# Assignment 3 - Build CNN Model for Classification Of Flowers

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```
import splitfolders
import numpy as np
import tensorflow as tf
from tensorflow.keras.preprocessing.image import
ImageDataGenerator from tensorflow.keras.preprocessing import
image from tensorflow.keras import layers
from tensorflow.keras.models import Sequential
from tensorflow.keras.models import load_model
from tensorflow.keras.layers import Dense,Convolution2D,MaxPooling2D,Flatten from
tensorflow.keras.applications.resnet50 import preprocess_input, decode_predicti
from tensorflow.keras.preprocessing import image import matplotlib.pyplot as plt
```

```
train_datagen = ImageDataGenerator(rescale=1./255,zoom_range=0.2,horizontal_flip=Tru
```

#### 2. Image Augmentation

In [2]:

```
In [3]:
         test datagen = ImageDataGenerator(rescale=1./255)
In [4]:
         input folder = '.\Flowers-Dataset\\flowers'
In [5]:
         splitfolders_ratio(input_folder,output="flowers",ratio=(.8,0,.2),group_prefix=None)
        Copying files: 4317 files [00:03, 1292.11 files/s]
In [6]:
         x_train=train_datagen_flow_from_directory(r".\flowers\train",target_size=(64,64),cla
        Found 3452 images belonging to 5 classes.
In [7]:
         x_test=test_datagen_flow_from_directory(r".\flowers\test",target_size=(64,64),class_
        Found 865 images belonging to 5 classes.
In [8]:
 x train_class indices
        {'daisy': 0, 'dandelion': 1, 'rose': 2, 'sunflower': 3, 'tulip': 4}
Out[8]:
```

#### 3. Create Model

In [9]:

model=Sequentia ()

## 4. Add Layers

#### 4.1. Convolution Layer

In [10]:

```
model_add(Convolution2D(32,(3,3),input_shape=(64,64,3),activation='relu'))
```

#### 4.2. MaxPooling Layer

In [11]:

```
model_add(MaxPooling2D(pool_size=(2,2)))
```

#### 4.3. Flatten Layer

In [12]:

```
model_add(Flatten())
```

#### 4.4. Dense Layer

In [14]:

model\_summary()

Model: "sequential"

Non-trainable params: 0

Layer (type)	Output Shape	Param #
= conv2d (Conv2D) max_pooling2d (MaxPooling2D)	(None, 62, 62, 32) (None, 31, 31, 32)	896 0
flatten (Flatten) dense (Dense) dense_1 (Dense)	(None, 30752) (None, 300) (None, 150)	0 9225900 45150
======================================		========

```
model_add(Dense(5,activation='softmax*))

model_summary()
```

#### 4.5. Output Layer

In [15]:

In [16]:

Model: "sequential"

Layer (type)	Output Shape	Param #	
=======================================			====
conv2d (Conv2D)	(None, 62, 62, 32)	896	
max_pooling2d (MaxPooling2D	(None, 31, 31, 32)	0	)
flatten (Flatten)	(None, 30752)	0	
dense (Dense)	(None, 300)	9225900	
dense 1 (Dense)	(None, 150)	45150	
dense_2 (Dense)	(None, 5)	755	
=======================================	=======================================		====

Total params: 9,272,701 Trainable params: 9,272,701 Non-trainable params: 0

model\_compile(loss='categorical\_crossentropy",optimizer='adam",metrics=['accuracy'])
len(x\_train)

## 5. Compile The Model

In [17]:

144

Out[17]:

```
epo=20
history =
model_fit(x_train, steps_per_epoch=len(x_train), validation_data=x_test, vali
```

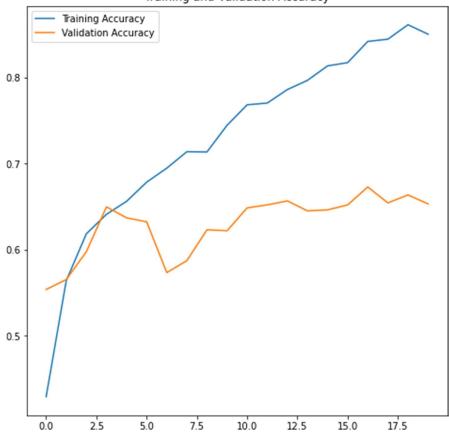
#### 6. Fit The Model

In [18]:

```
0.4293 - val loss: 1.1148 - val accuracy: 0.5538
Epoch 2/20
accuracy:
0.5640 - val_loss: 1.0807 - val_accuracy: 0.5653
Epoch 3/20
accuracy:
0.6185 - val loss: 1.0689 - val accuracy: 0.5977
Epoch 4/20
accuracy:
0.6411 - val_loss: 0.9561 - val_accuracy: 0.6497
Epoch 5/20
accuracy:
0.6561 - val_loss: 0.9766 - val_accuracy: 0.6370
Epoch 6/20
accuracy:
0.6784 - val loss: 1.0373 - val accuracy: 0.6324
Epoch 7/20
accuracy:
0.6947 - val loss: 1.1446 - val accuracy: 0.5734
Epoch 8/20
accuracy:
0.7138 - val loss: 1.1979 - val accuracy: 0.5873
Epoch 9/20
accuracy:
0.7135 - val loss: 1.0924 - val accuracy: 0.6231
Epoch 10/20
accuracy:
0.7445 - val loss: 1.1218 - val accuracy: 0.6220
Epoch 11/20
accuracy:
0.7683 - val_loss: 1.0576 - val_accuracy: 0.6486
Epoch 12/20
accuracy:
0.7703 - val loss: 1.0454 - val accuracy: 0.6520
Epoch 13/20
accuracy:
0.7859 - val loss: 1.0735 - val accuracy: 0.6566
Epoch 14/20
accuracy:
0.7966 - val_loss: 1.1083 - val_accuracy: 0.6451
Epoch 15/20
accuracy:
0.8134 - val_loss: 1.0815 - val_accuracy: 0.6462
Epoch 16/20
accuracy:
```

```
0.8172 - val loss: 1.0991 - val accuracy: 0.6520
      Epoch 17/20
      accuracy:
      0.8418 - val_loss: 1.2605 - val_accuracy: 0.6728
      Epoch 18/20
      accuracy:
      0.8444 - val loss: 1.1316 - val accuracy: 0.6543
      Epoch 19/20
      accuracy:
      0.8612 - val_loss: 1.1264 - val_accuracy: 0.6636
      Epoch 20/20
      accuracy:
      0.8502 - val_loss: 1.1911 - val_accuracy: 0.6532
In [19]:
      epochs_range = range(epo)
      plt_figure(figsize=(8, 8))
      plt_plot(epochs range, history_history['accuracy'], label='Training Accuracy')
      plt_plot(epochs_range, history_history['val_accuracy"], label='Validation
      Accuracy') plt.legend()
      plt.title('Training and Validation Accuracy')
      plt_show()
```

#### Training and Validation Accuracy



```
plt_figure(figsize=(8, 8))
plt_plot(epochs_range, history_history["loss"], label="Training Loss")
plt_plot(epochs_range, history_history["val_loss"], label="Validation Loss")
plt.legend()
plt_title('Training and Validation Loss')
plt_show()
```



#### 7. Save the Model

2.5

5.0

7.5

In [21]:

0.4

0.0

```
model_save('flowers.h5")
```

10.0

12.5

15.0

17.5

```
img=image_load_img(r".\flowers\test\daisy\3706420943_66f3214862_n.jpg",target_size=(
x=image_img_to_array(img) x=np_expand_dims(x,axis=0)
y=np_argmax(model_predict(x),axis=1) x_train_class_indices
index=['daisy','dandellion','rose','sunflower','tulip'] index[y[0]]
```

## 8. Test the Model

In [22]:

Out[22]:1/1 [=======] - 0s 77ms/step 'daisy'

```
Downloading data from
In [23]:
         https://storage.googleapis.com/download.tensorflow.org/example_images/592px-
         Red_sunflower.jpg
img_url = 17948/117948 [==========] - 0s Ous/step
"https://storage.googleapis.com/attps://storage.googleapis.com/tensorflow/keras-
img_path applications/ueaset/eenetsp-weightsutweim_orderingnts_kerinels.h5
         102967424/102967424 [=========
img = ima \sqrt[4]{\frac{10000 - img(img-path, -target-size=(224)}{24})} \sqrt[8]{68} ms/step
img arravDewnloading datarframing)
 img_batchttps://storagedgaggleapia.opm/slownload.tensorflow.org/data/im
         agenet_class_index.json
img_prep3536365363[eprocess_input(img_batch) ======] - 0s 0us/step
 model = [('n11939491', 'daisy', 0.5775759), ('n02206856', 'bee', 0.24938338),
 tf_keras (a)39220624s_restet50_R4484930)]
 prediction = model_predict(img preprocessed)
  print(decode_predictions(prediction, top=3)[0])
 score = tf_nn_softmax(prediction[0])
```