```
import tensorflow as tf
from tensorflow import keras
from keras import Sequential
from keras.layers import Dense, Flatten, Dropout, BatchNormalization,
Conv2D, MaxPooling2D
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import warnings
warnings.filterwarnings('ignore')
from tensorflow.keras.preprocessing.image import ImageDataGenerator,
load_img, img_to_array, save_img, array_to_img
df = pd.read csv('/content/emergency classification.csv')
df.head()
{"summary":"{\n \"name\": \"df\",\n \"rows\": 2352,\n \"fields\":
      {\n \"column\": \"image_names\",\n \"properties\": {\
[\n
        \"dtype\": \"string\",\n \"num_unique_values\": 2352,\
        \"samples\": [\n
                                 \"1960.jpg\",\n
\"668.jpg\",\n
               \"2082.jpg\"\n
                                           ],\n
\"semantic_type\": \"\",\n \"description\": \"\"\n
                                                           }\
          {\n \"column\": \"emergency_or_not\",\n
    },\n
\"properties\": {\n \"dtype\": \"number\",\n
                                                     \"std\":
           \"min\": 0,\n \"max\": 1,\n
0,\n
\"num unique values\": 2,\n
                                \"samples\": [\n
                                                         0.\n
1\n     ],\n \"semantic_type\": \"\",\n
\"description\": \"\"\n }\n
                                 }\n ]\
n}","type":"dataframe","variable name":"df"}
df['emergency or not'].value counts()
emergency or not
    1361
1
     991
Name: count, dtype: int64
import os
len(os.listdir('/content/sample data/Images'))
2345
```

For some reason all the 2352 images were not uploading, tried partially uploading them too. Hence, will upload the remaining 7 images in a different folder and will merge them later.

Handling Class Imbalance

```
data dir = '/content/sample data/Images'
output dir = '/content/sample data/aug images 2'
os.makedirs(output dir, exist ok=True)
class counts = df['emergency or not'].value counts() # Counts images
per class
minority class = class counts.idxmin() # Class with the fewest images
majority_class_count = class_counts.max() # Count of the majority
num augmented images = majority class count - class counts.min() #
Number of images to generate
num augmented images
370
datagen = ImageDataGenerator(
    rotation_range=30,
width_shift_range=0.2,
height_shift_range=0.2,
shear_range=0.2,
zoom_range=0.2,
horizontal_flip=True,
fill_mode='nearest'
# Random rotation up to 30 degrees
# Random horizontal shifts up to 20%
# Random vertical shifts up to 20%
# Shearing transformations
# Random zoom
# Randomly flip images horizontally
# Fill empty pixels after
     rotation range=30,
                                        # Random rotation up to 30 degrees
transformation
minority images = df[df['emergency or not'] == minority class]
['image_names']
current count = 0
for image name in minority images:
  img_path = os.path.join(data_dir, image name)
  img = load img(img path)
  img arr = img_to_array(img)
  img arr = np.expand dims(img arr, axis = \frac{0}{0})
  #Genearte augmented images
  for _ in range(num_augmented_images - current count):
     aug img = next(datagen.flow(img arr, batch size = 1))
[0].astype('uint8')
     save img(os.path.join(output dir, f'aug {current count}.jpg'),
```

```
aug_img)
    current_count +=1

if current_count >= num_augmented_images:
    break

print(f"Generated {num_augmented_images} augmented images.")

Generated 370 augmented images.
```

Created a folder with the name 'extra' for the 7 remaining images

```
len(os.listdir('/content/sample_data/extra'))

x = list(os.listdir(data_dir))
y = list(os.listdir(output_dir))
z = list(os.listdir('/content/sample_data/extra'))
len(x), len(y), len(z)

(2345, 370, 7)

res = x+y+z
len(res)

2722
```

Created a new dataframe for augmented images that will balance my minority class

```
df 1 = pd.DataFrame({'image names' : [file for file in
os.listdir(output dir)],
                     'emergency or not' : [1 for i in
range(len(os.listdir(output dir)))]})
df 1
{"summary":"{\n \"name\": \"df 1\",\n \"rows\": 370,\n \"fields\":
      {\n \"column\": \"image_names\",\n \"properties\": {\
[\n
         \"dtype\": \"string\",\n \"num unique values\": 370,\n
\"samples\": [\n \"aug_317.jpg\",\n \"aug_9.j
\"aug_70.jpg\"\n ],\n \"semantic_type\": \"\",\n
                                              \"dtvpe\":
\"number\",\n \"std\": 0,\n \"min\": 1,\n \"max\": 1,\n \"num_unique_values\": 1,\n \"samp\" [\n 1\n ],\n \"semantic_type\": \"\",\n
                                                      \"samples\":
\"description\": \"\"\n
                          }\n
                                   }\n 1\
n}","type":"dataframe","variable_name":"df_1"}
```

Combining all the dataframes and handling image paths so that i can use flow_from_dataframe() method

```
ex df = df.tail(7)
len(df), len(df 1), len(ex df)
df 2 = df.iloc[:len(df)-len(ex df)]
len(df 2)
2345
image data dir = '/content/sample data/Images'
aug data dir = '/content/sample data/aug images 2'
ex dir = '/content/sample data/extra'
ex df = df.tail(7)
df = df.iloc[:len(df)-len(ex df)]
combined_df = pd.concat([df, df_1, ex_df], ignore_index=True)
combined df['image paths'] = combined df['image names'].apply(lambda x
: image data dir + '/' + x if x in df['image names'].values else
(aug data dir + x if x in df 1['image names'].values else ex dir + x))
combined df
{"summary":"{\n \"name\": \"combined df\",\n \"rows\": 2722,\n
\"num unique values\": 2722,\n
                                   \"samples\": [\n
           ,\n \"408.jpg\",\n \"semantic_type\": \"\",\n
\"1061.jpg\",\n
                                              \"aug 81.jpg\"\n
                                           \"description\": \"\"\n
],\n
              {\n \"column\": \"emergency_or_not\",\n
}\n
      },\n
                       \"dtype\": \"number\\\",\n
\"properties\": {\n
                                                        \"std\":
                             \"max\": 1,\n
0,\n
           \"min\": 0,\n
\"num unique values\": 2,\n \"samples\": [\n
                                                          0, n
          ],\n \"semantic type\": \"\",\n
\"description\": \"\"\n
                          }\n },\n {\n
                                                  \"column\":
\"image paths\",\n
                   \"properties\": {\n
                                                \"dtvpe\":
\"string\",\n \"num
\"samples\": [\n
                   \"num unique values\": 2722,\n
                        \"/content/sample data/Images/1061.jpg\",\n
\"/content/sample data/Images/408.jpg\"\n
\"semantic type\": \"\",\n \"description\": \"\"\n
    }\n ]\n}","type":"dataframe","variable_name":"combined_df"}
n
combined df = combined df.sample(2722) # Reshuffling my dataframe
combined df['emergency or not'] =
combined df['emergency or not'].astype('str')
len(combined df)
2722
combined_df['emergency_or_not'].value_counts()
```

```
emergency_or_not
     1361
1
     1361
Name: count, dtype: int64
train data = combined df.sample(frac = 0.8, random state = 42)
test data = combined df.drop(train data.index)
# Create ImageDataGenerator for training data (with augmentations)
train datagen = ImageDataGenerator(rescale=1./255, # Rescale pixel
values to be between 0 and 1
    rotation_range=20,
    width shift range=0.2,
    height_shift_range=0.2,
    shear_range=0.2,
    zoom range=0.2,
    horizontal flip=True,
    fill mode='nearest'
)
# Create ImageDataGenerator for test data (only rescaling, no
augmentations)
test datagen = ImageDataGenerator(rescale=1./255)
# Create the training data generator
train generator = train datagen.flow from dataframe(train data,
directory = None,
                                                     x col =
'image paths',
                                                     y col =
'emergency or not',
                                                     target size =
(224, 224),
                                                     batch size = 32,
                                                     class mode =
'binary')
#Create the test data generator
test_generator = test_datagen.flow_from_dataframe(test data,
                                                   directory = None,
                                                   x_col =
'image paths',
                                                   y col =
'emergency or not',
                                                   batch size = 32,
                                                   target size =
(224,224),
                                                   class mode =
'binary')
```

Found 1875 validated image filenames belonging to 2 classes. Found 463 validated image filenames belonging to 2 classes.

#VGG16

```
model = Sequential()
#Block 1
model.add(Conv2D(32, kernel size = (3,3), activation = 'relu',
input shape = (224, 224, 3))
model.add(BatchNormalization())
model.add(Conv2D(32, kernel size = (3,3), activation = 'relu'))
model.add(BatchNormalization())
model.add(MaxPooling2D(pool size = (2,2)))
model.add(Dropout(0.25))
#Block 2
model.add(Conv2D(64, kernel size = (3,3), activation = 'relu'))
model.add(BatchNormalization())
model.add(Conv2D(64, kernel_size = (3,3), activation = 'relu'))
model.add(BatchNormalization())
model.add(MaxPooling2D(pool size = (2,2)))
model.add(Dropout(0.25))
#Block 3
model.add(Conv2D(128, kernel size = (3,3), activation = 'relu'))
model.add(BatchNormalization())
model.add(Conv2D(128, kernel size = (3,3), activation = 'relu'))
model.add(BatchNormalization())
model.add(MaxPooling2D(pool size = (2,2)))
model.add(Dropout(0.25))
#Flatten layer
model.add(Flatten())
#Dense layers
model.add(Dense(128, activation = 'relu', kernel regularizer =
keras.regularizers.l2(0.01)))
model.add(Dropout(0.5))
model.add(Dense(1, activation = 'sigmoid'))
model.summary()
```

| Model: "sequential_2" | |
|---|----------------------|
| Layer (type) Param # | Output Shape |
| conv2d_13 (Conv2D) 896 | (None, 222, 222, 32) |
| batch_normalization_13 128 (BatchNormalization) | (None, 222, 222, 32) |
| conv2d_14 (Conv2D) 9,248 | (None, 220, 220, 32) |
| batch_normalization_14 128 (BatchNormalization) | (None, 220, 220, 32) |
| max_pooling2d_7 (MaxPooling2D) | (None, 110, 110, 32) |
| dropout_9 (Dropout) | (None, 110, 110, 32) |
| conv2d_15 (Conv2D) 18,496 | (None, 108, 108, 64) |
| batch_normalization_15 256 (BatchNormalization) | (None, 108, 108, 64) |
| conv2d_16 (Conv2D) 36,928 | (None, 106, 106, 64) |
| | |

```
batch normalization 16
                                       (None, 106, 106, 64)
256 l
 (BatchNormalization)
 max pooling2d 8 (MaxPooling2D)
                                       (None, 53, 53, 64)
 dropout_10 (Dropout)
                                       (None, 53, 53, 64)
 conv2d_17 (Conv2D)
                                       (None, 51, 51, 128)
73,856
 batch normalization 17
                                       (None, 51, 51, 128)
 (BatchNormalization)
 conv2d 18 (Conv2D)
                                       (None, 49, 49, 128)
147,584
 batch normalization 18
                                       (None, 49, 49, 128)
 (BatchNormalization)
 max pooling2d 9 (MaxPooling2D)
                                       (None, 24, 24, 128)
 dropout 11 (Dropout)
                                       (None, 24, 24, 128)
 flatten_2 (Flatten)
                                       (None, 73728)
0 l
dense 4 (Dense)
                                       (None, 128)
9,437,312
```

```
dropout 12 (Dropout)
                                       (None, 128)
0
dense 5 (Dense)
                                       (None, 1)
129
Total params: 9,726,241 (37.10 MB)
Trainable params: 9,725,345 (37.10 MB)
Non-trainable params: 896 (3.50 KB)
model.compile(optimizer = 'adam',
             loss = 'binary_crossentropy',
             metrics = ['accuracy'])
history = model.fit(train generator, epochs = 5, batch size = 64,
validation data = test generator)
Epoch 1/5
                  _____ 52s 656ms/step - accuracy: 0.5377 - loss:
59/59 —
13.6902 - val accuracy: 0.5940 - val loss: 6.7442
Epoch 2/5
                  _____ 25s 385ms/step - accuracy: 0.5808 - loss:
59/59 -
5.8600 - val accuracy: 0.4060 - val loss: 18.3813
Epoch 3/5
               ______ 25s 383ms/step - accuracy: 0.6506 - loss:
59/59 ———
4.0446 - val accuracy: 0.5940 - val loss: 3.4718
Epoch 4/5
                 _____ 25s 383ms/step - accuracy: 0.6518 - loss:
59/59 ———
3.0382 - val accuracy: 0.4752 - val loss: 3.6349
Epoch 5/5
                  _____ 25s 382ms/step - accuracy: 0.6793 - loss:
59/59 -
2.2889 - val accuracy: 0.5940 - val_loss: 2.2644
history = model.fit(train generator, epochs = 5, batch size = 64,
validation data = test generator)
Epoch 1/5
             ______ 25s 365ms/step - accuracy: 0.7042 - loss:
1.7871 - val accuracy: 0.5961 - val loss: 2.2076
Epoch 2/5
                      —— 40s 373ms/step - accuracy: 0.7117 - loss:
59/59 —
1.5605 - val accuracy: 0.5940 - val loss: 1.6798
Epoch 3/5
                   _____ 25s 393ms/step - accuracy: 0.7357 - loss:
59/59 —
1.3938 - val accuracy: 0.6069 - val loss: 1.5436
```

```
Epoch 4/5
            41s 392ms/step - accuracy: 0.7477 - loss:
59/59 -
1.1556 - val accuracy: 0.6933 - val loss: 1.1716
Epoch 5/5
          ______ 25s 380ms/step - accuracy: 0.7445 - loss:
59/59 ----
1.0020 - val accuracy: 0.7646 - val loss: 1.0005
history = model.fit(train generator, epochs = \frac{5}{100}, batch size = \frac{64}{1000},
validation data = test generator)
Epoch 1/5
59/59 ______ 24s 375ms/step - accuracy: 0.7697 - loss:
1.0497 - val accuracy: 0.7927 - val loss: 1.0564
Epoch 2/5
59/59 ————
               ______ 24s 366ms/step - accuracy: 0.7689 - loss:
1.0638 - val accuracy: 0.7689 - val loss: 1.1063
Epoch 3/5
59/59 ————— 24s 361ms/step - accuracy: 0.7572 - loss:
1.0632 - val_accuracy: 0.7192 - val loss: 1.1973
Epoch 4/5
                     25s 365ms/step - accuracy: 0.7522 - loss:
59/59 —
1.2554 - val accuracy: 0.7711 - val loss: 1.3010
Epoch 5/5
                  40s 373ms/step - accuracy: 0.7693 - loss:
59/59 —
1.1286 - val accuracy: 0.7495 - val loss: 1.3508
```

ALexnet

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import warnings
warnings.filterwarnings('ignore')
import os
import tensorflow as tf
from tensorflow import keras
from keras import Sequential
from keras.layers import Dense, Flatten, Dropout, Conv2D,
MaxPooling2D, BatchNormalization
from tensorflow.keras.preprocessing.image import ImageDataGenerator,
load img, save img, img to array, array to img
df = pd.read csv('/content/emergency classification.csv')
df.head()
{"summary":"{\n \"name\": \"df\",\n \"rows\": 2352,\n \"fields\":
[\n {\n \"column\": \"image names\",\n \"properties\": {\
```

```
\"dtype\": \"string\",\n
                                       \"num unique values\": 2352,\
        \"samples\": [\n
                                 \"1960.jpg\",\n
n
\"668.jpg\",\n
                       \"2082.jpg\"\n
                                             ],\n
\"semantic type\": \"\",\n \"description\": \"\"\n
                                                             }\
    },\n {\n \"column\": \"emergency_or_not\",\n
\"properties\": {\n \"dtype\": \"number\",\n
                                                         \"std\":
           \"min\": 0,\n
                               \"max\": 1,\n
0,\n
\"num_unique_values\": 2,\n
                                  \"samples\": [\n
                                                           0.\n
                   \"semantic type\": \"\",\n
          ],\n
\"description\": \"\"\n
                           n}","type":"dataframe","variable_name":"df"}
df['emergency or not'].value counts()
emergency or not
     1361
1
      991
Name: count, dtype: int64
data dir = '/content/sample data/Images'
len(os.listdir(data dir))
2352
aug dir = '/content/sample data/aug images'
os.makedirs(aug dir, exist ok = True)
datagen = ImageDataGenerator(rotation range=0.4,
width shift range=0.25,
                            height shift range=0.12,
                            shear range=0.21, zoom range=0.38,
horizontal flip=True,
                            vertical flip=True, rescale=1./255,
minority class = df['emergency or not'].value counts().min()
majority class count = df['emergency or not'].value counts().max()
num augmented images = majority class count - minority class
num augmented images
370
minority images = df[df['emergency or not'] == 1]['image names']
current count = 0
for image name in minority images:
  if current count >= num augmented images:
   break
  img path = os.path.join(data dir, image name)
  img = load img(img_path)
```

```
img_array = img to array(imq)
  img array = np.expand dims(img array, axis = 0)
  for in range(num augmented images - current count):
    aug image = next(datagen.flow(img array, batch size = 1))
[0].astype('uint8')
    save_img(os.path.join(aug_dir, f'aug_{current_count}.jpg'),
aug image)
    current count += 1
    if current count >= num augmented images:
       break
print(f'Generated {num augmented images} augmented images.')
Generated 370 augmented images.
data_dir = '/content/sample_data/Images' aug_dir = '/content/sample_data/aug_images'
os.makedirs(aug_dir, exist_ok = True)
minority_images = df[df['emergency_or_not' == 1]]['image_names']
minority_class_count = df['emergency_or_not'].value_counts().min() majority_class_count =
df['emergency_or_not'].value_counts().max() num_augmented_images = minority_Class_count
- majority_class_count
current_count = 0 for image_file in minority_images: if current_count >=
num_augmented_images: break
image_path = os.path.join(data_dir, image_file) img = img.load(image_path) img_array =
img_to_array(img) img_array = np.expand_dims(img_array, axis = 0)
for _ in range(num_augmented_images - current_count): aug_image =
next(datagen.flow(img_array, batch_size = 1))[0].astype('uint8')
img.save(os.path.join(output dir, f'aug {current count}.jpg'),aug image) current count += 1
if current count >= num augmented images:
  break
print(f'Generated {num_augmented_images} images')
```

```
\"samples\": [\n \"aug_317.jpg\",\n \"aug_9.j
\"aug_70.jpg\"\n ],\n \"semantic_type\": \"\",\n
                                                       \"aug 9.jpg\",\n
\"description\": \"\"\n }\n },\n {\n \"emergency_or_not\",\n \"properties\": {\n
                                                      \"column\":
                                                          \"dtvpe\":
\"number\",\n \"std\": 0,\n \"min\": 1,\n \"max\": 1,\n \"num_unique_values\": 1,\n \"samples\": [\n 1\n ],\n \"semantic_type\": \"\",\n
\"description\": \"\"\n }\n
                                     }\n ]\
n}","type":"dataframe","variable name":"df 1"}
combined_df = pd.concat([df, df_1], ignore_index = True)
combined df
{"summary":"{\n \"name\": \"combined df\",\n \"rows\": 2722,\n
\"fields\": [\n \\"column\\\\": \\"image_names\\\\,\n
                            \"dtype\": \"string\\\",\n
\"properties\": {\n
\"num_unique_values\": 2722,\n \"samples\": [\n
\"1061.jpg\",\n \"408.jpg\",\n \"aug_94.jpg\"\n \",\n \"description\": \"\"\n
}\n },\n {\n \"column\": \"emergency_or_not\",\n \"properties\": {\n \"dtype\": \"number\",\n \"
                         \"dtype\": \"number\",\n \"std\":
0,\n \"min\": 0,\n \"max\": 1,\n \"num_unique_values\": 2,\n \"samples\": [\n
                                                                0, n
1\n ],\n \"semantic_type\": \"\",\n
\"description\": \"\\n }\n ]\
n}","type":"dataframe","variable name":"combined df"}
combined df['emergency or not'].value counts()
emergency_or_not
     1361
1
0
     1361
Name: count, dtype: int64
combined df['image paths'] = combined df['image names'].apply(lambda x
: data dir + '/' + x if x in df['image names'].values else aug dir +
'/' + x
combined df
{"summary":"{\n \"name\": \"combined df\",\n \"rows\": 2722,\n
\"num unique values\": 2722,\n \"samples\": [\n
\"1061.jpg\",\n \"408.jpg\",\n \"aug_94.jpg\"\n \],\n \"semantic_type\": \"\",\n \"description\": \"\"\n
       },\n {\n \"column\": \"emergency_or_not\",\n
}\n
\"properties\": {\n \"dtype\": \"number\",\n \"std\":
0,\n \"min\": 0,\n \"max\": 1,\n \"num_unique_values\": 2,\n \"samples\": [\n
                                                                0, n
1\n ],\n \"semantic_type\": \"\",\n
\"description\": \"\"\n }\n },\n
                                                       \"column\":
```

```
\"image paths\",\n
                       \"properties\": {\n
                                                   \"dtvpe\":
\"string\",\n \"n
\"samples\": [\n
                     \"num unique values\": 2722,\n
                          \"/content/sample data/Images/1061.jpg\",\n
\"/content/sample data/Images/408.jpg\"\n
\"semantic type\": \"\",\n \"description\": \"\"\n
     }\n ]\n}","type":"dataframe","variable_name":"combined_df"}
combined df = combined df.sample(2722)
combined df['emergency or not'] =
combined df['emergency or not'].astype('str')
train data = combined df.sample(frac = 0.8, random state = 42)
test data = combined df.drop(train data.index)
train datagen = ImageDataGenerator(rotation range=0.4,
width shift range=0.25,
                             height shift range=0.12,
                             shear range=0.21, zoom range=0.38,
horizontal flip=True,
                             vertical flip=True, rescale=1./255)
test datagen = ImageDataGenerator(rescale = 1./255)
train generator = train datagen.flow from dataframe(train data,
                                                     directory = None,
                                                     x col =
'image paths',
                                                     y col =
'emergency or not',
                                                     target size =
(224, 224),
                                                     batch size = 32,
                                                     class mode =
'binary')
test generator = test datagen.flow from dataframe(test data,
                                                   directory = None,
                                                   x col =
'image paths',
                                                   y col =
'emergency or not',
                                                   target_size =
(224, 224),
                                                   batch size = 32,
                                                   class mode =
'binary')
Found 2178 validated image filenames belonging to 2 classes.
Found 544 validated image filenames belonging to 2 classes.
```

```
# ALexnet model building
model = Sequential()
model.add(Conv2D(32, kernel size = (5,5), strides = 2, activation =
'relu', kernel regularizer=keras.regularizers.l2(0.01), input shape =
(224,224,3)))
model.add(BatchNormalization())
model.add(MaxPooling2D(pool size = (2,2), strides = 2))
model.add(Dropout(0.25))
model.add(Conv2D(64, kernel size = (3,3), padding = 'same', activation')
= 'relu'))
model.add(BatchNormalization())
model.add(MaxPooling2D(pool size = (2,2), strides = 2))
model.add(Dropout(0.25))
model.add(Conv2D(128, kernel size = (3,3), padding = 'same',
activation = 'relu'))
model.add(BatchNormalization())
model.add(Conv2D(128, kernel_size = (3,3), padding = 'same',
activation = 'relu'))
model.add(BatchNormalization())
model.add(Conv2D(128, kernel size = (3,3), padding = 'same',
activation = 'relu'))
model.add(BatchNormalization())
model.add(MaxPooling2D(pool size = (2,2), strides = 2))
model.add(Dropout(0.30))
model.add(Flatten())
model.add(Dense(128, activation = 'relu', kernel regularizer =
keras.regularizers.l2(0.01)))
print(model.output shape)
model.add(Dropout(0.5))
model.add(Dense(1, activation = 'sigmoid'))
model.summary()
(None, 128)
Model: "sequential 2"
Layer (type)
                                        Output Shape
Param # |
 conv2d 10 (Conv2D)
                                         (None, 110, 110, 32)
2,432
```

```
batch normalization 10
                                       (None, 110, 110, 32)
128
  (BatchNormalization)
 max pooling2d 6 (MaxPooling2D)
                                       (None, 55, 55, 32)
 dropout_8 (Dropout)
                                       (None, 55, 55, 32)
 conv2d_11 (Conv2D)
                                       (None, 55, 55, 64)
18,496
 batch normalization 11
                                        (None, 55, 55, 64)
  (BatchNormalization)
 max pooling2d 7 (MaxPooling2D)
                                       (None, 27, 27, 64)
0
 dropout 9 (Dropout)
                                       (None, 27, 27, 64)
 conv2d 12 (Conv2D)
                                       (None, 27, 27, 128)
73,856
 batch normalization 12
                                       (None, 27, 27, 128)
512 l
  (BatchNormalization)
 conv2d_13 (Conv2D)
                                       (None, 27, 27, 128)
147,584
 batch normalization 13
                                        (None, 27, 27, 128)
512
 (BatchNormalization)
```

```
conv2d 14 (Conv2D)
                                        (None, 27, 27, 128)
147,584
  batch normalization 14
                                         (None, 27, 27, 128)
512
  (BatchNormalization)
 max pooling2d 8 (MaxPooling2D)
                                         (None, 13, 13, 128)
 dropout_10 (Dropout)
                                        (None, 13, 13, 128)
0
 flatten_2 (Flatten)
                                         (None, 21632)
0
 dense 4 (Dense)
                                        (None, 128)
2,769,0\overline{2}4
 dropout_11 (Dropout)
                                         (None, 128)
0
 dense 5 (Dense)
                                         (None, 1)
129
Total params: 3,161,025 (12.06 MB)
Trainable params: 3,160,065 (12.05 MB)
Non-trainable params: 960 (3.75 KB)
model.compile(optimizer = 'adam',
              loss = 'binary_crossentropy',
              metrics = ['accuracy'])
history = model.fit(train generator, epochs = 5, batch size = 64,
validation data = test generator)
```

```
Epoch 1/5
69/69 33s 441ms/step - accuracy: 0.6092 - loss:
6.0481 - val accuracy: 0.5276 - val loss: 3.0425
2.7357 - val accuracy: 0.5276 - val loss: 3.2858
Epoch 3/5
69/69 40s 345ms/step - accuracy: 0.6683 - loss:
2.1260 - val accuracy: 0.5276 - val loss: 2.4026
Epoch 4/5
1.7314 - val_accuracy: 0.5276 - val_loss: 2.0263
Epoch 5/5
            _____ 25s 330ms/step - accuracy: 0.7148 - loss:
69/69 ----
1.3863 - val accuracy: 0.6618 - val loss: 1.4901
history = model.fit(train generator, epochs = 5, validation data =
test generator)
Epoch 1/5
69/69 _______ 26s 333ms/step - accuracy: 0.7397 - loss:
1.1060 - val accuracy: 0.5662 - val loss: 1.2707
Epoch 2/5
60/60 40s 334ms/step - accuracy: 0.7187 - loss:
1.2497 - val accuracy: 0.5625 - val loss: 2.0264
1.3348 - val accuracy: 0.5331 - val loss: 2.2575
1.0142 - val accuracy: 0.6746 - val loss: 1.4489
0.8914 - val accuracy: 0.6618 - val loss: 1.6831
```