Image Classification

- Import the necessry libraries
- Preprocessing on the datastet
 - Traing and test data generators

In [114]:

tf.keras.preprocessing.image.load_img(r"C:\Users\younu\Desktop\My Py Sc

Out[114]:



Out[115]:

	iiiiaye_iiaiiies	emergency_or_not
0	0.jpg	1
1	1.jpg	1
2	2.jpg	1
3	3.jpg	1
4	4.jpg	1

image names emergency or not

```
In [116]:
             1 | df['emergency_or_not'].value_counts()
Out[116]: emergency_or_not
           0
                 1361
           1
                  991
           Name: count, dtype: int64
In [117]:
             1 df.tail()
Out[117]:
                  image_names emergency_or_not
            2347
                                             0
                      2347.jpg
            2348
                      2348.jpg
                                             0
            2349
                      2349.jpg
                                             0
            2350
                      2350.jpg
                                             0
                                             0
            2351
                      2351.jpg
In [118]:
                train_df = df.groupby('emergency_or_not', group_keys = False).apply(lam
                test_df = df.drop(train_df.index)
               train_df.head()
Out[118]:
                  image_names emergency_or_not
            2126
                      2126.jpg
                                             0
            1648
                      1648.jpg
                                             0
            2021
                                             0
                      2021.jpg
            1911
                      1911.jpg
                                             0
            1790
                      1790.jpg
                                             0
In [119]:
                test_df.head()
Out[119]:
                image_names emergency_or_not
             9
                                            1
                       9.jpg
            11
                                            1
                      11.jpg
            19
                      19.jpg
            23
                      23.jpg
            28
                      28.jpg
                                            1
In [120]:
                print(train_df.shape, test_df.shape)
           (1882, 2) (470, 2)
In [121]:
             1 test_df['emergency_or_not'].value_counts()
Out[121]:
           emergency_or_not
                 272
                 198
           Name: count, dtype: int64
```

```
In [122]:
               test_df = df.groupby('emergency_or_not', group_keys = False).apply(lamb
            2
               test_df.head()
Out[122]:
                 image_names emergency_or_not
           2126
                     2126.jpg
                                           0
           1648
                                           0
                     1648.jpg
           2021
                     2021.jpg
                                           0
           1911
                     1911.jpg
                                           0
           1790
                     1790.jpg
                                           0
In [123]:
               train_df['emergency_or_not'] = train_df['emergency_or_not'].astype(str)
               test_df['emergency_or_not'] = test_df['emergency_or_not'].astype(str)
In [124]:
               print(train_df.shape, test_df.shape)
           (1882, 2) (470, 2)
In [125]:
            1 test_df['emergency_or_not'].value_counts()
Out[125]: emergency_or_not
                272
          0
           1
                198
          Name: count, dtype: int64
               training_data_generator = ImageDataGenerator(rotation_range = 30,
In [126]:
            2
                                                        width_shift_range = 0.2,
            3
                                                        height_shift_range = 0.2,
            4
                                                        shear_range = 0.2,
            5
                                                        zoom_range = 0.3,
                                                        channel_shift_range = 0.2,
            6
            7
                                                        fill_mode = 'nearest',
            8
                                                        horizontal_flip = True,
            9
                                                        vertical_flip = True,
           10
                                                        rescale = 1./255)
           11
               testing data generator = ImageDataGenerator(rescale = 1./255)
In [127]:
               folder path = r"C:\Users\younu\Desktop\My Py Scripts\Git Repos\11 Emerg
               train_generator = training_data_generator.flow_from_dataframe(train_df,
In [128]:
            1
            2
                                                                               x col = 'i
            3
                                                                               target_siz
           Found 1882 validated image filenames belonging to 2 classes.
In [129]:
               test_generator = testing_data_generator.flow_from_dataframe(test_df, di
                                                                               x_{col} = 'i
            2
            3
                                                                               target_siz
```

Found 470 validated image filenames belonging to 2 classes.

```
In [130]: 1 vgg16_model = VGG16(include_top = False, weights = 'imagenet', input_sh
2 vgg16_model.trainable = False # We give false to not change the actual

In [132]: 1 output = vgg16_model.layers[-1].output # -1 to access the last layer (
2 flatten = Flatten()(output)
```

Model: "functional_1"

Layer (type)	Output Shape	Param #
<pre>input_layer_1 (InputLayer)</pre>	(None, 200, 200, 3)	0
block1_conv1 (Conv2D)	(None, 200, 200, 64)	1,792
block1_conv2 (Conv2D)	(None, 200, 200, 64)	36,928
block1_pool (MaxPooling2D)	(None, 100, 100, 64)	0
block2_conv1 (Conv2D)	(None, 100, 100, 128)	73,856
block2_conv2 (Conv2D)	(None, 100, 100, 128)	147,584
block2_pool (MaxPooling2D)	(None, 50, 50, 128)	0
block3_conv1 (Conv2D)	(None, 50, 50, 256)	295,168
block3_conv2 (Conv2D)	(None, 50, 50, 256)	590,080
block3_conv3 (Conv2D)	(None, 50, 50, 256)	590,080
block3_pool (MaxPooling2D)	(None, 25, 25, 256)	0
block4_conv1 (Conv2D)	(None, 25, 25, 512)	1,180,160
block4_conv2 (Conv2D)	(None, 25, 25, 512)	2,359,808
block4_conv3 (Conv2D)	(None, 25, 25, 512)	2,359,808
block4_pool (MaxPooling2D)	(None, 12, 12, 512)	0
block5_conv1 (Conv2D)	(None, 12, 12, 512)	2,359,808
block5_conv2 (Conv2D)	(None, 12, 12, 512)	2,359,808
block5_conv3 (Conv2D)	(None, 12, 12, 512)	2,359,808
block5_pool (MaxPooling2D)	(None, 6, 6, 512)	0
flatten_2 (Flatten)	(None, 18432)	0
dense_4 (Dense)	(None, 512)	9,437,696
dense_5 (Dense)	(None, 512)	262,656
emergency_or_not (Dense)	(None, 1)	513

Total params: 24,415,553 (93.14 MB)

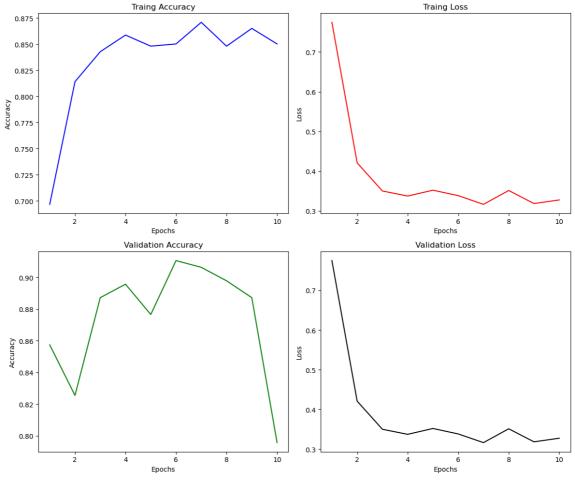
Trainable params: 9,700,865 (37.01 MB)

Non-trainable params: 14,714,688 (56.13 MB)

```
In [134]:
           1 model.compile(optimizer = 'adam', loss = 'binary_crossentropy', metrics
In [135]:
           1 history = model.fit(train_generator, batch_size = 32, epochs = 10, vali
          Epoch 1/10
          C:\Users\younu\anaconda3\Lib\site-packages\keras\src\trainers\data_adapter
          s\py_dataset_adapter.py:121: UserWarning: Your `PyDataset` class should ca
          11 `super().__init__(**kwargs)` in its constructor. `**kwargs` can include
          `workers`, `use_multiprocessing`, `max_queue_size`. Do not pass these argu
          ments to `fit()`, as they will be ignored.
            self._warn_if_super_not_called()
                            337s 6s/step - accuracy: 0.6189 - loss: 1.2496
          - val_accuracy: 0.8574 - val_loss: 0.3211
          Epoch 2/10
          59/59 -----
                          483s 8s/step - accuracy: 0.8109 - loss: 0.4330
          - val_accuracy: 0.8255 - val_loss: 0.3859
          Epoch 3/10
                                  - 424s 7s/step - accuracy: 0.8478 - loss: 0.3490
          - val_accuracy: 0.8872 - val_loss: 0.2832
          Epoch 4/10
          59/59 -
                                  - 407s 7s/step - accuracy: 0.8543 - loss: 0.3476
          - val_accuracy: 0.8957 - val_loss: 0.2506
          59/59 407s 7s/step - accuracy: 0.8610 - loss: 0.3285
          - val accuracy: 0.8766 - val loss: 0.2838
          Epoch 6/10
                                 - 392s 7s/step - accuracy: 0.8596 - loss: 0.3147
          59/59 -
          - val_accuracy: 0.9106 - val_loss: 0.2271
          Epoch 7/10
          59/59 -
                                  - 411s 7s/step - accuracy: 0.8845 - loss: 0.2983
          - val_accuracy: 0.9064 - val_loss: 0.2154
          Epoch 8/10
                            444s 7s/step - accuracy: 0.8594 - loss: 0.3289
          59/59 ----
          - val_accuracy: 0.8979 - val_loss: 0.2255
          Epoch 9/10
                             407s 7s/step - accuracy: 0.8542 - loss: 0.3200
          59/59 ----
          - val accuracy: 0.8872 - val loss: 0.2738
          Epoch 10/10
          59/59 -
                                 - 381s 6s/step - accuracy: 0.8543 - loss: 0.3266
          - val_accuracy: 0.7957 - val_loss: 0.4908
In [136]:
          1 history.history.keys()
```

Out[136]: dict_keys(['accuracy', 'loss', 'val_accuracy', 'val_loss'])

```
In [137]:
              x = np.linspace(1, 10, 10)
              fig, axs = plt.subplots(2, 2, figsize = (12, 10))
              axs[0, 0].plot(x, history.history['accuracy'], 'b')
            5
              axs[0, 0].set_title('Traing Accuracy')
            7
               axs[0, 0].set_xlabel('Epochs')
              axs[0, 0].set_ylabel('Accuracy')
            8
            9
              axs[0, 1].plot(x, history.history['loss'], 'r')
           10
           11
              axs[0, 1].set_title('Traing Loss')
               axs[0, 1].set_xlabel('Epochs')
           12
           13
              axs[0, 1].set_ylabel('Loss')
           14
           15 | axs[1, 0].plot(x, history.history['val_accuracy'], 'g')
           16
              axs[1, 0].set_title('Validation Accuracy')
              axs[1, 0].set_xlabel('Epochs')
           17
              axs[1, 0].set_ylabel('Accuracy')
           18
           19
              axs[1, 1].plot(x, history.history['loss'], 'k')
           20
           21
              axs[1, 1].set_title('Validation Loss')
              axs[1, 1].set_xlabel('Epochs')
           22
           23
              axs[1, 1].set_ylabel('Loss')
           24
           25
              plt.tight_layout()
           26
              plt.show()
```



```
In [139]:
     # Checking the predicton probability of each class in test dataset
    1
     predictions = model.predict(test_generator, verbose = 1)
     predictions
   15/15 •
            74s 5s/step
Out[139]: array([[0.99331623],
      [0.98817325],
      [0.08325341],
      [0.07799871],
      [0.9643081],
      [0.9911224],
      [0.6962938],
      [0.580725],
      [0.03578774],
      [0.9861002],
      [0.9969041],
      [0.99850017],
      [0.9985583],
      [0.9384076],
      [0.50707495],
      [0.8069996],
      [0.0292578],
      [0 00005806]
In [140]:
     # Converting the predictions into labels
    2
    3
     predicted_classes = np.argmax(predictions, axis = 1)
    4
     predicted_classes
0, 0, 0, 0, 0, 0, 0], dtype=int64)
In [141]:
     true_classes = test_generator.labels
In [142]:
     accuracy_score(true_classes, predicted_classes)
Out[142]: 0.5787234042553191
```

	precision	recall	f1-score	support
0	0.58	1.00	0.73	272
1	0.00	0.00	0.00	198
accuracy			0.58	470
macro avg	0.29	0.50	0.37	470
weighted avg	0.33	0.58	0.42	470

C:\Users\younu\anaconda3\Lib\site-packages\sklearn\metrics_classificatio n.py:1344: UndefinedMetricWarning: Precision and F-score are ill-defined a nd being set to 0.0 in labels with no predicted samples. Use `zero_divisio n` parameter to control this behavior.

_warn_prf(average, modifier, msg_start, len(result))

C:\Users\younu\anaconda3\Lib\site-packages\sklearn\metrics_classificatio n.py:1344: UndefinedMetricWarning: Precision and F-score are ill-defined a nd being set to 0.0 in labels with no predicted samples. Use `zero_divisio n` parameter to control this behavior.

_warn_prf(average, modifier, msg_start, len(result))

C:\Users\younu\anaconda3\Lib\site-packages\sklearn\metrics_classificatio n.py:1344: UndefinedMetricWarning: Precision and F-score are ill-defined a nd being set to 0.0 in labels with no predicted samples. Use `zero_divisio n` parameter to control this behavior.

_warn_prf(average, modifier, msg_start, len(result))

```
In [144]: 1 test_generator.class_indices
```

Out[144]: {'0': 0, '1': 1}

```
confu_matrix = confusion_matrix(true_classes, predictions)
In [145]:
              ns.heatmap(confu_matrix, annot=True, fmt='d', cmap='Blues', xticklabels
                        _____
          ValueError
                                                    Traceback (most recent call las
          t)
          Cell In[145], line 1
          ----> 1 confu_matrix = confusion_matrix(true_classes, predictions)
                2 ns.heatmap(confu_matrix, annot=True, fmt='d', cmap='Blues', xtickl
          abels=test_generator.class_indices.keys(), yticklabels=test_generator.clas
          s_indices.keys())
          File ~\anaconda3\Lib\site-packages\sklearn\metrics\_classification.py:317,
          in confusion_matrix(y_true, y_pred, labels, sample_weight, normalize)
              232 def confusion_matrix(
                      y_true, y_pred, *, labels=None, sample_weight=None, normalize=
              233
          None
              234 ):
                      """Compute confusion matrix to evaluate the accuracy of a clas
              235
          sification.
              236
              237
                      By definition a confusion matrix :math:`C` is such that :math:
          `C_{i, j}`
             (\ldots)
              315
                      (0, 2, 1, 1)
              316
                      y_type, y_true, y_pred = _check_targets(y_true, y_pred)
          --> 317
                      if y_type not in ("binary", "multiclass"):
              318
              319
                          raise ValueError("%s is not supported" % y_type)
          File ~\anaconda3\Lib\site-packages\sklearn\metrics\_classification.py:95,
          in _check_targets(y_true, y_pred)
                      y_type = {"multiclass"}
               92
               94 if len(y_type) > 1:
                      raise ValueError(
          ---> 95
                          "Classification metrics can't handle a mix of {0} and {1}
               96
          targets".format(
               97
                              type_true, type_pred
               98
                          )
               99
              101 # We can't have more than one value on y type => The set is no mor
          e needed
              102 y_type = y_type.pop()
          ValueError: Classification metrics can't handle a mix of binary and contin
          uous targets
```