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✓ Fine Tuning for model InceptionResNetV2

```
from google.colab import drive
drive.mount('/content/drive')
```

Mounted at /content/drive

```
import tensorflow as tf
```

```
from tensorflow import keras
tf.random.set_seed(42)
```

```
import numpy as np
np.random.seed(42)
```

```
import matplotlib.pyplot as plt
%matplotlib inline
```

```
from numpy import load
```

```
X_train_std = load('/content/drive/MyDrive/DLPROJECT/X_train_std.npy')
X_test_std = load('/content/drive/MyDrive/DLPROJECT/X_test_std.npy')
```

```
y_train = load('/content/drive/MyDrive/DLPROJECT/y_train.npy')
y_test = load('/content/drive/MyDrive/DLPROJECT/y_test.npy')
```


```
print("X_train_std_shape: {}".format(X_train_std.shape))
print("X_test_std_shape: {}".format(X_test_std.shape))
```

X_train_std_shape: (381, 299, 299, 3)
X_test_std_shape: (127, 299, 299, 3)

```
model1_FT = keras.models.load_model('/content/drive/MyDrive/DLPROJECT/01_InceptionResNetV2_TransferLearning_Best_Model.keras')
```

/usr/local/lib/python3.10/dist-packages/keras/src/saving/saving_lib.py:713: UserWarning: Skipping variable loading for optimizer 'rn
saveable.load_own_variables(weights_store.get(inner_path))

```
model1_FT.summary()
```

 Model: "functional"

Layer (type)	Output Shape	Param #	Connected to
input_layer_1 (InputLayer)	(None, None, None, 3)	0	-
conv2d_203 (Conv2D)	(None, None, None, 32)	864	input_layer_1[0][0]
batch_normalization_203 (BatchNormalization)	(None, None, None, 32)	96	conv2d_203[0][0]
activation_203 (Activation)	(None, None, None, 32)	0	batch_normalization_2...
conv2d_204 (Conv2D)	(None, None, None, 32)	9,216	activation_203[0][0]
batch_normalization_204 (BatchNormalization)	(None, None, None, 32)	96	conv2d_204[0][0]
activation_204 (Activation)	(None, None, None, 32)	0	batch_normalization_2...
conv2d_205 (Conv2D)	(None, None, None, 64)	18,432	activation_204[0][0]
batch_normalization_205 (BatchNormalization)	(None, None, None, 64)	192	conv2d_205[0][0]
activation_205 (Activation)	(None, None, None, 64)	0	batch_normalization_2...
max_pooling2d_4 (MaxPooling2D)	(None, None, None, 64)	0	activation_205[0][0]
conv2d_206 (Conv2D)	(None, None, None, 80)	5,120	max_pooling2d_4[0][0]
batch_normalization_206 (BatchNormalization)	(None, None, None, 80)	240	conv2d_206[0][0]
activation_206 (Activation)	(None, None, None, 80)	0	batch_normalization_2...
conv2d_207 (Conv2D)	(None, None, None, 192)	138,240	activation_206[0][0]
batch_normalization_207 (BatchNormalization)	(None, None, None, 192)	576	conv2d_207[0][0]
activation_207 (Activation)	(None, None, None, 192)	0	batch_normalization_2...
max_pooling2d_5 (MaxPooling2D)	(None, None, None, 192)	0	activation_207[0][0]
conv2d_211 (Conv2D)	(None, None, None, 64)	12,288	max_pooling2d_5[0][0]
batch_normalization_211 (BatchNormalization)	(None, None, None, 64)	192	conv2d_211[0][0]

```
print(len(model1_FT.layers))
print(int(0.25*(len(model1_FT.layers))))
```

 784
196


```
for layer in model1_FT.layers[:int(0.25*(len(model1_FT.layers)))]:
    layer.trainable = False

for layer in model1_FT.layers[int(0.25*(len(model1_FT.layers)))]:
    layer.trainable = True
    | activation_207 | (None, None, None, 40) | 0 | batch_normalization_2... |

model1_FT.compile(loss='sparse_categorical_crossentropy',
                  optimizer='adam',
                  metrics=['accuracy'])

callbacks_FineTune = [
    keras.callbacks.ModelCheckpoint("bestFT.weights.h5",
                                    monitor='val_accuracy',
                                    save_weights_only=True,
                                    save_best_only=True)
]

history_FineTune = model1_FT.fit(x = X_train_std, y = y_train, epochs=10,
                                validation_split=0.10, batch_size=16, callbacks=callbacks_FineTune)
```

 Epoch 1/10
22/22 ————— 791s 31s/step - accuracy: 0.8302 - loss: 1.0834 - val_accuracy: 0.3333 - val_loss: 12770.4150

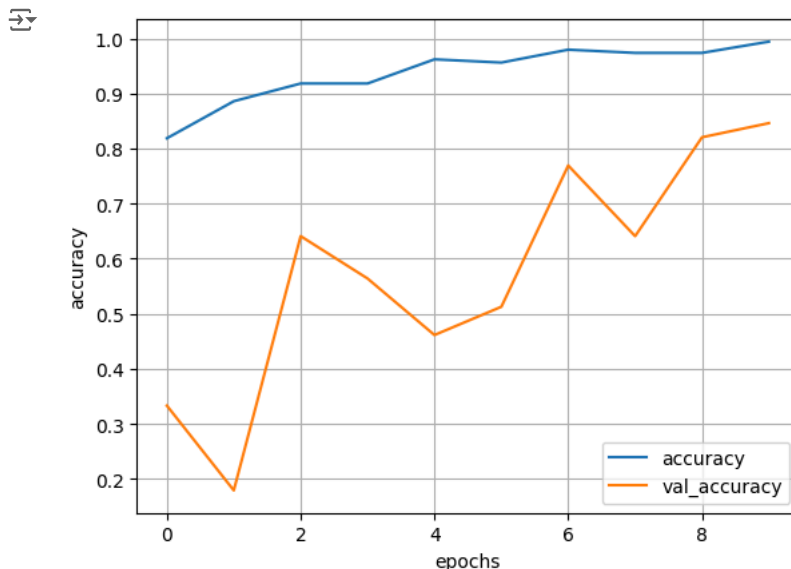
```
Epoch 2/10
22/22 ----- 706s 29s/step - accuracy: 0.8639 - loss: 0.4402 - val_accuracy: 0.1795 - val_loss: 185.3277
Epoch 3/10
22/22 ----- 699s 30s/step - accuracy: 0.9162 - loss: 0.3402 - val_accuracy: 0.6410 - val_loss: 0.9667
Epoch 4/10
22/22 ----- 665s 30s/step - accuracy: 0.9040 - loss: 0.2349 - val_accuracy: 0.5641 - val_loss: 564.7798
Epoch 5/10
22/22 ----- 664s 30s/step - accuracy: 0.9728 - loss: 0.1586 - val_accuracy: 0.4615 - val_loss: 5042.0513
Epoch 6/10
22/22 ----- 658s 29s/step - accuracy: 0.9606 - loss: 0.1158 - val_accuracy: 0.5128 - val_loss: 0.9618
Epoch 7/10
22/22 ----- 714s 31s/step - accuracy: 0.9829 - loss: 0.0610 - val_accuracy: 0.7692 - val_loss: 0.6547
Epoch 8/10
22/22 ----- 676s 30s/step - accuracy: 0.9630 - loss: 0.2243 - val_accuracy: 0.6410 - val_loss: 10.8359
Epoch 9/10
22/22 ----- 680s 31s/step - accuracy: 0.9728 - loss: 0.0856 - val_accuracy: 0.8205 - val_loss: 0.5760
Epoch 10/10
22/22 ----- 673s 31s/step - accuracy: 0.9979 - loss: 0.0221 - val_accuracy: 0.8462 - val_loss: 0.6175
```

```
keys = ['accuracy', 'val_accuracy']
progress = {k:v for k,v in history_FineTune.history.items() if k in keys}
```

```
import pandas as pd
pd.DataFrame(progress).plot()
```

```
plt.xlabel("epochs")
plt.ylabel("accuracy")
```

```
plt.grid(True)
plt.show()
```



```
testLoss_FineTune, testAccuracy_FineTune = model1_FT.evaluate(x = X_test_std, y = y_test)
print("Test-loss: %f, Test-accuracy: %f" % (testLoss_FineTune, testAccuracy_FineTune))
```

```
4/4 ----- 78s 19s/step - accuracy: 0.7190 - loss: 0.8304
Test-loss: 0.823912, Test-accuracy: 0.732283
```

```
model1_FT.load_weights("bestFT.weights.h5")
```

```
testLoss_FineTune, testAccuracy_FineTune = model1_FT.evaluate(x = X_test_std, y = y_test)
```

```
print("Test-loss: %f, Test-accuracy: %f" % (testLoss_FineTune, testAccuracy_FineTune))
```

```
4/4 ----- 80s 20s/step - accuracy: 0.7190 - loss: 0.8304
Test-loss: 0.823912, Test-accuracy: 0.732283
```

```
y_proba = model1_FT.predict(X_test_std)
y_predict = np.argmax(y_proba, axis=-1)
print(y_predict)
```

```
4/4 ----- 94s 22s/step
[3 3 0 0 2 1 3 3 1 1 3 0 1 3 1 2 3 2 0 1 1 1 0 3 0 0 3 1 1 1 1 3 3 3 1 1 1
 3 3 2 2 2 1 3 3 0 2 0 1 0 1 2 3 3 1 1 3 0 1 1 3 2 0 0 0 0 1 1 0 1 0 3 3 3
 3 3 0 3 1 3 2 1 3 2 3 3 3 3 1 3 0 1 0 1 2 0 3 0 3 3 0 3 1 1 3 2 3 1 1 0 2
 0 3 0 3 1 0 3 3 3 1 3 3 3 3 3 3]
```

```

from sklearn.metrics import confusion_matrix
confusion_matrix(y_true = y_test, y_pred = y_predict)

array([[16,  4,  0,  6],
       [ 7, 23,  0,  3],
       [ 2,  4, 14,  1],
       [ 2,  4,  0, 40]])

cm = confusion_matrix(y_true = y_test, y_pred = y_predict)

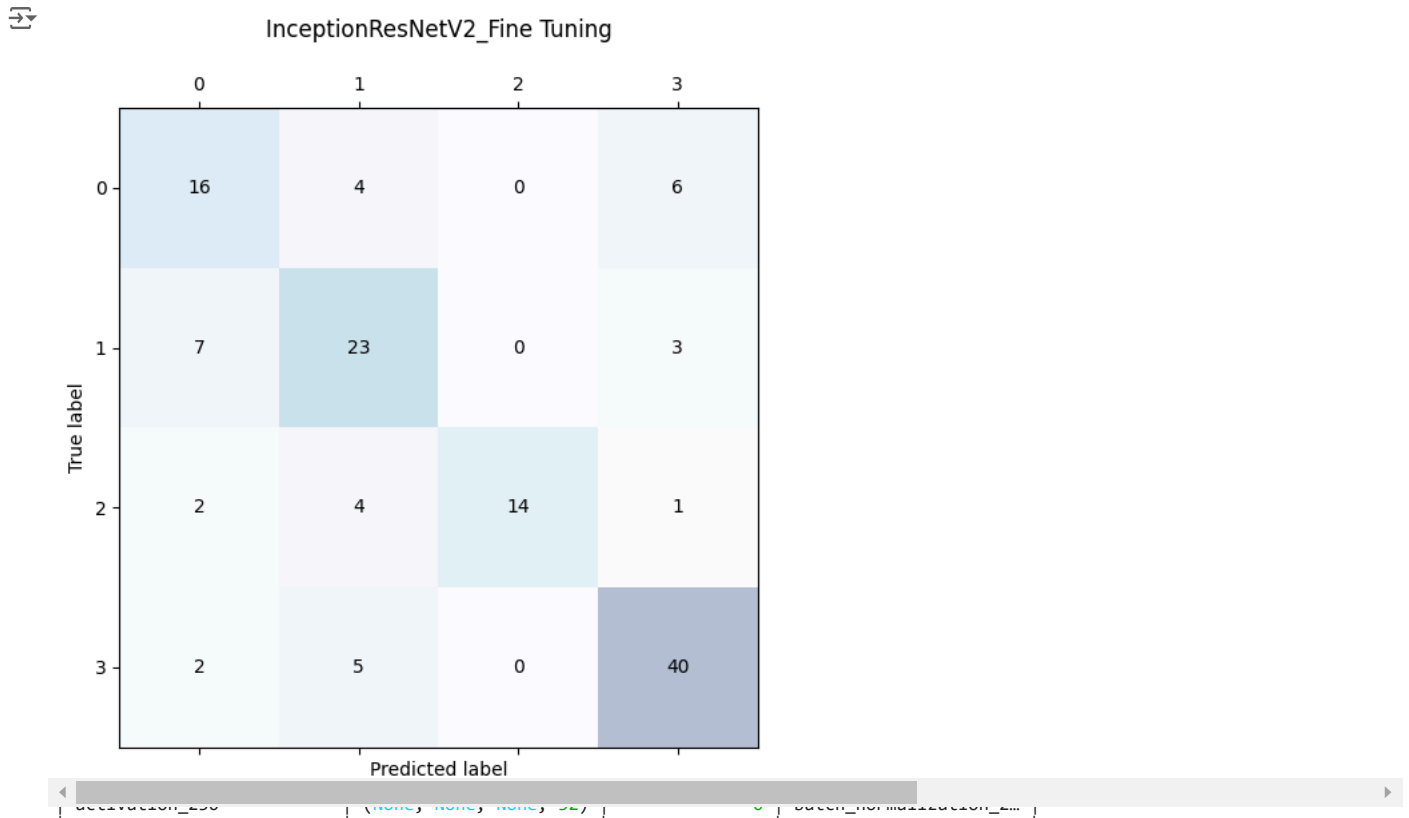
fig, ax = plt.subplots(figsize=(6, 6))
ax.matshow(cm, cmap=plt.cm.Blues, alpha=0.3)

for i in range(cm.shape[0]):
    for j in range(cm.shape[1]):
        ax.text(x=j, y=i, s=cm[i, j], va='center', ha='center')

ax.title.set_text('InceptionResNetV2_Fine Tuning')
plt.xlabel('Predicted label')
plt.ylabel('True label')

plt.tight_layout()
plt.savefig("ConfusionMatrix.png", dpi=300, format='png', pad_inches=0.3)
plt.show()

```



```

from sklearn.metrics import precision_score, recall_score, f1_score

pScore = precision_score(y_true= y_test, y_pred = y_predict, average = 'weighted')
print("Precision: ", pScore)

rScore = recall_score(y_true= y_test, y_pred = y_predict, average = 'weighted')
print("Recall: ", rScore)

fScore = f1_score(y_true= y_test, y_pred = y_predict, average = 'weighted')
print("F1-score: ", fScore)

print("\n\n")

Precision:  0.7487459900845728

```