

# galaxy-classifier-Inceptionv3

February 11, 2021

## 0.0.1 Classification of Galaxies Based on their Morphology using Transfer Learning

```
[4]: #Importing libraries
%matplotlib inline
import matplotlib.pyplot as plt
import numpy as np
import pandas as pd
import os, random, shutil
import tensorflow as tf
import seaborn
```

```
[59]: # paths to the archives
base_path = r'../input/galaxy-zoo-the-galaxy-challenge/'
training_solutions = os.path.join(base_path, 'training_solutions_rev1.zip')
training_images = os.path.join(base_path, 'images_training_rev1.zip')
```

```
[60]: df = pd.read_csv(training_solutions, compression="zip")
cols = df.columns
new = list(map(lambda s: s.replace('Class','Q'), cols))
df.columns = new
df.head()
```

```
[60]:
```

	GalaxyID	Q1.1	Q1.2	Q1.3	Q2.1	Q2.2	Q3.1	\
0	100008	0.383147	0.616853	0.000000	0.000000	0.616853	0.038452	
1	100023	0.327001	0.663777	0.009222	0.031178	0.632599	0.467370	
2	100053	0.765717	0.177352	0.056931	0.000000	0.177352	0.000000	
3	100078	0.693377	0.238564	0.068059	0.000000	0.238564	0.109493	
4	100090	0.933839	0.000000	0.066161	0.000000	0.000000	0.000000	

		Q3.2	Q4.1	Q4.2	...	Q9.3	Q10.1	Q10.2	Q10.3	\
0	0.578401	0.418398	0.198455	...	0.000000	0.279952	0.138445	0.000000		
1	0.165229	0.591328	0.041271	...	0.018764	0.000000	0.131378	0.459950		
2	0.177352	0.000000	0.177352	...	0.000000	0.000000	0.000000	0.000000		
3	0.129071	0.189098	0.049466	...	0.000000	0.094549	0.000000	0.094549		
4	0.000000	0.000000	0.000000	...	0.000000	0.000000	0.000000	0.000000		

		Q11.1	Q11.2	Q11.3	Q11.4	Q11.5	Q11.6
0	0.000000	0.092886	0.0	0.0	0.0	0.325512	

1	0.000000	0.591328	0.0	0.0	0.0	0.000000
2	0.000000	0.000000	0.0	0.0	0.0	0.000000
3	0.189098	0.000000	0.0	0.0	0.0	0.000000
4	0.000000	0.000000	0.0	0.0	0.0	0.000000

[5 rows x 38 columns]

## 0.0.2 Class Segregation

The columns from the above dataset signifies the actual classes, considering both computations and my novice knowledge, I'm restricting the classes to three types which are root categories anyway

```
[91]: ellipticals = df[(df['Q1.1']>0.7) & (df['Q7.1']>0.4)]['GalaxyID'].tolist()
      lenticulars = df[(df['Q1.1']>0.7) & (df['Q7.2']>0.4)]['GalaxyID'].tolist()
      spirals = df[(df['Q1.2']>0.7) & (df['Q2.1']>0.4)]['GalaxyID'].tolist()

      print('Total number of elliptical examples: ', len(ellipticals))
      print('Total number of lenticular examples: ', len(lenticulars))
      print('Total number of spiral examples: ', len(spirals))
```

Total number of elliptical examples: 7311

Total number of lenticular examples: 6625

Total number of spiral examples: 4635

Taking only 500 random samples from each category

```
[92]: def return500(category):
      category = np.array(category)
      rn_500 = sorted(list(np.random.randint(1,4500,500)))
      return list(category[rn_500])

      ellipticals = return500(ellipticals)
      lenticulars = return500(lenticulars)
      spirals = return500(spirals)
```

```
[93]: len(spirals)
```

```
[93]: 500
```

## 0.0.3 Test and Validation Splits

```
[98]: # This subroutine is lifted from a kaggle notebook
      def _proc_images(src, dst, label, arr, percent):
          train_dir = os.path.join(dst, 'train')
          val_dir = os.path.join(dst, 'validation')
```

```

train_dest = os.path.join(train_dir, label)
val_dest   = os.path.join(val_dir, label)

if not os.path.exists(train_dest):
    os.makedirs(train_dest)

if not os.path.exists(val_dest):
    os.makedirs(val_dest)

random.shuffle(arr)

idx = int(len(arr)*percent)
for i in arr[0:idx]:
    shutil.copyfile(os.path.join(src, str(i)+'.jpg'), os.path.
→join(train_dest, str(i)+'.jpg'))
    for i in arr[idx:]:
        shutil.copyfile(os.path.join(src, str(i)+'.jpg'), os.path.
→join(val_dest, str(i)+'.jpg'))

print(label, 'done!')

```

```

[99]: training_solutions = os.path.join(base_path, 'training_solutions_rev1')
      training_images    = os.path.join(base_path, 'images_training_rev1')

```

```

[100]: my_data = '/Users/Shared/Relocated/Security/Codes/Deep_Learning/input/
→galaxy-zoo-clean/data'

```

```

[101]: _proc_images(training_images, '../input/galaxy-zoo-clean/data/', 'elliptical',
→ellipticals, 0.90)
_proc_images(training_images, '../input/galaxy-zoo-clean/data/', 'lenticular',
→lenticulars, 0.90)
_proc_images(training_images, '../input/galaxy-zoo-clean/data/', 'spiral',
→spirals, 0.90)

print('Elliptical:', len(os.listdir(os.path.join(my_data, 'train',
→'elliptical'))))
print('Total train lenticular:', len(os.listdir(os.path.join(my_data, 'train',
→'lenticular'))))
print('Total train spiral:', len(os.listdir(os.path.join(my_data, 'train',
→'spiral'))))

print('Total validation elliptical:', len(os.listdir(os.path.join(my_data,
→'validation', 'elliptical'))))

```

```
print('Total validation lenticular:', len(os.listdir(os.path.join(my_data,
↳ 'validation', 'lenticular'))))
print('Total validation spiral:', len(os.listdir(os.path.join(my_data,
↳ 'validation', 'spiral'))))
```

```
elliptical done!
lenticular done!
spiral done!
Elliptical: 425
Total train lenticular: 431
Total train spiral: 433
Total validation elliptical: 49
Total validation lenticular: 50
Total validation spiral: 50
```

```
[5]: train_dir = '../input/galaxy-zoo-clean/data/train'
validation_dir = '../input/galaxy-zoo-clean/data/validation'
```

```
[30]: total_train = 0
for c in ['elliptical', 'lenticular', 'spiral']:
    total_train += len(os.listdir(os.path.join(train_dir, c)))
print('Total train:', total_train)

total_validation = 0
for c in ['elliptical', 'lenticular', 'spiral']:
    total_validation += len(os.listdir(os.path.join(validation_dir, c)))
print('Total validation:', total_validation)
```

```
Total train: 1289
Total validation: 149
```

#### 0.0.4 Data Overview

```
[31]: from PIL import Image
import random
def return_imgarr(train_dir, label):
    img_arr = []
    paths = os.listdir(os.path.join(train_dir, label))

    for num, file in enumerate(random.choices(paths, k=3)):
        pic = Image.open(os.path.join(train_dir, label, file))
        img_arr.append(np.array(pic))

    return img_arr
```

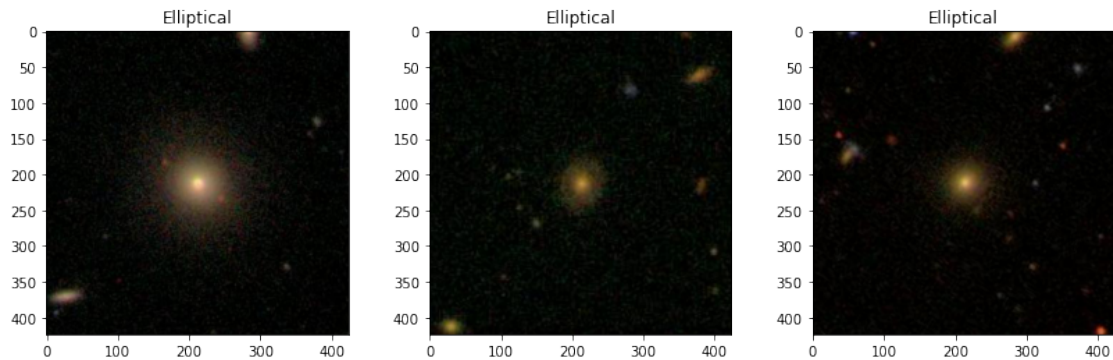
```
[32]: elliptical_arr = return_imgarr(train_dir, label='elliptical')
lenticular_arr = return_imgarr(train_dir, label='lenticular')
```

```
spiral_arr = return_imgarr(train_dir, label='spiral')
```

### 0.0.5 *Three Random Samples of Elliptical Galaxies*

```
[33]: fig, ax = plt.subplots(1, 3, figsize=(14, 4))

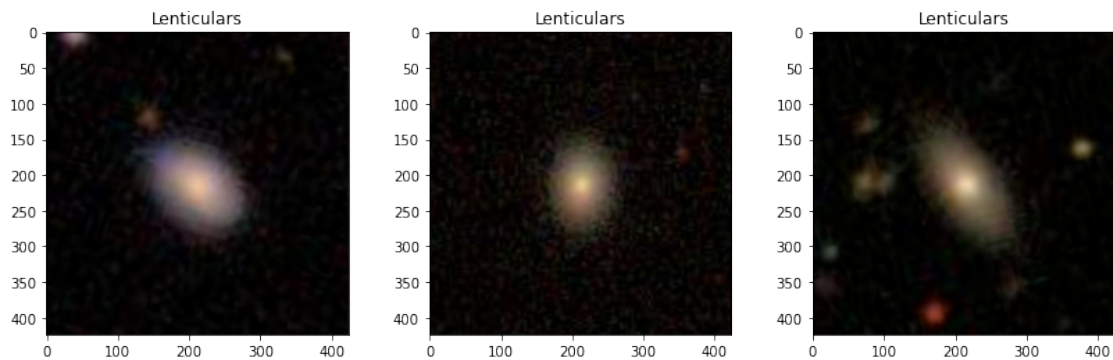
for i, ax in enumerate(ax.ravel()):
    ax.set_title("Elliptical".format(i))
    ax.imshow(elliptical_arr[i])
```



### 0.0.6 *Three Random Samples of Lenticular Galaxies*

```
[34]: fig, ax = plt.subplots(1, 3, figsize=(14, 4))

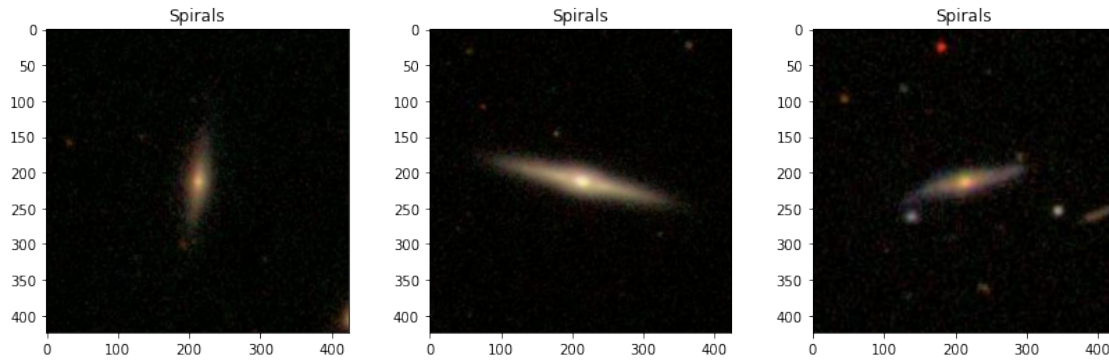
for i, ax in enumerate(ax.ravel()):
    ax.set_title("Lenticulars".format(i))
    ax.imshow(lenticular_arr[i])
```



### 0.0.7 Three Random Samples of Spiral Galaxies

```
[35]: fig, ax = plt.subplots(1, 3, figsize=(14, 4))

for i, ax in enumerate(ax.ravel()):
    ax.set_title("Spirals".format(i))
    ax.imshow(spiral_arr[i])
```



### 0.0.8 Image Generators

```
[7]: BS = 64 # batch_size
train_datagen = tf.keras.preprocessing.image.ImageDataGenerator(
    rescale=1.0/255,
    rotation_range=25,
    width_shift_range=.15,
    height_shift_range=.15,
    horizontal_flip=True,
    zoom_range=0.2)
validation_datagen = tf.keras.preprocessing.image.ImageDataGenerator(rescale=1.
    ↪0/255.)

train_generator = train_datagen.flow_from_directory(train_dir,
    target_size=(180,180),
    batch_size=BS,
    shuffle=True,
    class_mode='categorical')
validation_generator = train_datagen.flow_from_directory(validation_dir,
    target_size=(180,180),
    batch_size=BS,
    shuffle=True,
    ↪class_mode='categorical')
```

Found 1289 images belonging to 3 classes.

Found 149 images belonging to 3 classes.

```
[8]: # Required Layer artefacts
from tensorflow.keras.applications import ResNet50
from tensorflow.keras.callbacks import EarlyStopping
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import
    ↳Flatten,Dense,BatchNormalization,Activation,Dropout
from tensorflow.keras.optimizers import Adam
from tensorflow.keras.models import Model
```

```
[9]: train_generator.image_shape
```

```
[9]: (180, 180, 3)
```

```
[83]: from tensorflow.keras.applications import InceptionV3
```

```
[109]: inception = InceptionV3(include_top=False, weights='imagenet',
    ↳input_shape=train_generator.image_shape)
```

```
[110]: # Freezing the Weights
for layer in inception.layers:
    layer.trainable = False
```

```
[111]: inception.summary()
```

Model: "inception\_v3"

```
-----
Layer (type)                Output Shape          Param #   Connected to
=====
input_6 (InputLayer)        [(None, 180, 180, 3) 0
-----
conv2d_282 (Conv2D)         (None, 89, 89, 32)   864       input_6[0][0]
-----
batch_normalization_282 (BatchN (None, 89, 89, 32)   96
conv2d_282[0][0]
-----
activation_282 (Activation) (None, 89, 89, 32)   0
batch_normalization_282[0][0]
-----
conv2d_283 (Conv2D)         (None, 87, 87, 32)   9216
activation_282[0][0]
```

```

-----
-----
batch_normalization_283 (BatchN (None, 87, 87, 32) 96
conv2d_283[0][0]
-----
-----
activation_283 (Activation) (None, 87, 87, 32) 0
batch_normalization_283[0][0]
-----
-----
conv2d_284 (Conv2D) (None, 87, 87, 64) 18432
activation_283[0][0]
-----
-----
batch_normalization_284 (BatchN (None, 87, 87, 64) 192
conv2d_284[0][0]
-----
-----
activation_284 (Activation) (None, 87, 87, 64) 0
batch_normalization_284[0][0]
-----
-----
max_pooling2d_12 (MaxPooling2D) (None, 43, 43, 64) 0
activation_284[0][0]
-----
-----
conv2d_285 (Conv2D) (None, 43, 43, 80) 5120
max_pooling2d_12[0][0]
-----
-----
batch_normalization_285 (BatchN (None, 43, 43, 80) 240
conv2d_285[0][0]
-----
-----
activation_285 (Activation) (None, 43, 43, 80) 0
batch_normalization_285[0][0]
-----
-----
conv2d_286 (Conv2D) (None, 41, 41, 192) 138240
activation_285[0][0]
-----
-----
batch_normalization_286 (BatchN (None, 41, 41, 192) 576
conv2d_286[0][0]
-----
-----
activation_286 (Activation) (None, 41, 41, 192) 0
batch_normalization_286[0][0]

```



```

-----
-----
max_pooling2d_13 (MaxPooling2D) (None, 20, 20, 192) 0
activation_286[0][0]
-----
-----
conv2d_290 (Conv2D) (None, 20, 20, 64) 12288
max_pooling2d_13[0][0]
-----
-----
batch_normalization_290 (BatchN (None, 20, 20, 64) 192
conv2d_290[0][0]
-----
-----
activation_290 (Activation) (None, 20, 20, 64) 0
batch_normalization_290[0][0]
-----
-----
conv2d_288 (Conv2D) (None, 20, 20, 48) 9216
max_pooling2d_13[0][0]
-----
-----
conv2d_291 (Conv2D) (None, 20, 20, 96) 55296
activation_290[0][0]
-----
-----
batch_normalization_288 (BatchN (None, 20, 20, 48) 144
conv2d_288[0][0]
-----
-----
batch_normalization_291 (BatchN (None, 20, 20, 96) 288
conv2d_291[0][0]
-----
-----
activation_288 (Activation) (None, 20, 20, 48) 0
batch_normalization_288[0][0]
-----
-----
activation_291 (Activation) (None, 20, 20, 96) 0
batch_normalization_291[0][0]
-----
-----
average_pooling2d_27 (AveragePo (None, 20, 20, 192) 0
max_pooling2d_13[0][0]
-----
-----
conv2d_287 (Conv2D) (None, 20, 20, 64) 12288
max_pooling2d_13[0][0]

```

```

-----
conv2d_289 (Conv2D)          (None, 20, 20, 64)    76800
activation_288[0][0]

-----

conv2d_292 (Conv2D)          (None, 20, 20, 96)    82944
activation_291[0][0]

-----

conv2d_293 (Conv2D)          (None, 20, 20, 32)    6144
average_pooling2d_27[0][0]

-----

batch_normalization_287 (BatchN (None, 20, 20, 64)    192
conv2d_287[0][0]

-----

batch_normalization_289 (BatchN (None, 20, 20, 64)    192
conv2d_289[0][0]

-----

batch_normalization_292 (BatchN (None, 20, 20, 96)    288
conv2d_292[0][0]

-----

batch_normalization_293 (BatchN (None, 20, 20, 32)    96
conv2d_293[0][0]

-----

activation_287 (Activation)    (None, 20, 20, 64)    0
batch_normalization_287[0][0]

-----

activation_289 (Activation)    (None, 20, 20, 64)    0
batch_normalization_289[0][0]

-----

activation_292 (Activation)    (None, 20, 20, 96)    0
batch_normalization_292[0][0]

-----

activation_293 (Activation)    (None, 20, 20, 32)    0
batch_normalization_293[0][0]

-----

mixed0 (Concatenate)          (None, 20, 20, 256)   0
activation_287[0][0]

```

```

activation_289[0][0]
activation_292[0][0]
activation_293[0][0]
-----

conv2d_297 (Conv2D)          (None, 20, 20, 64)    16384    mixed0[0][0]
-----

batch_normalization_297 (BatchN (None, 20, 20, 64)    192
conv2d_297[0][0]
-----

activation_297 (Activation)    (None, 20, 20, 64)    0
batch_normalization_297[0][0]
-----

conv2d_295 (Conv2D)          (None, 20, 20, 48)    12288    mixed0[0][0]
-----

conv2d_298 (Conv2D)          (None, 20, 20, 96)    55296
activation_297[0][0]
-----

batch_normalization_295 (BatchN (None, 20, 20, 48)    144
conv2d_295[0][0]
-----

batch_normalization_298 (BatchN (None, 20, 20, 96)    288
conv2d_298[0][0]
-----

activation_295 (Activation)    (None, 20, 20, 48)    0
batch_normalization_295[0][0]
-----

activation_298 (Activation)    (None, 20, 20, 96)    0
batch_normalization_298[0][0]
-----

average_pooling2d_28 (AveragePo (None, 20, 20, 256)    0          mixed0[0][0]
-----

conv2d_294 (Conv2D)          (None, 20, 20, 64)    16384    mixed0[0][0]
-----

conv2d_296 (Conv2D)          (None, 20, 20, 64)    76800
activation_295[0][0]
-----

```

```

-----
conv2d_299 (Conv2D)          (None, 20, 20, 96)    82944
activation_298[0][0]
-----

```

```

-----
conv2d_300 (Conv2D)          (None, 20, 20, 64)    16384
average_pooling2d_28[0][0]
-----

```

```

-----
batch_normalization_294 (BatchN (None, 20, 20, 64)    192
conv2d_294[0][0]
-----

```

```

-----
batch_normalization_296 (BatchN (None, 20, 20, 64)    192
conv2d_296[0][0]
-----

```

```

-----
batch_normalization_299 (BatchN (None, 20, 20, 96)    288
conv2d_299[0][0]
-----

```

```

-----
batch_normalization_300 (BatchN (None, 20, 20, 64)    192
conv2d_300[0][0]
-----

```

```

-----
activation_294 (Activation)    (None, 20, 20, 64)    0
batch_normalization_294[0][0]
-----

```

```

-----
activation_296 (Activation)    (None, 20, 20, 64)    0
batch_normalization_296[0][0]
-----

```

```

-----
activation_299 (Activation)    (None, 20, 20, 96)    0
batch_normalization_299[0][0]
-----

```

```

-----
activation_300 (Activation)    (None, 20, 20, 64)    0
batch_normalization_300[0][0]
-----

```

```

-----
mixed1 (Concatenate)          (None, 20, 20, 288)   0
activation_294[0][0]
activation_296[0][0]
activation_299[0][0]
activation_300[0][0]
-----
-----

```

conv2d_304 (Conv2D)	(None, 20, 20, 64)	18432	mixed1[0][0]
-----			
batch_normalization_304 (Batch Normalization)	(None, 20, 20, 64)	192	
conv2d_304[0][0]			
-----			
activation_304 (Activation)	(None, 20, 20, 64)	0	
batch_normalization_304[0][0]			
-----			
conv2d_302 (Conv2D)	(None, 20, 20, 48)	13824	mixed1[0][0]
-----			
conv2d_305 (Conv2D)	(None, 20, 20, 96)	55296	
activation_304[0][0]			
-----			
batch_normalization_302 (Batch Normalization)	(None, 20, 20, 48)	144	
conv2d_302[0][0]			
-----			
batch_normalization_305 (Batch Normalization)	(None, 20, 20, 96)	288	
conv2d_305[0][0]			
-----			
activation_302 (Activation)	(None, 20, 20, 48)	0	
batch_normalization_302[0][0]			
-----			
activation_305 (Activation)	(None, 20, 20, 96)	0	
batch_normalization_305[0][0]			
-----			
average_pooling2d_29 (Average Pooling)	(None, 20, 20, 288)	0	mixed1[0][0]
-----			
conv2d_301 (Conv2D)	(None, 20, 20, 64)	18432	mixed1[0][0]
-----			
conv2d_303 (Conv2D)	(None, 20, 20, 64)	76800	
activation_302[0][0]			
-----			
conv2d_306 (Conv2D)	(None, 20, 20, 96)	82944	
activation_305[0][0]			
-----			
-----			

conv2d_307 (Conv2D)	(None, 20, 20, 64)	18432	
average_pooling2d_29[0][0]			
-----			
batch_normalization_301 (BatchN	(None, 20, 20, 64)	192	
conv2d_301[0][0]			
-----			
batch_normalization_303 (BatchN	(None, 20, 20, 64)	192	
conv2d_303[0][0]			
-----			
batch_normalization_306 (BatchN	(None, 20, 20, 96)	288	
conv2d_306[0][0]			
-----			
batch_normalization_307 (BatchN	(None, 20, 20, 64)	192	
conv2d_307[0][0]			
-----			
activation_301 (Activation)	(None, 20, 20, 64)	0	
batch_normalization_301[0][0]			
-----			
activation_303 (Activation)	(None, 20, 20, 64)	0	
batch_normalization_303[0][0]			
-----			
activation_306 (Activation)	(None, 20, 20, 96)	0	
batch_normalization_306[0][0]			
-----			
activation_307 (Activation)	(None, 20, 20, 64)	0	
batch_normalization_307[0][0]			
-----			
mixed2 (Concatenate)	(None, 20, 20, 288)	0	
activation_301[0][0]			
activation_303[0][0]			
activation_306[0][0]			
activation_307[0][0]			
-----			
conv2d_309 (Conv2D)	(None, 20, 20, 64)	18432	mixed2[0][0]
-----			
batch_normalization_309 (BatchN	(None, 20, 20, 64)	192	
conv2d_309[0][0]			

```

-----
activation_309 (Activation)      (None, 20, 20, 64)    0
batch_normalization_309[0][0]

-----

conv2d_310 (Conv2D)             (None, 20, 20, 96)    55296
activation_309[0][0]

-----

batch_normalization_310 (BatchN (None, 20, 20, 96)    288
conv2d_310[0][0]

-----

activation_310 (Activation)      (None, 20, 20, 96)    0
batch_normalization_310[0][0]

-----

conv2d_308 (Conv2D)             (None, 9, 9, 384)     995328    mixed2[0][0]

-----

conv2d_311 (Conv2D)             (None, 9, 9, 96)      82944
activation_310[0][0]

-----

batch_normalization_308 (BatchN (None, 9, 9, 384)     1152
conv2d_308[0][0]

-----

batch_normalization_311 (BatchN (None, 9, 9, 96)      288
conv2d_311[0][0]

-----

activation_308 (Activation)      (None, 9, 9, 384)     0
batch_normalization_308[0][0]

-----

activation_311 (Activation)      (None, 9, 9, 96)      0
batch_normalization_311[0][0]

-----

max_pooling2d_14 (MaxPooling2D) (None, 9, 9, 288)     0          mixed2[0][0]

-----

mixed3 (Concatenate)            (None, 9, 9, 768)     0
activation_308[0][0]
activation_311[0][0]
max_pooling2d_14[0][0]

```

conv2d_316 (Conv2D)	(None, 9, 9, 128)	98304	mixed3[0][0]
batch_normalization_316 (Batch Normalization)	(None, 9, 9, 128)	384	
conv2d_316[0][0]			
activation_316 (Activation)	(None, 9, 9, 128)	0	
batch_normalization_316[0][0]			
conv2d_317 (Conv2D)	(None, 9, 9, 128)	114688	
activation_316[0][0]			
batch_normalization_317 (Batch Normalization)	(None, 9, 9, 128)	384	
conv2d_317[0][0]			
activation_317 (Activation)	(None, 9, 9, 128)	0	
batch_normalization_317[0][0]			
conv2d_313 (Conv2D)	(None, 9, 9, 128)	98304	mixed3[0][0]
conv2d_318 (Conv2D)	(None, 9, 9, 128)	114688	
activation_317[0][0]			
batch_normalization_313 (Batch Normalization)	(None, 9, 9, 128)	384	
conv2d_313[0][0]			
batch_normalization_318 (Batch Normalization)	(None, 9, 9, 128)	384	
conv2d_318[0][0]			
activation_313 (Activation)	(None, 9, 9, 128)	0	
batch_normalization_313[0][0]			
activation_318 (Activation)	(None, 9, 9, 128)	0	
batch_normalization_318[0][0]			



conv2d_314 (Conv2D)	(None, 9, 9, 128)	114688	
activation_313[0][0]			
-----			
conv2d_319 (Conv2D)	(None, 9, 9, 128)	114688	
activation_318[0][0]			
-----			
batch_normalization_314 (BatchN	(None, 9, 9, 128)	384	
conv2d_314[0][0]			
-----			
batch_normalization_319 (BatchN	(None, 9, 9, 128)	384	
conv2d_319[0][0]			
-----			
activation_314 (Activation)	(None, 9, 9, 128)	0	
batch_normalization_314[0][0]			
-----			
activation_319 (Activation)	(None, 9, 9, 128)	0	
batch_normalization_319[0][0]			
-----			
average_pooling2d_30 (AveragePo	(None, 9, 9, 768)	0	mixed3[0][0]
-----			
conv2d_312 (Conv2D)	(None, 9, 9, 192)	147456	mixed3[0][0]
-----			
conv2d_315 (Conv2D)	(None, 9, 9, 192)	172032	
activation_314[0][0]			
-----			
conv2d_320 (Conv2D)	(None, 9, 9, 192)	172032	
activation_319[0][0]			
-----			
conv2d_321 (Conv2D)	(None, 9, 9, 192)	147456	
average_pooling2d_30[0][0]			
-----			
batch_normalization_312 (BatchN	(None, 9, 9, 192)	576	
conv2d_312[0][0]			
-----			
batch_normalization_315 (BatchN	(None, 9, 9, 192)	576	
conv2d_315[0][0]			

```

-----
-----
batch_normalization_320 (BatchN (None, 9, 9, 192)    576
conv2d_320[0][0]
-----
-----
batch_normalization_321 (BatchN (None, 9, 9, 192)    576
conv2d_321[0][0]
-----
-----
activation_312 (Activation)      (None, 9, 9, 192)    0
batch_normalization_312[0][0]
-----
-----
activation_315 (Activation)      (None, 9, 9, 192)    0
batch_normalization_315[0][0]
-----
-----
activation_320 (Activation)      (None, 9, 9, 192)    0
batch_normalization_320[0][0]
-----
-----
activation_321 (Activation)      (None, 9, 9, 192)    0
batch_normalization_321[0][0]
-----
-----
mixed4 (Concatenate)            (None, 9, 9, 768)    0
activation_312[0][0]
activation_315[0][0]
activation_320[0][0]
activation_321[0][0]
-----
-----
conv2d_326 (Conv2D)              (None, 9, 9, 160)    122880    mixed4[0][0]
-----
-----
batch_normalization_326 (BatchN (None, 9, 9, 160)    480
conv2d_326[0][0]
-----
-----
activation_326 (Activation)      (None, 9, 9, 160)    0
batch_normalization_326[0][0]
-----
-----
conv2d_327 (Conv2D)              (None, 9, 9, 160)    179200
activation_326[0][0]
-----
-----

```

batch_normalization_327 (BatchN	(None, 9, 9, 160)	480	
conv2d_327[0][0]			
-----			
activation_327 (Activation)	(None, 9, 9, 160)	0	
batch_normalization_327[0][0]			
-----			
conv2d_323 (Conv2D)	(None, 9, 9, 160)	122880	mixed4[0][0]
-----			
conv2d_328 (Conv2D)	(None, 9, 9, 160)	179200	
activation_327[0][0]			
-----			
batch_normalization_323 (BatchN	(None, 9, 9, 160)	480	
conv2d_323[0][0]			
-----			
batch_normalization_328 (BatchN	(None, 9, 9, 160)	480	
conv2d_328[0][0]			
-----			
activation_323 (Activation)	(None, 9, 9, 160)	0	
batch_normalization_323[0][0]			
-----			
activation_328 (Activation)	(None, 9, 9, 160)	0	
batch_normalization_328[0][0]			
-----			
conv2d_324 (Conv2D)	(None, 9, 9, 160)	179200	
activation_323[0][0]			
-----			
conv2d_329 (Conv2D)	(None, 9, 9, 160)	179200	
activation_328[0][0]			
-----			
batch_normalization_324 (BatchN	(None, 9, 9, 160)	480	
conv2d_324[0][0]			
-----			
batch_normalization_329 (BatchN	(None, 9, 9, 160)	480	
conv2d_329[0][0]			
-----			
activation_324 (Activation)	(None, 9, 9, 160)	0	

```

batch_normalization_324[0][0]
-----
-----
activation_329 (Activation)      (None, 9, 9, 160)      0
batch_normalization_329[0][0]
-----
-----
average_pooling2d_31 (AveragePo (None, 9, 9, 768)      0      mixed4[0][0]
-----
-----
conv2d_322 (Conv2D)              (None, 9, 9, 192)      147456      mixed4[0][0]
-----
-----
conv2d_325 (Conv2D)              (None, 9, 9, 192)      215040
activation_324[0][0]
-----
-----
conv2d_330 (Conv2D)              (None, 9, 9, 192)      215040
activation_329[0][0]
-----
-----
conv2d_331 (Conv2D)              (None, 9, 9, 192)      147456
average_pooling2d_31[0][0]
-----
-----
batch_normalization_322 (BatchN (None, 9, 9, 192)      576
conv2d_322[0][0]
-----
-----
batch_normalization_325 (BatchN (None, 9, 9, 192)      576
conv2d_325[0][0]
-----
-----
batch_normalization_330 (BatchN (None, 9, 9, 192)      576
conv2d_330[0][0]
-----
-----
batch_normalization_331 (BatchN (None, 9, 9, 192)      576
conv2d_331[0][0]
-----
-----
activation_322 (Activation)      (None, 9, 9, 192)      0
batch_normalization_322[0][0]
-----
-----
activation_325 (Activation)      (None, 9, 9, 192)      0
batch_normalization_325[0][0]
-----

```

```

-----
activation_330 (Activation)      (None, 9, 9, 192)      0
batch_normalization_330[0][0]

-----

activation_331 (Activation)      (None, 9, 9, 192)      0
batch_normalization_331[0][0]

-----

mixed5 (Concatenate)           (None, 9, 9, 768)      0
activation_322[0][0]
activation_325[0][0]
activation_330[0][0]
activation_331[0][0]

-----

conv2d_336 (Conv2D)             (None, 9, 9, 160)      122880      mixed5[0][0]

-----

batch_normalization_336 (BatchN (None, 9, 9, 160)      480
conv2d_336[0][0]

-----

activation_336 (Activation)      (None, 9, 9, 160)      0
batch_normalization_336[0][0]

-----

conv2d_337 (Conv2D)             (None, 9, 9, 160)      179200
activation_336[0][0]

-----

batch_normalization_337 (BatchN (None, 9, 9, 160)      480
conv2d_337[0][0]

-----

activation_337 (Activation)      (None, 9, 9, 160)      0
batch_normalization_337[0][0]

-----

conv2d_333 (Conv2D)             (None, 9, 9, 160)      122880      mixed5[0][0]

-----

conv2d_338 (Conv2D)             (None, 9, 9, 160)      179200
activation_337[0][0]

-----

batch_normalization_333 (BatchN (None, 9, 9, 160)      480
conv2d_333[0][0]

```

```

-----
-----
batch_normalization_338 (BatchN (None, 9, 9, 160)    480
conv2d_338[0][0]
-----
-----
activation_333 (Activation)    (None, 9, 9, 160)    0
batch_normalization_333[0][0]
-----
-----
activation_338 (Activation)    (None, 9, 9, 160)    0
batch_normalization_338[0][0]
-----
-----
conv2d_334 (Conv2D)            (None, 9, 9, 160)    179200
activation_333[0][0]
-----
-----
conv2d_339 (Conv2D)            (None, 9, 9, 160)    179200
activation_338[0][0]
-----
-----
batch_normalization_334 (BatchN (None, 9, 9, 160)    480
conv2d_334[0][0]
-----
-----
batch_normalization_339 (BatchN (None, 9, 9, 160)    480
conv2d_339[0][0]
-----
-----
activation_334 (Activation)    (None, 9, 9, 160)    0
batch_normalization_334[0][0]
-----
-----
activation_339 (Activation)    (None, 9, 9, 160)    0
batch_normalization_339[0][0]
-----
-----
average_pooling2d_32 (AveragePo (None, 9, 9, 768)    0          mixed5[0][0]
-----
-----
conv2d_332 (Conv2D)            (None, 9, 9, 192)    147456     mixed5[0][0]
-----
-----
conv2d_335 (Conv2D)            (None, 9, 9, 192)    215040
activation_334[0][0]
-----
-----

```

conv2d_340 (Conv2D)	(None, 9, 9, 192)	215040	
activation_339[0][0]			
-----			
conv2d_341 (Conv2D)	(None, 9, 9, 192)	147456	
average_pooling2d_32[0][0]			
-----			
batch_normalization_332 (BatchN	(None, 9, 9, 192)	576	
conv2d_332[0][0]			
-----			
batch_normalization_335 (BatchN	(None, 9, 9, 192)	576	
conv2d_335[0][0]			
-----			
batch_normalization_340 (BatchN	(None, 9, 9, 192)	576	
conv2d_340[0][0]			
-----			
batch_normalization_341 (BatchN	(None, 9, 9, 192)	576	
conv2d_341[0][0]			
-----			
activation_332 (Activation)	(None, 9, 9, 192)	0	
batch_normalization_332[0][0]			
-----			
activation_335 (Activation)	(None, 9, 9, 192)	0	
batch_normalization_335[0][0]			
-----			
activation_340 (Activation)	(None, 9, 9, 192)	0	
batch_normalization_340[0][0]			
-----			
activation_341 (Activation)	(None, 9, 9, 192)	0	
batch_normalization_341[0][0]			
-----			
mixed6 (Concatenate)	(None, 9, 9, 768)	0	
activation_332[0][0]			
activation_335[0][0]			
activation_340[0][0]			
activation_341[0][0]			
-----			
conv2d_346 (Conv2D)	(None, 9, 9, 192)	147456	mixed6[0][0]

```

-----
-----
batch_normalization_346 (BatchN (None, 9, 9, 192)    576
conv2d_346[0][0]
-----
-----
activation_346 (Activation)      (None, 9, 9, 192)    0
batch_normalization_346[0][0]
-----
-----
conv2d_347 (Conv2D)              (None, 9, 9, 192)    258048
activation_346[0][0]
-----
-----
batch_normalization_347 (BatchN (None, 9, 9, 192)    576
conv2d_347[0][0]
-----
-----
activation_347 (Activation)      (None, 9, 9, 192)    0
batch_normalization_347[0][0]
-----
-----
conv2d_343 (Conv2D)              (None, 9, 9, 192)    147456      mixed6[0][0]
-----
-----
conv2d_348 (Conv2D)              (None, 9, 9, 192)    258048
activation_347[0][0]
-----
-----
batch_normalization_343 (BatchN (None, 9, 9, 192)    576
conv2d_343[0][0]
-----
-----
batch_normalization_348 (BatchN (None, 9, 9, 192)    576
conv2d_348[0][0]
-----
-----
activation_343 (Activation)      (None, 9, 9, 192)    0
batch_normalization_343[0][0]
-----
-----
activation_348 (Activation)      (None, 9, 9, 192)    0
batch_normalization_348[0][0]
-----
-----
conv2d_344 (Conv2D)              (None, 9, 9, 192)    258048
activation_343[0][0]
-----

```



conv2d_349 (Conv2D)	(None, 9, 9, 192)	258048	
activation_348[0][0]			
-----			
batch_normalization_344 (BatchN	(None, 9, 9, 192)	576	
conv2d_344[0][0]			
-----			
batch_normalization_349 (BatchN	(None, 9, 9, 192)	576	
conv2d_349[0][0]			
-----			
activation_344 (Activation)	(None, 9, 9, 192)	0	
batch_normalization_344[0][0]			
-----			
activation_349 (Activation)	(None, 9, 9, 192)	0	
batch_normalization_349[0][0]			
-----			
average_pooling2d_33 (AveragePo	(None, 9, 9, 768)	0	mixed6[0][0]
-----			
conv2d_342 (Conv2D)	(None, 9, 9, 192)	147456	mixed6[0][0]
-----			
conv2d_345 (Conv2D)	(None, 9, 9, 192)	258048	
activation_344[0][0]			
-----			
conv2d_350 (Conv2D)	(None, 9, 9, 192)	258048	
activation_349[0][0]			
-----			
conv2d_351 (Conv2D)	(None, 9, 9, 192)	147456	
average_pooling2d_33[0][0]			
-----			
batch_normalization_342 (BatchN	(None, 9, 9, 192)	576	
conv2d_342[0][0]			
-----			
batch_normalization_345 (BatchN	(None, 9, 9, 192)	576	
conv2d_345[0][0]			
-----			
batch_normalization_350 (BatchN	(None, 9, 9, 192)	576	

```

conv2d_350[0][0]
-----
-----
batch_normalization_351 (BatchN (None, 9, 9, 192)    576
conv2d_351[0][0]
-----
-----
activation_342 (Activation)      (None, 9, 9, 192)    0
batch_normalization_342[0][0]
-----
-----
activation_345 (Activation)      (None, 9, 9, 192)    0
batch_normalization_345[0][0]
-----
-----
activation_350 (Activation)      (None, 9, 9, 192)    0
batch_normalization_350[0][0]
-----
-----
activation_351 (Activation)      (None, 9, 9, 192)    0
batch_normalization_351[0][0]
-----
-----
mixed7 (Concatenate)            (None, 9, 9, 768)    0
activation_342[0][0]
activation_345[0][0]
activation_350[0][0]
activation_351[0][0]
-----
-----
conv2d_354 (Conv2D)              (None, 9, 9, 192)    147456    mixed7[0][0]
-----
-----
batch_normalization_354 (BatchN (None, 9, 9, 192)    576
conv2d_354[0][0]
-----
-----
activation_354 (Activation)      (None, 9, 9, 192)    0
batch_normalization_354[0][0]
-----
-----
conv2d_355 (Conv2D)              (None, 9, 9, 192)    258048
activation_354[0][0]
-----
-----
batch_normalization_355 (BatchN (None, 9, 9, 192)    576
conv2d_355[0][0]
-----

```

```

-----
activation_355 (Activation)      (None, 9, 9, 192)      0
batch_normalization_355[0][0]
-----

-----
conv2d_352 (Conv2D)              (None, 9, 9, 192)      147456      mixed7[0][0]
-----

-----
conv2d_356 (Conv2D)              (None, 9, 9, 192)      258048
activation_355[0][0]
-----

-----
batch_normalization_352 (BatchN (None, 9, 9, 192)      576
conv2d_352[0][0]
-----

-----
batch_normalization_356 (BatchN (None, 9, 9, 192)      576
conv2d_356[0][0]
-----

-----
activation_352 (Activation)      (None, 9, 9, 192)      0
batch_normalization_352[0][0]
-----

-----
activation_356 (Activation)      (None, 9, 9, 192)      0
batch_normalization_356[0][0]
-----

-----
conv2d_353 (Conv2D)              (None, 4, 4, 320)      552960
activation_352[0][0]
-----

-----
conv2d_357 (Conv2D)              (None, 4, 4, 192)      331776
activation_356[0][0]
-----

-----
batch_normalization_353 (BatchN (None, 4, 4, 320)      960
conv2d_353[0][0]
-----

-----
batch_normalization_357 (BatchN (None, 4, 4, 192)      576
conv2d_357[0][0]
-----

-----
activation_353 (Activation)      (None, 4, 4, 320)      0
batch_normalization_353[0][0]
-----

```

activation_357 (Activation)	(None, 4, 4, 192)	0	
batch_normalization_357[0][0]			
-----			
max_pooling2d_15 (MaxPooling2D)	(None, 4, 4, 768)	0	mixed7[0][0]
-----			
mixed8 (Concatenate)	(None, 4, 4, 1280)	0	
activation_353[0][0]			
activation_357[0][0]			
max_pooling2d_15[0][0]			
-----			
conv2d_362 (Conv2D)	(None, 4, 4, 448)	573440	mixed8[0][0]
-----			
batch_normalization_362 (BatchN	(None, 4, 4, 448)	1344	
conv2d_362[0][0]			
-----			
activation_362 (Activation)	(None, 4, 4, 448)	0	
batch_normalization_362[0][0]			
-----			
conv2d_359 (Conv2D)	(None, 4, 4, 384)	491520	mixed8[0][0]
-----			
conv2d_363 (Conv2D)	(None, 4, 4, 384)	1548288	
activation_362[0][0]			
-----			
batch_normalization_359 (BatchN	(None, 4, 4, 384)	1152	
conv2d_359[0][0]			
-----			
batch_normalization_363 (BatchN	(None, 4, 4, 384)	1152	
conv2d_363[0][0]			
-----			
activation_359 (Activation)	(None, 4, 4, 384)	0	
batch_normalization_359[0][0]			
-----			
activation_363 (Activation)	(None, 4, 4, 384)	0	
batch_normalization_363[0][0]			
-----			
conv2d_360 (Conv2D)	(None, 4, 4, 384)	442368	

```

activation_359[0][0]
-----

conv2d_361 (Conv2D)          (None, 4, 4, 384)    442368
activation_359[0][0]
-----

conv2d_364 (Conv2D)          (None, 4, 4, 384)    442368
activation_363[0][0]
-----

conv2d_365 (Conv2D)          (None, 4, 4, 384)    442368
activation_363[0][0]
-----

average_pooling2d_34 (AveragePo (None, 4, 4, 1280)    0          mixed8[0][0]
-----

conv2d_358 (Conv2D)          (None, 4, 4, 320)    409600    mixed8[0][0]
-----

batch_normalization_360 (BatchN (None, 4, 4, 384)    1152
conv2d_360[0][0]
-----

batch_normalization_361 (BatchN (None, 4, 4, 384)    1152
conv2d_361[0][0]
-----

batch_normalization_364 (BatchN (None, 4, 4, 384)    1152
conv2d_364[0][0]
-----

batch_normalization_365 (BatchN (None, 4, 4, 384)    1152
conv2d_365[0][0]
-----

conv2d_366 (Conv2D)          (None, 4, 4, 192)    245760
average_pooling2d_34[0][0]
-----

batch_normalization_358 (BatchN (None, 4, 4, 320)    960
conv2d_358[0][0]
-----

activation_360 (Activation)    (None, 4, 4, 384)    0
batch_normalization_360[0][0]
-----

```

```

-----
activation_361 (Activation)      (None, 4, 4, 384)      0
batch_normalization_361[0][0]

-----

activation_364 (Activation)      (None, 4, 4, 384)      0
batch_normalization_364[0][0]

-----

activation_365 (Activation)      (None, 4, 4, 384)      0
batch_normalization_365[0][0]

-----

batch_normalization_366 (BatchN (None, 4, 4, 192)      576
conv2d_366[0][0]

-----

activation_358 (Activation)      (None, 4, 4, 320)      0
batch_normalization_358[0][0]

-----

mixed9_0 (Concatenate)          (None, 4, 4, 768)      0
activation_360[0][0]
activation_361[0][0]

-----

concatenate_6 (Concatenate)      (None, 4, 4, 768)      0
activation_364[0][0]
activation_365[0][0]

-----

activation_366 (Activation)      (None, 4, 4, 192)      0
batch_normalization_366[0][0]

-----

mixed9 (Concatenate)            (None, 4, 4, 2048)      0
activation_358[0][0]
mixed9_0[0][0]

concatenate_6[0][0]
activation_366[0][0]

-----

conv2d_371 (Conv2D)              (None, 4, 4, 448)      917504      mixed9[0][0]

-----

batch_normalization_371 (BatchN (None, 4, 4, 448)      1344
conv2d_371[0][0]

-----

```

activation_371 (Activation)	(None, 4, 4, 448)	0	
batch_normalization_371[0][0]			
conv2d_368 (Conv2D)	(None, 4, 4, 384)	786432	mixed9[0][0]
conv2d_372 (Conv2D)	(None, 4, 4, 384)	1548288	
activation_371[0][0]			
batch_normalization_368 (BatchN	(None, 4, 4, 384)	1152	
conv2d_368[0][0]			
batch_normalization_372 (BatchN	(None, 4, 4, 384)	1152	
conv2d_372[0][0]			
activation_368 (Activation)	(None, 4, 4, 384)	0	
batch_normalization_368[0][0]			
activation_372 (Activation)	(None, 4, 4, 384)	0	
batch_normalization_372[0][0]			
conv2d_369 (Conv2D)	(None, 4, 4, 384)	442368	
activation_368[0][0]			
conv2d_370 (Conv2D)	(None, 4, 4, 384)	442368	
activation_368[0][0]			
conv2d_373 (Conv2D)	(None, 4, 4, 384)	442368	
activation_372[0][0]			
conv2d_374 (Conv2D)	(None, 4, 4, 384)	442368	
activation_372[0][0]			
average_pooling2d_35 (AveragePo	(None, 4, 4, 2048)	0	mixed9[0][0]
conv2d_367 (Conv2D)	(None, 4, 4, 320)	655360	mixed9[0][0]

```

-----
-----
batch_normalization_369 (BatchN (None, 4, 4, 384)    1152
conv2d_369[0][0]

-----
-----
batch_normalization_370 (BatchN (None, 4, 4, 384)    1152
conv2d_370[0][0]

-----
-----
batch_normalization_373 (BatchN (None, 4, 4, 384)    1152
conv2d_373[0][0]

-----
-----
batch_normalization_374 (BatchN (None, 4, 4, 384)    1152
conv2d_374[0][0]

-----
-----
conv2d_375 (Conv2D)                (None, 4, 4, 192)    393216
average_pooling2d_35[0][0]

-----
-----
batch_normalization_367 (BatchN (None, 4, 4, 320)    960
conv2d_367[0][0]

-----
-----
activation_369 (Activation)          (None, 4, 4, 384)    0
batch_normalization_369[0][0]

-----
-----
activation_370 (Activation)          (None, 4, 4, 384)    0
batch_normalization_370[0][0]

-----
-----
activation_373 (Activation)          (None, 4, 4, 384)    0
batch_normalization_373[0][0]

-----
-----
activation_374 (Activation)          (None, 4, 4, 384)    0
batch_normalization_374[0][0]

-----
-----
batch_normalization_375 (BatchN (None, 4, 4, 192)    576
conv2d_375[0][0]

-----
-----
activation_367 (Activation)          (None, 4, 4, 320)    0
batch_normalization_367[0][0]

```



```

-----
mixed9_1 (Concatenate)      (None, 4, 4, 768)    0
activation_369[0][0]
activation_370[0][0]
-----
concatenate_7 (Concatenate) (None, 4, 4, 768)    0
activation_373[0][0]
activation_374[0][0]
-----
activation_375 (Activation)  (None, 4, 4, 192)    0
batch_normalization_375[0][0]
-----
mixed10 (Concatenate)      (None, 4, 4, 2048)    0
activation_367[0][0]
mixed9_1[0][0]
concatenate_7[0][0]
activation_375[0][0]
=====
=====
Total params: 21,802,784
Trainable params: 0
Non-trainable params: 21,802,784
-----
-----

```

```

[112]: model = Sequential()
model.add(inception)
model.add(Flatten())

model.add(Dense(512,activation='relu'))
model.add(Dropout(0.4))
model.add(Dense(256,activation='relu'))
model.add(Dense(64,activation='relu'))
model.add(Dropout(0.4))
model.add(Dense(3,activation='softmax'))

```

```

[113]: model.summary()

```

Model: "sequential\_3"

```

-----
Layer (type)                 Output Shape              Param #
=====
inception_v3 (Functional)    (None, 4, 4, 2048)       21802784
-----

```

flatten_3 (Flatten)	(None, 32768)	0
-----		
dense_12 (Dense)	(None, 512)	16777728
-----		
dropout_6 (Dropout)	(None, 512)	0
-----		
dense_13 (Dense)	(None, 256)	131328
-----		
dense_14 (Dense)	(None, 64)	16448
-----		
dropout_7 (Dropout)	(None, 64)	0
-----		
dense_15 (Dense)	(None, 3)	195
=====		
Total params: 38,728,483		
Trainable params: 16,925,699		
Non-trainable params: 21,802,784		
-----		

```
[114]: model.
        ↪ compile(loss='categorical_crossentropy', optimizer='adam', metrics=['accuracy'])
        #callbacks
        early_stop = EarlyStopping(monitor = 'val_accuracy', patience=3)
```

```
[115]: import time

        start = time.perf_counter()

        model.fit(train_generator, epochs=30,
        ↪ validation_data=validation_generator, callbacks=[early_stop])

        elapsed = time.perf_counter() - start
```

```
Epoch 1/30
21/21 [=====] - 39s 2s/step - loss: 3.0289 - accuracy:
0.3841 - val_loss: 0.9531 - val_accuracy: 0.4362
Epoch 2/30
21/21 [=====] - 42s 2s/step - loss: 1.0111 - accuracy:
0.4777 - val_loss: 0.7067 - val_accuracy: 0.5705
Epoch 3/30
21/21 [=====] - 35s 2s/step - loss: 0.8604 - accuracy:
0.5536 - val_loss: 0.6962 - val_accuracy: 0.5705
Epoch 4/30
21/21 [=====] - 40s 2s/step - loss: 0.7697 - accuracy:
0.5776 - val_loss: 0.6254 - val_accuracy: 0.6309
Epoch 5/30
```

```

21/21 [=====] - 36s 2s/step - loss: 0.7364 - accuracy:
0.6293 - val_loss: 0.5875 - val_accuracy: 0.6443
Epoch 6/30
21/21 [=====] - 35s 2s/step - loss: 0.7181 - accuracy:
0.6244 - val_loss: 0.6263 - val_accuracy: 0.6443
Epoch 7/30
21/21 [=====] - 42s 2s/step - loss: 0.7045 - accuracy:
0.6436 - val_loss: 0.5552 - val_accuracy: 0.6510
Epoch 8/30
21/21 [=====] - 41s 2s/step - loss: 0.6553 - accuracy:
0.6186 - val_loss: 0.5519 - val_accuracy: 0.6779
Epoch 9/30
21/21 [=====] - 29s 1s/step - loss: 0.6612 - accuracy:
0.6665 - val_loss: 0.6127 - val_accuracy: 0.6443
Epoch 10/30
21/21 [=====] - 27s 1s/step - loss: 0.8600 - accuracy:
0.5894 - val_loss: 0.5974 - val_accuracy: 0.6711
Epoch 11/30
21/21 [=====] - 29s 1s/step - loss: 0.6950 - accuracy:
0.6459 - val_loss: 0.5552 - val_accuracy: 0.6443

```

```

[121]: import pandas as pd
history = pd.DataFrame(model.history.history)
history

```

```

[121]:      loss  accuracy  val_loss  val_accuracy
0    2.179312  0.418154  0.953067      0.436242
1    0.961773  0.522110  0.706701      0.570470
2    0.829118  0.577192  0.696200      0.570470
3    0.753635  0.591932  0.625395      0.630872
4    0.750024  0.617533  0.587511      0.644295
5    0.718781  0.623739  0.626285      0.644295
6    0.707123  0.631497  0.555202      0.651007
7    0.694791  0.630722  0.551940      0.677852
8    0.654065  0.645462  0.612667      0.644295
9    0.806783  0.613654  0.597448      0.671141
10   0.675678  0.644686  0.555222      0.644295

```

```

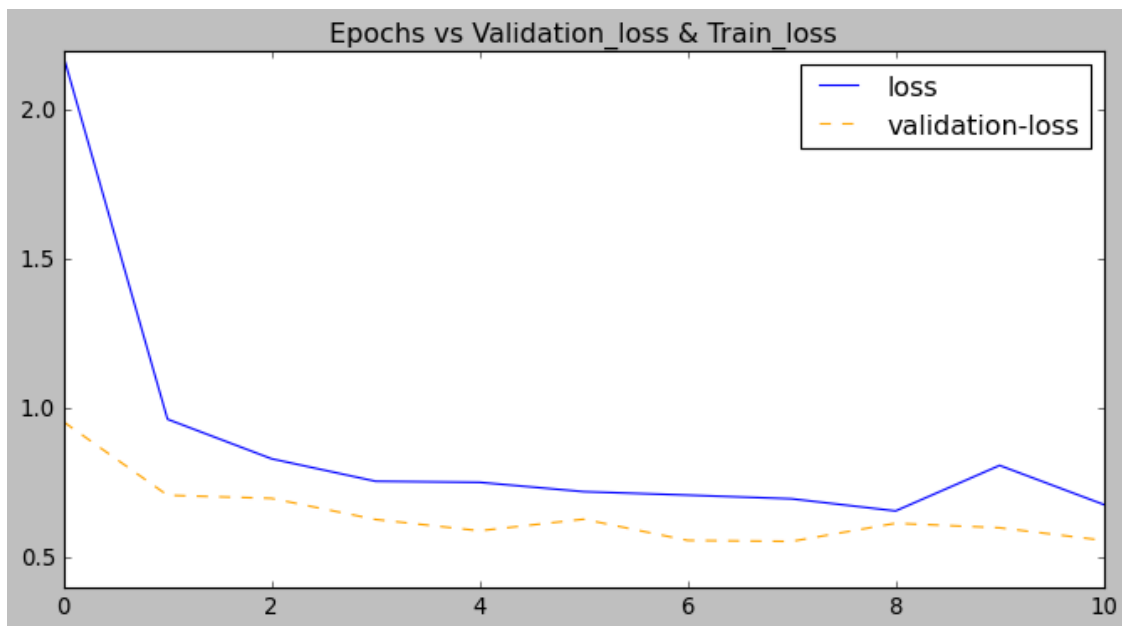
[129]: plt.style.use('classic')
plt.figure(figsize=(10,5))
plt.plot(history['loss'],label='loss')
plt.plot(history['val_loss'],ls='--',color='orange',label='validation-loss')
plt.title('Epochs vs Validation_loss & Train_loss')
plt.legend()

```

```

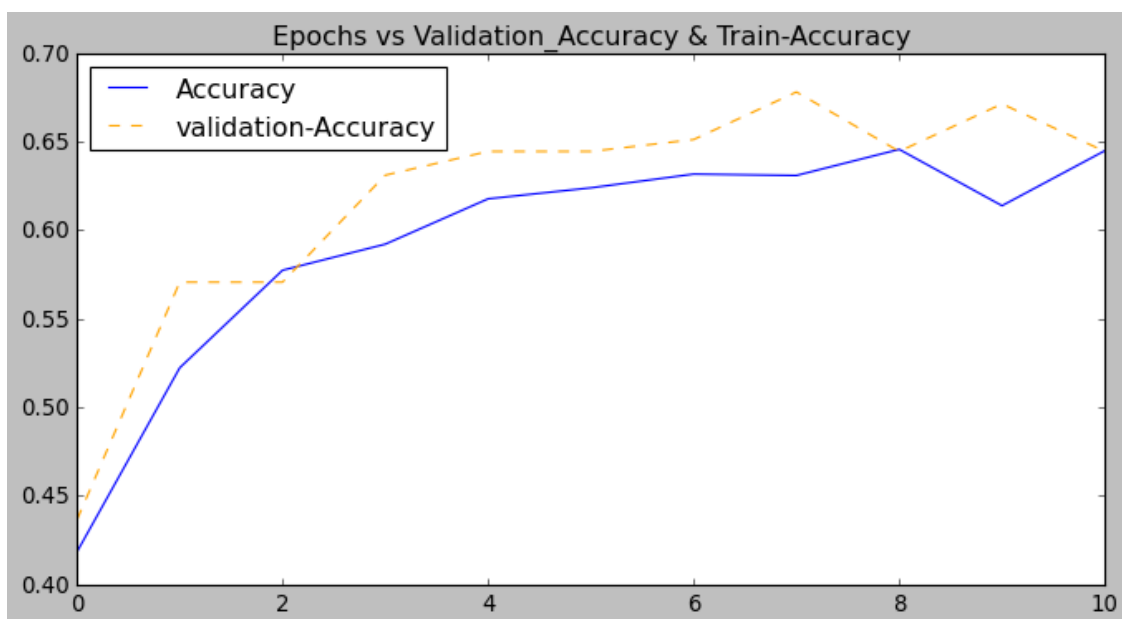
[129]: <matplotlib.legend.Legend at 0x7fd0b6975dc0>

```



```
[132]: plt.style.use('classic')
plt.figure(figsize=(10,5))
plt.plot(history['accuracy'],label='Accuracy')
plt.
    ↳plot(history['val_accuracy'],ls='--',color='orange',label='validation-Accuracy')
plt.title('Epochs vs Validation_Accuracy & Train-Accuracy')
plt.legend(loc='upper left')
```

[132]: <matplotlib.legend.Legend at 0x7fd0b0bc60a0>



```
[133]: predictions = model.predict_classes(validation_generator)
```

```
/Users/rohan/opt/anaconda3/lib/python3.8/site-  
packages/tensorflow/python/keras/engine/sequential.py:450: UserWarning:  
`model.predict_classes()` is deprecated and will be removed after 2021-01-01.  
Please use instead: * `np.argmax(model.predict(x), axis=-1)`, if your model  
does multi-class classification (e.g. if it uses a `softmax` last-layer  
activation). * `(model.predict(x) > 0.5).astype("int32")`, if your model does  
binary classification (e.g. if it uses a `sigmoid` last-layer activation).  
warnings.warn("`model.predict_classes()` is deprecated and '
```

```
[134]: predictions
```

```
[134]: array([0, 0, 0, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 2, 0, 2, 0, 0, 0, 0, 2, 0,  
        0, 2, 0, 2, 2, 0, 0, 2, 0, 0, 0, 2, 0, 0, 0, 2, 0, 0, 0, 2, 2, 0,  
        0, 2, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 2, 0, 0, 2, 2, 2, 2, 0, 0,  
        0, 2, 0, 2, 0, 0, 0, 0, 0, 2, 0, 0, 0, 2, 2, 0, 0, 2, 0, 2, 0, 2,  
        2, 0, 0, 0, 0, 0, 0, 0, 2, 0, 2, 0, 2, 0, 0, 0, 0, 0, 0, 0,  
        0, 0, 0, 0, 2, 0, 0, 0, 0, 2, 0, 0, 2, 0, 2, 0, 0, 2, 2, 0, 0, 0,  
        0, 0, 0, 2, 0, 2, 0, 2, 0, 0, 2, 0, 2, 0, 0, 0, 0])
```

```
[139]: # saving model  
model.save('inception-model.h5')
```

```
[137]: # Deconstructing Validation Set as x_test, and y_test for classification report
```

```
[236]: x_train, y_train = next(train_generator)  
x_test, y_test = next(validation_generator)  
class_names = validation_generator.class_indices
```

```
[237]: y_test[index].argmax()
```

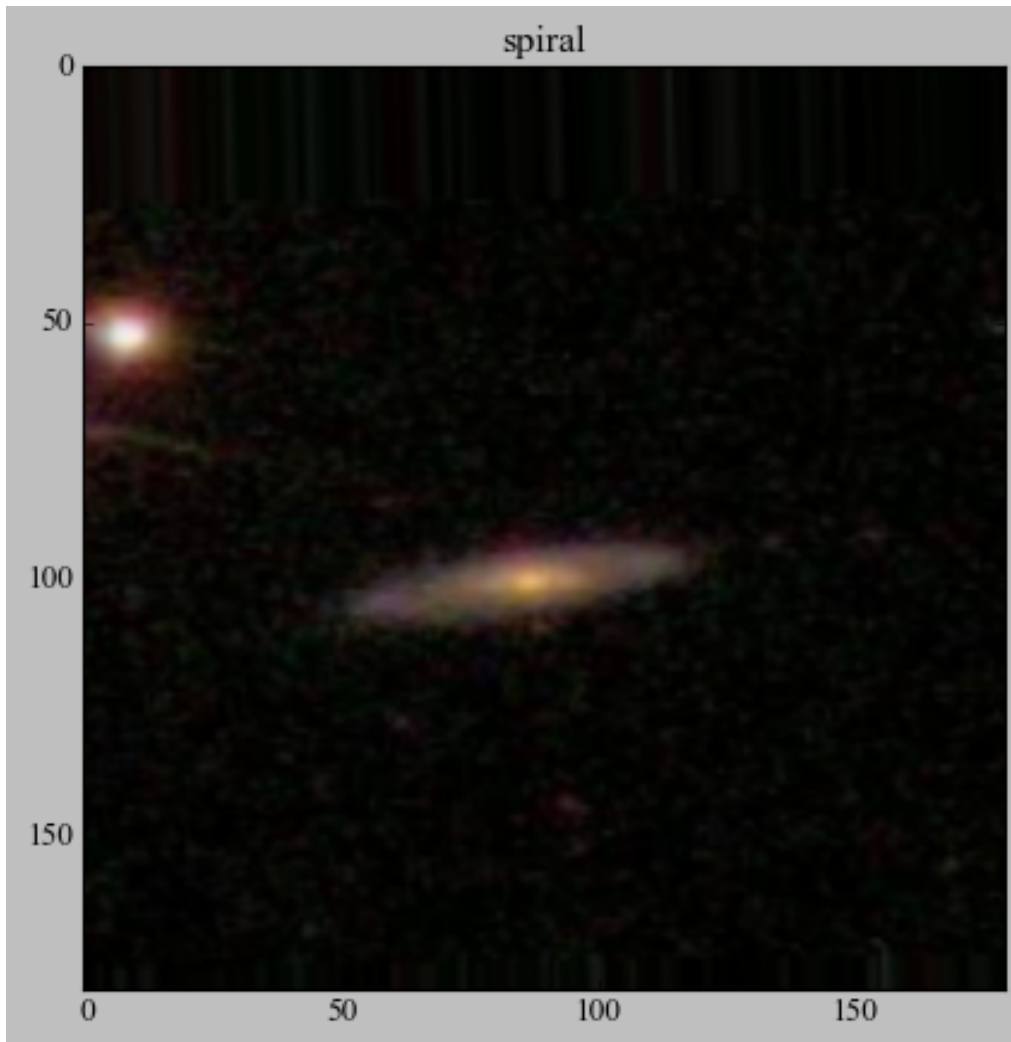
```
[237]: 2
```

```
[238]: x_test.shape
```

```
[238]: (64, 180, 180, 3)
```

```
[227]: index = 63  
plt.imshow(x_test[index])  
plt.title(list(class_names.keys())[y_test[index].argmax()])
```

```
[227]: Text(0.5, 1.0, 'spiral')
```



```
[228]: predictions
```

```
[228]: array([0, 0, 0, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 2, 0, 2, 0, 0, 0, 0, 2, 0,
          0, 2, 0, 2, 2, 0, 0, 2, 0, 0, 0, 2, 0, 0, 0, 2, 0, 0, 0, 2, 2, 0,
          0, 2, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 2, 0, 0, 2, 2, 2, 2, 2, 0, 0,
          0, 2, 0, 2, 0, 0, 0, 0, 0, 2, 0, 0, 0, 2, 2, 0, 0, 2, 0, 2, 0, 2,
          2, 0, 0, 0, 0, 0, 0, 0, 2, 0, 2, 0, 2, 0, 0, 0, 0, 0, 0, 0, 0,
          0, 0, 0, 0, 2, 0, 0, 0, 0, 2, 0, 0, 2, 0, 2, 0, 0, 2, 2, 0, 0, 0,
          0, 0, 0, 2, 0, 2, 0, 2, 0, 0, 2, 0, 2, 0, 0, 0, 0])
```

```
[248]: predictions = model.predict_classes(x_train)
```

```
/Users/rohan/opt/anaconda3/lib/python3.8/site-
packages/tensorflow/python/keras/engine/sequential.py:450: UserWarning:
`model.predict_classes()` is deprecated and will be removed after 2021-01-01.
Please use instead: * `np.argmax(model.predict(x), axis=-1)`, if your model
```

does multi-class classification (e.g. if it uses a `softmax` last-layer activation).  
 .\* `(model.predict(x) > 0.5).astype("int32")`, if your model does binary classification (e.g. if it uses a `sigmoid` last-layer activation).  
 warnings.warn("`model.predict\_classes()` is deprecated and "

```
[242]: y_test = np.array([np.where(r==1)[0][0] for r in y_test])
```

```
[243]: y_test
```

```
[243]: array([1, 1, 0, 1, 2, 2, 0, 1, 1, 1, 0, 0, 0, 0, 2, 0, 2, 2, 1, 1, 2, 0,
          0, 2, 2, 1, 0, 1, 1, 2, 0, 0, 1, 1, 2, 0, 2, 0, 0, 1, 2, 0, 2, 1,
          1, 0, 2, 0, 0, 2, 2, 1, 0, 0, 0, 2, 2, 2, 2, 1, 1, 2, 0, 2])
```

```
[245]: len(y_train)
```

```
[245]: 64
```

```
[250]: from sklearn.metrics import classification_report, confusion_matrix
```

```
[256]: history
```

```
[256]:      loss accuracy val_loss val_accuracy
0    2.179312  0.418154  0.953067      0.436242
1    0.961773  0.522110  0.706701      0.570470
2    0.829118  0.577192  0.696200      0.570470
3    0.753635  0.591932  0.625395      0.630872
4    0.750024  0.617533  0.587511      0.644295
5    0.718781  0.623739  0.626285      0.644295
6    0.707123  0.631497  0.555202      0.651007
7    0.694791  0.630722  0.551940      0.677852
8    0.654065  0.645462  0.612667      0.644295
9    0.806783  0.613654  0.597448      0.671141
10   0.675678  0.644686  0.555222      0.644295
```

## 0.0.9 Load Model

```
[10]: from tensorflow.keras.models import load_model
```

```
[11]: inceptionNet = load_model('inception-model.h5')
```

```
[15]: inceptionNet.summary()
```

Model: "sequential\_3"

Layer (type)	Output Shape	Param #
inception_v3 (Functional)	(None, 4, 4, 2048)	21802784

flatten_3 (Flatten)	(None, 32768)	0
-----		
dense_12 (Dense)	(None, 512)	16777728
-----		
dropout_6 (Dropout)	(None, 512)	0
-----		
dense_13 (Dense)	(None, 256)	131328
-----		
dense_14 (Dense)	(None, 64)	16448
-----		
dropout_7 (Dropout)	(None, 64)	0
-----		
dense_15 (Dense)	(None, 3)	195
=====		
Total params: 38,728,483		
Trainable params: 16,925,699		
Non-trainable params: 21,802,784		
-----		

#### 0.0.10 Miscellaneous

```
[17]: import os
path = '../samples/'
overview_path = '../samples/overview.txt'
eval_path = '../samples/evaluate.txt'

if os.path.exists(path):

    print('samples dir, exists..checking for dictionaries existence..')

    if os.path.exists(overview_path) and os.path.exists(eval_path):
        print('Data exists. no need of overwritting.')
    else:
        print("overview and eval doesn't exist, proceed to step-2")

else:
    print("samples/ dir is non-existent, Establishing one..")
    os.mkdir(path) # samples directory
```

```
samples dir, exists..checking for dictionaries existence..
overview and eval doesn't exist, proceed to step-2
```

```
[21]: shape = (train_generator.samples,*train_generator.image_shape)
shape
```

```
[21]: (1289, 180, 180, 3)
```

```
[23]: train_generator.class_indices.values()
```



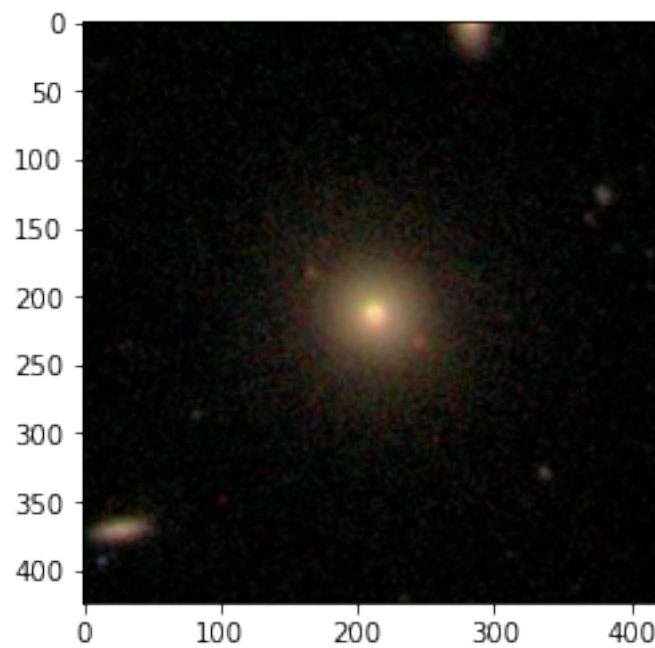
```
[23]: dict_values([0, 1, 2])
```

```
[37]: sample = list()
sample.append(elliptical_arr[0])
sample.append(lenticular_arr[0])
sample.append(spiral_arr[1])
```

```
[39]: sample = np.array(sample)
```

```
[43]:
```

```
[43]: <matplotlib.image.AxesImage at 0x7fbbc18aa400>
```



```
[44]: x_test, y_test = next(validation_generator)
```

```
[73]: # dictionary init
overview_dict = {}
eval_dict = {}

# fill the following -
# for overview
#string
kind = 'Image Data'
#tuple
dimensions = shape
```

```

#labels : str(list of unique target values)
targets = list(train_generator.class_indices.values())
#nd.array
data = sample
#nd.array or class_names
labels = ['elliptical','lenticular','spiral']

vars0 = ['kind','dimensions', 'targets', 'data', 'labels']

# filling overview_dict
for x in vars0:
    try:
        overview_dict[x] = eval(x)
    except:
        overview_dict[x] = x

# evaluate_dict

eval_dict = {'test_cases' : x_test, 'true': y_test, 'class_names':␣
→list(validation_generator.class_indices.keys()) , 'model':'/
→inceptionv3-model.h5'}

```

```

[77]: import pickle
# dump 1
with open(overview_path,'wb') as f:
    pickle.dump(overview_dict,f)

# dump 2
with open(eval_path,'wb') as f:
    pickle.dump(eval_dict,f)

```

```

[ ]: # desc-----string
# project_name-----string
# framework-----string
# prediction_type-----string
# network_type-----string
# architecture-----model()
# layers-----int
# hidden_units-----int
# activations-----string(list)
# epochs-----int
# metrics-----string(list)
# loss-----string
# optimiser-----string
# learning_rate-----float
# batch_size-----int/string
# train_performance-----float

```

```
# test_performance-----float
# classification_report-----string
# elapsed-----float
# summary-----string
# ipynb-----path
# plots-----path
```

```
[81]: inceptionNet.summary()
```

Model: "sequential\_3"

Layer (type)	Output Shape	Param #
inception_v3 (Functional)	(None, 4, 4, 2048)	21802784
flatten_3 (Flatten)	(None, 32768)	0
dense_12 (Dense)	(None, 512)	16777728
dropout_6 (Dropout)	(None, 512)	0
dense_13 (Dense)	(None, 256)	131328
dense_14 (Dense)	(None, 64)	16448
dropout_7 (Dropout)	(None, 64)	0
dense_15 (Dense)	(None, 3)	195

Total params: 38,728,483  
 Trainable params: 16,925,699  
 Non-trainable params: 21,802,784

```
[82]: synopsis = '''Considering my computation limitations as well as being a
↳neophyte to use pre-trained networks, working on this project was quite
↳challenging for me. This dataset was sliced down to a considerable amount to
↳train an InceptionV3 with Imagenet weights. This implementation shows the
↳quick efficiency that Transfer Learning delivers. My analogy for only
↳getting a 60% accuracy is ideally because of using only 500 samples of each
↳class instead of the original amount which is over 5000 samples for each
↳class.'''
```

```
[94]:
```

```

desc = '''The Galaxy Zoo Project is an ensembled collection of Images of
↳different classes of Galaxies categorised based on their Morphology. Data
↳collection is a part of Sloan Digital Sky Survey. six categories include -
↳elliptical, clockwise spiral, anticlockwise spiral, edge-on , lenticular, or
↳merger.'''
project_name = 'Galaxy Classification'
framework = 'Keras'
prediction_type = 'Classification of 3 Classes'
network_type = 'INCEPTION-V3'
architecture = str(inceptionNet.summary())
layers = 'InceptionV3 (48) + 7 Layers'
hidden_units = 'None'
activations = "['relu','softmax']"
epochs = '30, with earlystopping(patience=3) - trained for 8 epochs'
metrics = 'Accuracy'
loss = 'Categorical Cross-Entropy'
optimiser = 'Adam'
learning_rate = 0.001
batch_size = 64
train_performance = '64.32%'
test_performance = '64.46%'
classification_report = 'None'
elapsed = '23.4 Mins'
summary = synopsis
ipynb = './Projects/galaxy-classifier/galaxy-classifier-Inceptionv3.ipynb'
plots = './Projects/galaxy-classifier/Plots'

```

Model: "sequential\_3"

Layer (type)	Output Shape	Param #
inception_v3 (Functional)	(None, 4, 4, 2048)	21802784
flatten_3 (Flatten)	(None, 32768)	0
dense_12 (Dense)	(None, 512)	16777728
dropout_6 (Dropout)	(None, 512)	0
dense_13 (Dense)	(None, 256)	131328
dense_14 (Dense)	(None, 64)	16448
dropout_7 (Dropout)	(None, 64)	0
dense_15 (Dense)	(None, 3)	195

Total params: 38,728,483  
Trainable params: 16,925,699  
Non-trainable params: 21,802,784

```
[95]: var = ['desc', 'project_name', 'framework', 'prediction_type', 'network_type',
'architecture', 'layers', 'hidden_units', 'activations', 'epochs',
'metrics', 'loss', 'optimiser', 'learning_rate', 'batch_size', 'train_performance', 'test_performance',
'ipynb', 'plots']
param = {}
for val in var:

    try:
        param[val] = eval(val)

    except:
        param[val] = val
```

```
[103]: param
```

```
[103]: {'desc': 'The Galaxy Zoo Project is an ensembled collection of Images of
different classes of Galaxies categorised based on their Morphology. Data
collection is a part of Sloan Digital Sky Survey. six categories include -
elliptical, clockwise spiral, anticlockwise spiral, edge-on , lenticular, or
merger.',
'project_name': 'Galaxy Classification',
'framework': 'Keras',
'prediction_type': 'Classification of 3 Classes',
'network_type': 'INCEPTION-V3',
'architecture': '\nModel: "sequential_3"\n
-----
Layer (type) Output Shape
Param # \n===== \n
inception_v3 (Functional) (None, 4, 4, 2048) 21802784
\n
\nflatten_3
(Flatten) (None, 32768) 0
\n
\ndense_12
(Dense) (None, 512) 16777728
\n
\ndropout_6
(Dropout) (None, 512) 0
\n
\ndense_13
(Dense) (None, 256) 131328
\n
\ndense_14
(Dense) (None, 64) 16448
\n
\ndropout_7
(Dropout) (None, 64) 0
\n
\ndense_15
(Dense) (None, 3) 195
```

```

\n=====Total
params: 38,728,483\nTrainable params: 16,925,699\nNon-trainable params: 21,802,7
84\n-----\n\n',
'layers': 'InceptionV3 (48) + 7 Layers',
'hidden_units': 'None',
'activations': "['relu','softmax']",
'epochs': '30, with earlystopping(patience=3) - trained for 8 epochs',
'metrics': 'Accuracy',
'loss': 'Categorical Cross-Entropy',
'optimiser': 'Adam',
'learning_rate': 0.001,
'batch_size': 64,
'train_performance': '64.32%',
'test_performance': '64.46%',
'classification_report': 'None',
'elapsed': '23.4 Mins',
'summary': 'Considering my computation limitations as well as being a neophyte
to use pre-trained networks, working on this project was quite challenging for
me. This dataset was sliced down to a considerable amount to train an
InceptionV3 with Imagenet weights. This implementation shows the quick
efficiency that Transfer Learning delivers. My analogy for only getting a 60%
accuracy is ideally because of using only 500 samples of each class instead of
the original amount which is over 5000 samples for each class.',
'ipynb': './Projects/galaxy-classifier/galaxy-classifier-Inceptionv3.ipynb',
'plots': './Projects/galaxy-classifier/Plots'}

```

---

[ ]: