

Model1_FT - Fine Tuning using InceptionNetV2

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

Mounted at /content/drive

```
import tensorflow as tf
print(tf.__version__)

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

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

import matplotlib.pyplot as plt
%matplotlib inline

import glob
import PIL
from PIL import Image
```

2.15.0

Loading the preprocessed dataset

```
# load numpy array from npy file
from numpy import load

X_train_std = load('/content/drive/MyDrive/Models/X_train_std_model1.npy')
X_test_std = load('/content/drive/MyDrive/Models/X_test_std_model1.npy')

y_train = load('/content/drive/MyDrive/Models/y_train_model1.npy')
y_test = load('/content/drive/MyDrive/Models/y_test_model1.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: (373, 299, 299, 3)
X_test_std_shape: (125, 299, 299, 3)
```

Loading the Transfer-learning Model

```
model1_FT = keras.models.load_model('/content/drive/MyDrive/Models/Model1_TL.h5')
model1_FT.summary()
```

Model: "model"

Layer (type)	Output Shape	Param #	Connected to
input_1 (InputLayer)	[(None, 299, 299, 3)]	0	[]
conv2d (Conv2D)	(None, 149, 149, 32)	864	['input_1[0][0]']
batch_normalization (Batch Normalization)	(None, 149, 149, 32)	96	['conv2d[0][0]']
activation (Activation)	(None, 149, 149, 32)	0	['batch_normalization[0][0]']
conv2d_1 (Conv2D)	(None, 147, 147, 32)	9216	['activation[0][0]']
batch_normalization_1 (Batch Normalization)	(None, 147, 147, 32)	96	['conv2d_1[0][0]']
activation_1 (Activation)	(None, 147, 147, 32)	0	['batch_normalization_1[0][0]']
conv2d_2 (Conv2D)	(None, 147, 147, 64)	18432	['activation_1[0][0]']
batch_normalization_2 (Batch Normalization)	(None, 147, 147, 64)	192	['conv2d_2[0][0]']

activation_2 (Activation)	(None, 147, 147, 64)	0	['batch_normalization_2[0][0]']
max_pooling2d (MaxPooling2D)	(None, 73, 73, 64)	0	['activation_2[0][0]']
conv2d_3 (Conv2D)	(None, 73, 73, 80)	5120	['max_pooling2d[0][0]']
batch_normalization_3 (Batch Normalization)	(None, 73, 73, 80)	240	['conv2d_3[0][0]']
activation_3 (Activation)	(None, 73, 73, 80)	0	['batch_normalization_3[0][0]']
conv2d_4 (Conv2D)	(None, 71, 71, 192)	138240	['activation_3[0][0]']
batch_normalization_4 (Batch Normalization)	(None, 71, 71, 192)	576	['conv2d_4[0][0]']
activation_4 (Activation)	(None, 71, 71, 192)	0	['batch_normalization_4[0][0]']
max_pooling2d_1 (MaxPooling2D)	(None, 35, 35, 192)	0	['activation_4[0][0]']
conv2d_8 (Conv2D)	(None, 35, 35, 64)	12288	['max_pooling2d_1[0][0]']
batch_normalization_8 (Batch Normalization)	(None, 35, 35, 64)	192	['conv2d_8[0][0]']
activation_8 (Activation)	(None, 35, 35, 64)	0	['batch_normalization_8[0][0]']

✧ Modifyng and Fine tuning the layers to be trained

```
total_layers = len(model1_FT.layers)
split_index = int(0.25 * total_layers)

for layer in model1_FT.layers[:split_index]:
    layer.trainable = False

for layer in model1_FT.layers[split_index:]:
    layer.trainable = True
```

✧ Compiling and Training the Model

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

callbacks_FineTune = [
    keras.callbacks.ModelCheckpoint("bestFT1.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=0.1, batch_size=16, callbacks=callbacks_FineTune)
```

```
Epoch 1/10
21/21 [=====] - 115s 1s/step - loss: 2.0020 - accuracy: 0.5403 - val_loss: 91.9486 - val_accuracy: 0.2105
Epoch 2/10
21/21 [=====] - 11s 515ms/step - loss: 0.8413 - accuracy: 0.7134 - val_loss: 448.7038 - val_accuracy: 0.2875
Epoch 3/10
21/21 [=====] - 10s 459ms/step - loss: 0.6594 - accuracy: 0.7731 - val_loss: 58.6408 - val_accuracy: 0.2105
Epoch 4/10
21/21 [=====] - 11s 515ms/step - loss: 0.5184 - accuracy: 0.8328 - val_loss: 94.6049 - val_accuracy: 0.3421
Epoch 5/10
21/21 [=====] - 9s 449ms/step - loss: 0.5477 - accuracy: 0.8269 - val_loss: 10469.3779 - val_accuracy: 0.2875
Epoch 6/10
21/21 [=====] - 9s 450ms/step - loss: 0.3422 - accuracy: 0.8746 - val_loss: 2586.4138 - val_accuracy: 0.3125
Epoch 7/10
21/21 [=====] - 11s 517ms/step - loss: 0.2396 - accuracy: 0.8985 - val_loss: 3.8631 - val_accuracy: 0.3947
Epoch 8/10
21/21 [=====] - 11s 537ms/step - loss: 0.2028 - accuracy: 0.9254 - val_loss: 2.0081 - val_accuracy: 0.4211
Epoch 9/10
21/21 [=====] - 11s 517ms/step - loss: 0.1190 - accuracy: 0.9522 - val_loss: 4.1281 - val_accuracy: 0.5263
Epoch 10/10
21/21 [=====] - 10s 459ms/step - loss: 0.1709 - accuracy: 0.9254 - val_loss: 16.6901 - val_accuracy: 0.4479
```

```

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()

```

✓ Evaluating the Model with Best weights

```

model1_FT.load_weights("bestFT1.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 [=====] - 12s 2s/step - loss: 1.7353 - accuracy: 0.6320
Test-loss: 1.735336, Test-accuracy: 0.632000

```

✓ Checking Model Performance

```

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

```

```

from sklearn.metrics import confusion_matrix
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('CNN\n')
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\n")

```

✓ Saving the Fine-Tuned Model

```

model1_FT.save('/content/drive/MyDrive/Models/model1_FT.h5')

```

```

/usr/local/lib/python3.10/dist-packages/keras/src/engine/training.py:3103: UserWarning: You are saving your model as an HDF5 file via
saving_api.save_model(

```

