Ex No: 5 TRANSFER LEARNING WITH CNN AND VISUALIZATION

Aim: To build a convolutional neural network with transfer learning and perform visualization

Procedure:

- 1. Download and load the dataset.
- 2. Perform analysis and preprocessing of the dataset.
- 3. Build a simple neural network model using Keras/TensorFlow.
- 4. Compile and fit the model.
- 5. Perform prediction with the test dataset.
- 6. Calculate performance metrics

Program:

import tensorflow as tf

from tensorflow.keras import layers, models

from tensorflow.keras.datasets import cifar10

import matplotlib.pyplot as plt

Load CIFAR-10 dataset

```
(x_train, y_train), (x_test, y_test) = cifar10.load_data()
```

Normalize pixel values to be between 0 and 1

```
x_{train}, x_{test} = x_{train} / 255.0, x_{test} / 255.0
```

Define the VGG16 model (adjusted for CIFAR-10)

```
def create_vgg16_model():
```

```
model = models.Sequential([
```

VGG16 architecture adapted for smaller input size (32x32)

```
layers.InputLayer(input_shape=(32, 32, 3)),
    # Convolutional Block 1
    layers.Conv2D(64, (3, 3), padding='same', activation='relu'),
    layers.Conv2D(64, (3, 3), activation='relu'),
    layers.MaxPooling2D((2, 2)),
    # Convolutional Block 2
    layers.Conv2D(128, (3, 3), padding='same', activation='relu'),
    layers.Conv2D(128, (3, 3), activation='relu'),
    layers.MaxPooling2D((2, 2)),
    # Convolutional Block 3
    layers.Conv2D(256, (3, 3), padding='same', activation='relu'),
    layers.Conv2D(256, (3, 3), activation='relu'),
    layers.MaxPooling2D((2, 2)),
    # Fully connected layers
    layers.Flatten(),
    layers.Dense(512, activation='relu'),
    layers.Dropout(0.5),
    layers.Dense(10, activation='softmax') # 10 classes for CIFAR-10
  return model
# Create the model
```

])

```
model = create_vgg16_model()
# Compile the model
model.compile(optimizer='adam',
        loss='sparse_categorical_crossentropy',
        metrics=['accuracy'])
# Train the model
history = model.fit(x_train, y_train, epochs=20, validation_data=(x_test, y_test),
batch size=64)
# Plot the accuracy
plt.figure(figsize=(12, 6))
plt.plot(history.history['accuracy'], label='Training Accuracy')
plt.plot(history.history['val_accuracy'], label='Validation Accuracy')
plt.title('Training and Validation Accuracy')
plt.xlabel('Epochs')
plt.ylabel('Accuracy')
plt.legend()
plt.show()
# Evaluate the model
test_loss, test_acc = model.evaluate(x_test, y_test, verbose=2)
print(f"Test Accuracy: {test_acc:.4f}")
Output:
 Epoch 20/20
                         – 21s 14ms/step - accuracy: 0.9615 - loss: 0.1141 - val_accuracy: 0.7742 - val_loss: 1.2498
 782/782 -
```



Result:

Thus to build a convolutional neural network with transfer learning and perform visualization was completed successfully.