Training**: Explain loss, optimizer, and show the live training output.
- 6. **Evaluation**: Show the final test accuracy or confusion matrix.
- 7. **Visualization**: Display the t-SNE plot to explain feature embedding.