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Dataset

The Car Images Dataset dataset is used for this project.

Original Dataset

- 1. Total images: 4165
- 2. Classes: 7 ['Audi', 'Hyundai Creta', 'Mahindra Scorpio', 'Rolls Royce', 'Swift', 'Tata Safari', 'Toyota Innova']
- 3. Images per class: [1013, 338, 391, 385, 526, 547, 965]
- 4. Training samples: 3123, Testing samples: 1042

Subset Dataset

To ease training, a selection of 200 samples per category was used to form the subset dataset.

- 1. Total images: 1400
- 2. Classes: 7 ['Audi', 'Hyundai Creta', 'Mahindra Scorpio', 'Rolls Royce', 'Swift', 'Tata Safari', 'Toyota Innova']
- 3. Images per class: [200, 200, 200, 200, 200, 200]
- 4. Training samples: 1050, Testing samples: 350

The subset dataset is split into 75% training and 25% testing.

Pretrained models

Task is to train the following three pretrained CNN models by applying both Transfer-Learning and Fine-Tuning.

- 1. Model-1: ResNet101V2
- 2. Model-2: InceptionResNetV2
- 3. Model-3: DenseNet201

Part 1

Model-1: ResNet101V2

Subtask 1: Apply the following modifications to the default classifier layers of the model during Transfer-Learning:

- 1. Model-1 → Include a Batch Normalization and Dropout layer with 25% drop rate before the output layer.
- 2. Train the model for 10 epochs, and preserve the best performing TL model using the callback. Use 10% of the training dataset for validation.

```
# Import necessary libraries
import numpy as np
import tensorflow as tf
from tensorflow import keras
from tensorflow.keras import layers
from sklearn.model_selection import train_test_split
import os
from glob import glob
import matplotlib.pyplot as plt
from PIL import Image
# Set random seeds for reproducibility
np.random.seed(42)
tf.random.set_seed(42)
# Subset the dataset
dataset_path = "../input/car-images-dataset/Car_Dataset"
categories = ["Audi", "Hyundai Creta", "Mahindra Scorpio", "Rolls Royce", "Swift", "Tata Safari", "Toyota Innova"]
# Load images and labels
images, labels = [], []
```

```
for i, category in enumerate(categories):
    image_files = glob(os.path.join(dataset_path, category, "*.jpg"))
    selected_files = image_files[:200] # Select subset (200) number of samples per category
    for file in selected_files:
        img = Image.open(file).convert("RGB") # Ensure all images are RGB
        img = img.resize((224, 224)) # Resizing for model input
        images.append(np.array(img))
        labels.append(i)
# Convert to numpy arrays
images = np.array(images) / 255.0 # Normalize
labels = np.array(labels)
# Split into train and test sets
X_train, X_test, y_train, y_test = train_test_split(images, labels, test_size=0.25, stratify=labels, random_state=42)
# Check dataset details
print(f"Total images: {len(images)}")
print(f"Classes: {categories}")
print(f"Images per class: {[labels.tolist().count(i) for i in range(len(categories))]}")
print(f"Training samples: {len(X_train)}, Testing samples: {len(X_test)}")
# Build the model using ResNet101V2
base_model = keras.applications.ResNet101V2(weights='imagenet', include_top=False, input_shape=(224, 224, 3))
base_model.trainable = False # Freeze base model layers
# Add classifier layers
global_avg_pooling = layers.GlobalAveragePooling2D()(base_model.output)
batch_norm = layers.BatchNormalization()(global_avg_pooling)
dropout = layers.Dropout(0.25)(batch_norm)
output_layer = layers.Dense(len(categories), activation='softmax')(dropout)
model = keras.models.Model(inputs=base_model.input, outputs=output_layer)
# Compile the model
model.compile(optimizer='adam', loss='sparse_categorical_crossentropy', metrics=['accuracy'])
# Train the model
callbacks = [
    keras.callbacks.ModelCheckpoint("resnet101v2_tl_best.keras", save_best_only=True, monitor='val_accuracy'),
    keras.callbacks.EarlyStopping(monitor='val_accuracy', patience=3)
]
history = model.fit(
    X_train, y_train,
    validation_split=0.1,
    epochs=10,
    batch_size=32,
    callbacks=callbacks
)
# Plot accuracy
plt.plot(history.history['accuracy'], label='Training Accuracy')
plt.plot(history.history['val_accuracy'], label='Validation Accuracy')
plt.xlabel('Epochs')
plt.ylabel('Accuracy')
plt.legend()
plt.title('ResNet101V2 Transfer Learning Accuracy')
plt.show()
# Evaluate the model on the test set
model.load_weights("resnet101v2_tl_best.keras")
test_loss, test_accuracy = model.evaluate(X_test, y_test)
print(f"Test Loss: {test_loss}, Test Accuracy: {test_accuracy}")
#model.summary()
```

12/29/24, 10:44 AM Model1_TL - Colab

```
→ Total images: 1400
         Classes: ['Audi', 'Hyundai Creta', 'Mahindra Scorpio', 'Rolls Royce', 'Swift', 'Tata Safari', 'Toyota Innova']
         Images per class: [200, 200, 200, 200, 200, 200, 200]
         Training samples: 1050, Testing samples: 350
         Downloading data from <a href="https://storage.googleapis.com/tensorflow/keras-applications/resnet/resnet101v2">https://storage.googleapis.com/tensorflow/keras-applications/resnet/resnet101v2</a> weights tf dim ordering the sound of the so
         171317808/171317808
                                                                                               1s Ous/step
         Epoch 1/10
                                                                - 42s 763ms/step - accuracy: 0.2419 - loss: 2.3118 - val_accuracy: 0.5619 - val_loss: 1.1279
         30/30 -
         Epoch 2/10
         30/30 -
                                                                  5s 182ms/step - accuracy: 0.6993 - loss: 0.8094 - val_accuracy: 0.6857 - val_loss: 0.8625
         Epoch 3/10
         30/30 -
                                                                 - 5s 182ms/step - accuracy: 0.8392 - loss: 0.4903 - val_accuracy: 0.7905 - val_loss: 0.6863
         Epoch 4/10
         30/30 -
                                                                  6s 186ms/step - accuracy: 0.8859 - loss: 0.3469 - val_accuracy: 0.8095 - val_loss: 0.5815
         Epoch 5/10
                                                                  5s 184ms/step - accuracy: 0.9201 - loss: 0.2679 - val_accuracy: 0.8286 - val_loss: 0.5274
         30/30 -
         Epoch 6/10
                                                                 - 6s 185ms/step - accuracy: 0.9481 - loss: 0.2027 - val accuracy: 0.8381 - val loss: 0.4687
         30/30 -
         Epoch 7/10
         30/30 -
                                                                  6s 185ms/step - accuracy: 0.9644 - loss: 0.1601 - val_accuracy: 0.8571 - val_loss: 0.4289
         Epoch 8/10
         30/30 -
                                                                  4s 139ms/step - accuracy: 0.9822 - loss: 0.1330 - val_accuracy: 0.8476 - val_loss: 0.4118
         Epoch 9/10
                                                                - 6s 189ms/step - accuracy: 0.9799 - loss: 0.1286 - val_accuracy: 0.8667 - val_loss: 0.3908
         30/30 -
         Epoch 10/10
                                                                - 6s 192ms/step - accuracy: 0.9976 - loss: 0.1031 - val_accuracy: 0.8762 - val_loss: 0.3700
         30/30 -
                                               ResNet101V2 Transfer Learning Accuracy
                 1.0
                                         Training Accuracy
                                         Validation Accuracy
                 0.9
                 0.8
           Accuracy
                  0.7
                 0.6
                  0.5
                  0.4
                                                           2
                                                                                       4
                                                                                                                   6
                                                                                                                                               8
                                                                                         Epochs
                                                               - 8s 732ms/step - accuracy: 0.8260 - loss: 0.6041
         Test Loss: 0.622068464756012, Test Accuracy: 0.8171428442001343
```

model.save("/kaggle/working/resnet101v2_tl_best.keras")