Import packages

```
In [135]:
          import numpy as np
          import random
          import pandas as pd
          import matplotlib.pyplot as plt
          %matplotlib inline
          import tensorflow
          import os
          import glob
          import time
          from PIL import Image
          from tgdm import tgdm
          from sklearn.utils import shuffle
          from tensorflow.keras.preprocessing.image import ImageDataGenerator
          from tensorflow.keras.models import Sequential
          from tensorflow.keras.layers import InputLayer, Conv2D, BatchNormalizati
          on, MaxPool2D, Dropout, Flatten, Dense
          from tensorflow.keras.callbacks import EarlyStopping, ModelCheckpoint, R
          educeLROnPlateau
          from sklearn.model selection import train test split, GridSearchCV
          from sklearn.metrics import confusion_matrix, classification_report, acc
          uracy score
          from sklearn.decomposition import PCA #Principal Component Analysis
          from sklearn.discriminant analysis import LinearDiscriminantAnalysis
          from sklearn.ensemble import RandomForestClassifier
          from sklearn.linear model import LogisticRegressionCV, LogisticRegressio
          from tensorflow.keras.applications.vgg16 import VGG16
          import tensorflow.keras.applications.vgg16 as vgg16
          from tensorflow.keras.applications.densenet import DenseNet121
          import tensorflow.keras.applications.densenet as densenet
          from tensorflow.keras.applications.xception import Xception
          import tensorflow.keras.applications.xception as xception
          from tensorflow.keras.layers import Dense, Flatten
          from tensorflow.keras.models import Model
          from tensorflow.keras.optimizers import SGD, Adam
          import matplotlib.pyplot as plt
          import time
```

Load data

```
# reference: https://www.kaggle.com/gauravrajpal/leukemia-classification
In [4]:
        -v1-2-xception-66-52
        # load training data
        train 0 all = glob.glob('cs156 final/training data/fold 0/all/*.bmp')
        train 0 hem = glob.glob('cs156 final/training data/fold 0/hem/*.bmp')
        train 1 all = glob.glob('cs156 final/training data/fold 1/all/*.bmp')
        train 1 hem = glob.glob('cs156 final/training data/fold 1/hem/*.bmp')
        train 2 all = glob.glob('cs156 final/training data/fold 2/all/*.bmp')
        train 2 hem = glob.glob('cs156 final/training_data/fold_2/hem/*.bmp')
In [5]: # load labels for test data
        valid labels = pd.read csv('cs156 final/validation data/C-NMC test preli
        m phase data labels.csv')
        valid labels.head(10)
Out[5]:
                     Patient_ID new_names labels
         0
              UID_57_29_1_all.bmp
                                  1.bmp
                                           1
              UID_57_22_2_all.bmp
                                  2.bmp
                                           1
              UID_57_31_3_all.bmp
                                  3.bmp
         3 UID_H49_35_1_hem.bmp
                                  4.bmp
                                           0
              UID_58_6_13_all.bmp
                                  5.bmp
              UID_57_8_11_all.bmp
                                  6.bmp
         6 UID_H49_29_2_hem.bmp
                                  7.bmp
                                           0
            UID_H30_6_2_hem.bmp
                                  8.bmp
                                           0
               UID_58_2_1_all.bmp
                                  9.bmp
                                           1
         8
              UID 54 35 3 all.bmp
                                 10.bmp
                                           1
In [7]: # stack all training data
        train_all = np.hstack((np.array(train_0_all), np.array(train_1_all), np.
        array(train 2 all)))
        train_hem = np.hstack((np.array(train_0_hem), np.array(train_1_hem), np.
        array(train 2 hem)))
In [8]: # read and convert images to numpy array for healthy
        train hem np = []
        for i in tqdm(range(len(train_hem))):
             train hem np.append(np.array(Image.open(train hem[i]).resize((100,10
        0))))
        100% | 3389/3389 [00:08<00:00, 419.36it/s]
In [9]: # read and convert images to numpy array for leukemia
        train all_np = []
        for i in tqdm(range(len(train all))):
             train all np.append(np.array(Image.open(train all[i]).resize((100,10
        0))))
        100% 7272/7272 [00:18<00:00, 398.90it/s]
```

```
# stack healthy and leukemia for training
In [10]:
          train_np = np.vstack((np.array(train_all_np), np.array(train_hem_np)))
          train label = np.vstack((np.ones((len(train all np),1)), np.zeros((len(t
          rain hem np), 1)))
          train np, train label = shuffle(train np, train label, random state=123)
In [77]: # check shape of training set
          train np.shape
Out[77]: (10661, 100, 100, 3)
In [87]: # proportion of class 1 in training
          train_label.sum() / len(train_label)
Out[87]: 0.68211237219773
In [152]: # number of class 1 and class 0 in training
          print('leukemia', 10661*0.68211237219773)
          print('healthy', 10661 - 10661*0.68211237219773)
          leukemia 7272.0
          healthy 3389.0
In [78]: # sample randomly 3000 samples from each class in training to train
          ind_bal = shuffle(np.hstack((np.random.choice(np.where(train_label == 1)
          [0], 3000), np.random.choice(np.where(train label == 0)[0], 3000))))
          X train = train np[ind bal]
          y_train = train_label[ind bal]
In [98]: # check out the images for healthy and leukemia
          fig, ax = plt.subplots(nrows = 1, ncols = 6, figsize = (20,20))
          for i in tqdm(range(6)):
              ind = np.random.randint(len(X train))
              ax[i].imshow(X_train[ind])
              if y train[ind] == 1:
                  ax[i].set_title('Leukemia')
              else:
                  ax[i].set_title('Healthy')
          100%
                    6/6 [00:00<00:00, 426.85it/s]
               Healthy
```

```
In [43]:
         # load the test set
         test_np = []
         test label = []
         for i in tqdm(range(valid_labels.shape[0])):
             tail = f'{i+1}.bmp'
             path = 'cs156 final/validation data/C-NMC test prelim phase data/' +
         tail
             test_np.append(np.array(Image.open(path).resize((100,100))))
             test label.append(valid labels[valid labels['new names'] == tail]['1
         abels'].iloc[0])
         test np = np.array(test np)
         test_label = np.array(test_label).reshape(-1,1)
         100% | 1867/1867 [00:07<00:00, 256.28it/s]
In [79]: # randomly sample 500 healthy and leukemia samples for test
         ind bal test = shuffle(np.hstack((np.random.choice(np.where(test label =
         = 1)[0], 500), np.random.choice(np.where(test label == 0)[0], 500))))
         X_test = test_np[ind_bal_test]
         y_test = test_label[ind_bal_test]
```

Build the models

Deep Learning

```
In [124]: # reference: https://www.kaggle.com/gauravrajpal/leukemia-classification
          -v1-2-xception-66-52
          # build the deep learning model with transfer learning
          def individual model(app, model name, path to model, X train, y train, X
          _test, y_test, min_delta = 0.00003, k = 0, epochs=20, batch_size=20, thr
          eshold=0.5, drop rate=0.3, class weight=None):
              # augment the images
              X train, X val, y train, y val = train test split(X train, y train,
          random_state=123, test_size=0.2)
              datagen = ImageDataGenerator(horizontal_flip=True,
                                               vertical flip=True,
                                               preprocessing function=app.preproces
          s input)
              datagen.fit(X train)
              # pre-process test data
              X val = app.preprocess input(X val)
              X_test = app.preprocess_input(X_test)
              # get the base model
              base model = model name(include top=False, input shape=(100, 100, 3
          ), weights='imagenet')
              # get the output of the base model
              mod = base model.output
              # add new layers on top
              mod = Flatten()(mod)
              mod = Dense(1024, activation='relu')(mod)
              mod = Dropout(drop rate)(mod)
              mod = Dense(512, activation='relu')(mod)
              mod = Dropout(drop_rate)(mod)
              mod = Dense(256, activation='relu')(mod)
              mod = Dropout(drop_rate)(mod)
              # add output layers
              output = Dense(1, activation='sigmoid')(mod)
              # start time
              start = time.time()
              # make new model
              model = Model(inputs=base model.inputs, outputs=output)
              # unfreeze k last layers of base model
              for layer in model.layers[:len(base model.layers)-k]:
                  layer.trainable = False
              for layer in model.layers[len(base_model.layers)-k:]:
                  layer.trainable = True
              # compile the model
              model.compile(optimizer=SGD(lr=0.0001, momentum=0.9), loss='binary c
          rossentropy', metrics = ['accuracy'])
               model.compile(optimizer=Adam(lr=0.0001), loss='binary crossentrop
          y', metrics = ['accuracy'])
              # path to log checkpoints
              filepath = f'./best weights/{path to model}'
```

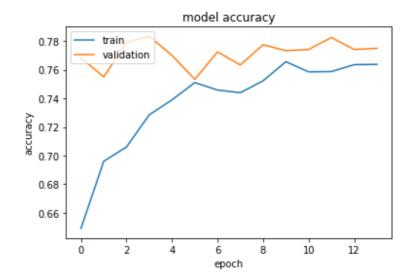
```
# early stopping if no improvement after 10 epochs
   earlystopping = EarlyStopping(monitor = 'val accuracy',
                                  mode = 'max' ,
                                  patience = 10,
                                  verbose = 1)
   # checkpoint for best performance
   checkpoint = ModelCheckpoint(filepath,
                                    monitor = 'val accuracy',
                                    mode='max',
                                    save best only=True,
                                    verbose = 1)
   # reduce learning rate if no improvement
   learning rate = ReduceLROnPlateau(monitor = 'val accuracy',
                                      mode = 'max',
                                      patience = 5,
                                      factor = 0.3,
                                      min delta = min delta)
   # callbacks
   callbacks = [earlystopping, checkpoint, learning rate]
   # train the model with validation and callbacks
   history = model.fit(datagen.flow(X_train, y_train), epochs=epochs,\
                        validation data = (X val, y val),
                        batch size = batch size, class weight = class we
ight, callbacks=callbacks)
     history = model.fit(X_train, y_train, epochs=epochs, \
                          validation data = (X test, y test),
                          batch size = batch size, class weight = class
weight, callbacks=callbacks)
   print('training time', time.time() - start)
   # test the model
   print('loss and accuracy', model.evaluate(X test, y test))
   # get predicted labels based on threshold
   y pred prob train = model.predict(X train, batch size=20, verbose=1)
   y_pred_prob = model.predict(X_test, batch_size=20, verbose=1)
   y_pred = [[1] if y_pred_prob[i][0] > threshold else [0] for i in ran
ge(len(y test))]
   # detailed report
   print('detailed report')
   print(classification report(y test, y pred))
   print('confusion matrix')
   print(confusion_matrix(y_test, y_pred))
   # plot accuracy of train and validation at each epoch
   plt.plot(history.history['accuracy'])
   plt.plot(history.history['val_accuracy'])
   plt.title('model accuracy')
   plt.ylabel('accuracy')
   plt.xlabel('epoch')
   plt.legend(['train', 'validation'], loc='upper left')
   plt.show()
   # return probability for each test sample
   return y_pred_prob_train, y_train, y_pred_prob
```

In [140]: # vgg16, 0 trainable layers in base model y_pred_prob_train_vgg1, y_train_vgg1, y_pred_prob_vgg1 = individual_mode l(vgg16, VGG16, 'vgg16', X_train, y_train, X_test, y_test, min_delta = 0.00003, k = 0, epochs=20, batch_size=20, threshold=0.5, drop_rate=0.3, class weight=None)

```
Epoch 1/20
uracy: 0.6492
Epoch 00001: val accuracy improved from -inf to 0.76833, saving model t
o ./best weights/vgg16
INFO:tensorflow:Assets written to: ./best_weights/vgg16/assets
150/150 [============== ] - 31s 206ms/step - loss: 0.969
6 - accuracy: 0.6492 - val loss: 0.4809 - val accuracy: 0.7683
Epoch 2/20
uracy: 0.6960
Epoch 00002: val accuracy did not improve from 0.76833
1 - accuracy: 0.6960 - val loss: 0.4962 - val accuracy: 0.7550
Epoch 3/20
uracy: 0.7060
Epoch 00003: val accuracy improved from 0.76833 to 0.77917, saving mode
l to ./best weights/vgg16
INFO:tensorflow:Assets written to: ./best_weights/vgg16/assets
4 - accuracy: 0.7060 - val loss: 0.4817 - val accuracy: 0.7792
Epoch 4/20
uracy: 0.7285
Epoch 00004: val accuracy improved from 0.77917 to 0.78333, saving mode
1 to ./best weights/vgg16
INFO:tensorflow:Assets written to: ./best weights/vgg16/assets
150/150 [=============== ] - 31s 208ms/step - loss: 0.556
6 - accuracy: 0.7285 - val loss: 0.4753 - val accuracy: 0.7833
Epoch 5/20
uracy: 0.7390
Epoch 00005: val accuracy did not improve from 0.78333
150/150 [============== ] - 28s 188ms/step - loss: 0.538
9 - accuracy: 0.7390 - val loss: 0.4707 - val accuracy: 0.7700
Epoch 6/20
uracy: 0.7510
Epoch 00006: val_accuracy did not improve from 0.78333
150/150 [=============== ] - 28s 187ms/step - loss: 0.524
0 - accuracy: 0.7510 - val loss: 0.5109 - val accuracy: 0.7533
Epoch 7/20
uracy: 0.7458
Epoch 00007: val accuracy did not improve from 0.78333
2 - accuracy: 0.7458 - val loss: 0.4702 - val accuracy: 0.7725
Epoch 8/20
uracy: 0.7440
Epoch 00008: val_accuracy did not improve from 0.78333
1 - accuracy: 0.7440 - val_loss: 0.4819 - val_accuracy: 0.7633
Epoch 9/20
uracy: 0.7523
Epoch 00009: val accuracy did not improve from 0.78333
7 - accuracy: 0.7523 - val_loss: 0.4605 - val_accuracy: 0.7775
Epoch 10/20
uracy: 0.7656
```

12/22/2020 CS156 Final-Copy1

```
Epoch 00010: val_accuracy did not improve from 0.78333
8 - accuracy: 0.7656 - val loss: 0.4625 - val accuracy: 0.7733
Epoch 11/20
uracy: 0.7585
Epoch 00011: val accuracy did not improve from 0.78333
9 - accuracy: 0.7585 - val_loss: 0.4611 - val_accuracy: 0.7742
Epoch 12/20
uracy: 0.7588
Epoch 00012: val accuracy did not improve from 0.78333
0 - accuracy: 0.7588 - val loss: 0.4606 - val accuracy: 0.7825
Epoch 13/20
uracy: 0.7635
Epoch 00013: val accuracy did not improve from 0.78333
150/150 [=============== ] - 29s 190ms/step - loss: 0.493
5 - accuracy: 0.7635 - val_loss: 0.4617 - val_accuracy: 0.7742
Epoch 14/20
uracy: 0.7638
Epoch 00014: val_accuracy did not improve from 0.78333
150/150 [=============== ] - 28s 190ms/step - loss: 0.484
6 - accuracy: 0.7638 - val loss: 0.4595 - val accuracy: 0.7750
Epoch 00014: early stopping
training time 405.6991868019104
32/32 [===============] - 4s 139ms/step - loss: 0.7037 -
accuracy: 0.5840
loss and accuracy [0.7037352323532104, 0.5839999914169312]
240/240 [========== ] - 23s 97ms/step
50/50 [======== ] - 5s 94ms/step
detailed report
         precision
                  recall f1-score
                                  500
       0
             0.68
                    0.32
                           0.43
       1
             0.55
                    0.85
                           0.67
                                  500
                           0.58
                                  1000
  accuracy
  macro avg
             0.62
                    0.58
                           0.55
                                  1000
weighted avg
             0.62
                    0.58
                           0.55
                                  1000
confusion matrix
[[159 341]
[ 75 425]]
```



In [141]: # vgg16, 7 trainable layers in base model y_pred_prob_train_vgg2, y_train_vgg2, y_pred_prob_vgg2 = individual_mode l(vgg16, VGG16, 'vgg16_2', X_train, y_train, X_test, y_test, min_delta = 0.00003, k = 7, epochs=20, batch_size=20, threshold=0.5, drop_rate=0.3, class weight=None)

```
Epoch 1/20
uracy: 0.6927
Epoch 00001: val accuracy improved from -inf to 0.75833, saving model t
o ./best weights/vgg16 2
INFO:tensorflow:Assets written to: ./best_weights/vgg16_2/assets
150/150 [============== ] - 76s 506ms/step - loss: 0.628
8 - accuracy: 0.6927 - val loss: 0.4849 - val accuracy: 0.7583
Epoch 2/20
uracy: 0.7544
Epoch 00002: val accuracy improved from 0.75833 to 0.77833, saving mode
1 to ./best weights/vgg16 2
INFO:tensorflow:Assets written to: ./best weights/vgg16 2/assets
150/150 [==============] - 73s 488ms/step - loss: 0.509
3 - accuracy: 0.7544 - val loss: 0.4780 - val accuracy: 0.7783
Epoch 3/20
uracy: 0.7671
Epoch 00003: val accuracy did not improve from 0.77833
150/150 [============== ] - 70s 466ms/step - loss: 0.485
4 - accuracy: 0.7671 - val loss: 0.4642 - val accuracy: 0.7733
Epoch 4/20
uracy: 0.7779
Epoch 00004: val accuracy did not improve from 0.77833
2 - accuracy: 0.7779 - val loss: 0.4828 - val accuracy: 0.7517
Epoch 5/20
uracy: 0.7821
Epoch 00005: val accuracy improved from 0.77833 to 0.80333, saving mode
1 to ./best_weights/vgg16_2
INFO:tensorflow:Assets written to: ./best weights/vgg16 2/assets
150/150 [===============] - 73s 486ms/step - loss: 0.460
4 - accuracy: 0.7821 - val loss: 0.4528 - val accuracy: 0.8033
Epoch 6/20
uracy: 0.7833
Epoch 00006: val_accuracy improved from 0.80333 to 0.80500, saving mode
1 to ./best weights/vgg16 2
INFO:tensorflow:Assets written to: ./best weights/vgg16 2/assets
150/150 [===============] - 72s 482ms/step - loss: 0.455
4 - accuracy: 0.7833 - val_loss: 0.4404 - val_accuracy: 0.8050
Epoch 7/20
uracy: 0.7844
Epoch 00007: val accuracy did not improve from 0.80500
150/150 [==============] - 70s 469ms/step - loss: 0.452
5 - accuracy: 0.7844 - val loss: 0.4730 - val accuracy: 0.7658
Epoch 8/20
uracy: 0.7917
Epoch 00008: val accuracy did not improve from 0.80500
150/150 [============== ] - 70s 467ms/step - loss: 0.440
0 - accuracy: 0.7917 - val loss: 0.4436 - val accuracy: 0.7933
Epoch 9/20
uracy: 0.7981
Epoch 00009: val_accuracy did not improve from 0.80500
150/150 [============================] - 70s 468ms/step - loss: 0.430
9 - accuracy: 0.7981 - val loss: 0.4555 - val accuracy: 0.8042
```

```
uracy: 0.8002
Epoch 00010: val accuracy improved from 0.80500 to 0.81083, saving mode
1 to ./best weights/vgg16 2
INFO:tensorflow:Assets written to: ./best_weights/vgg16_2/assets
1 - accuracy: 0.8002 - val loss: 0.4293 - val accuracy: 0.8108
Epoch 11/20
uracy: 0.8006
Epoch 00011: val accuracy did not improve from 0.81083
9 - accuracy: 0.8006 - val_loss: 0.4249 - val_accuracy: 0.8033
Epoch 12/20
uracy: 0.8062
Epoch 00012: val accuracy improved from 0.81083 to 0.81250, saving mode
1 to ./best weights/vgg16 2
INFO:tensorflow:Assets written to: ./best weights/vgg16 2/assets
150/150 [==============] - 73s 483ms/step - loss: 0.415
0 - accuracy: 0.8062 - val_loss: 0.4057 - val_accuracy: 0.8125
Epoch 13/20
uracy: 0.8087
Epoch 00013: val_accuracy did not improve from 0.81250
150/150 [============= ] - 70s 468ms/step - loss: 0.410
6 - accuracy: 0.8087 - val loss: 0.4209 - val accuracy: 0.8083
Epoch 14/20
uracy: 0.8150
Epoch 00014: val accuracy improved from 0.81250 to 0.81750, saving mode
1 to ./best weights/vgg16 2
INFO:tensorflow:Assets written to: ./best_weights/vgg16_2/assets
150/150 [===============] - 72s 480ms/step - loss: 0.402
0 - accuracy: 0.8150 - val_loss: 0.4165 - val_accuracy: 0.8175
Epoch 15/20
uracy: 0.8219
Epoch 00015: val_accuracy did not improve from 0.81750
150/150 [================] - 71s 471ms/step - loss: 0.387
2 - accuracy: 0.8219 - val loss: 0.4002 - val accuracy: 0.8058
Epoch 16/20
uracy: 0.8246
Epoch 00016: val accuracy improved from 0.81750 to 0.82417, saving mode
1 to ./best weights/vgg16 2
INFO:tensorflow:Assets written to: ./best_weights/vgg16_2/assets
150/150 [=============== ] - 73s 488ms/step - loss: 0.378
9 - accuracy: 0.8246 - val_loss: 0.3825 - val_accuracy: 0.8242
Epoch 17/20
uracy: 0.8269
Epoch 00017: val accuracy did not improve from 0.82417
4 - accuracy: 0.8269 - val_loss: 0.4569 - val_accuracy: 0.7825
Epoch 18/20
uracy: 0.8277
Epoch 00018: val_accuracy improved from 0.82417 to 0.83083, saving mode
1 to ./best weights/vgg16 2
INFO:tensorflow:Assets written to: ./best_weights/vgg16_2/assets
150/150 [=============== ] - 78s 522ms/step - loss: 0.373
1 - accuracy: 0.8277 - val_loss: 0.3858 - val_accuracy: 0.8308
```

```
Epoch 19/20
uracy: 0.8319
Epoch 00019: val accuracy did not improve from 0.83083
150/150 [=============== ] - 73s 488ms/step - loss: 0.364
1 - accuracy: 0.8319 - val loss: 0.4536 - val accuracy: 0.8075
Epoch 20/20
uracy: 0.8340
Epoch 00020: val accuracy improved from 0.83083 to 0.84500, saving mode
1 to ./best weights/vgg16 2
INFO:tensorflow:Assets written to: ./best weights/vgg16 2/assets
150/150 [============== ] - 76s 510ms/step - loss: 0.360
1 - accuracy: 0.8340 - val loss: 0.3892 - val accuracy: 0.8450
training time 1460.3803207874298
32/32 [============= ] - 4s 135ms/step - loss: 0.6762 -
accuracy: 0.6520
loss and accuracy [0.6762085556983948, 0.6520000100135803]
240/240 [=========== ] - 23s 97ms/step
50/50 [======== ] - 5s 96ms/step
detailed report
           precision
                     recall
                           f1-score
                                     support
        0
               0.75
                       0.45
                               0.57
                                        500
                       0.85
                                        500
               0.61
                               0.71
                               0.65
                                       1000
   accuracy
```

0.64

0.64

0.65

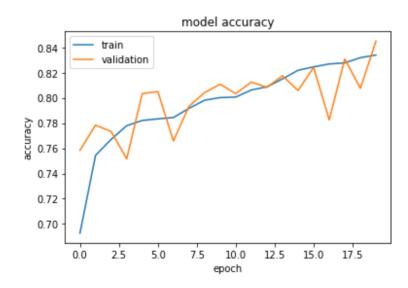
0.65

1000

1000

confusion matrix [[227 273] [75 425]]

macro avg
weighted avg



0.68

0.68

12/22/2020 CS156 Final-Copy1

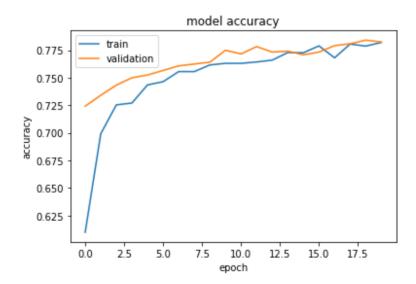
In [142]: # xception, 0 trainable layers in base model y_pred_prob_train_xception1, y_train_xception1, y_pred_prob_xception1 = individual_model(xception, Xception, 'xception', X_train, y_train, X_tes t, y_test, min_delta = 0.00003, k = 0, epochs=20, batch_size=20, thresho ld=0.5, drop rate=0.3, class weight=None)

```
Epoch 1/20
uracy: 0.6098
Epoch 00001: val accuracy improved from -inf to 0.72417, saving model t
o ./best weights/xception
INFO:tensorflow:Assets written to: ./best_weights/xception/assets
150/150 [============= ] - 43s 286ms/step - loss: 0.657
8 - accuracy: 0.6098 - val loss: 0.5801 - val accuracy: 0.7242
Epoch 2/20
uracy: 0.6992
Epoch 00002: val accuracy improved from 0.72417 to 0.73417, saving mode
1 to ./best weights/xception
INFO:tensorflow:Assets written to: ./best weights/xception/assets
0 - accuracy: 0.6992 - val loss: 0.5387 - val accuracy: 0.7342
Epoch 3/20
uracy: 0.7254
Epoch 00003: val accuracy improved from 0.73417 to 0.74333, saving mode
1 to ./best weights/xception
INFO: tensorflow: Assets written to: ./best weights/xception/assets
7 - accuracy: 0.7254 - val_loss: 0.5145 - val_accuracy: 0.7433
Epoch 4/20
uracy: 0.7271
Epoch 00004: val accuracy improved from 0.74333 to 0.75000, saving mode
1 to ./best weights/xception
INFO:tensorflow:Assets written to: ./best_weights/xception/assets
150/150 [============== ] - 41s 273ms/step - loss: 0.546
2 - accuracy: 0.7271 - val loss: 0.5033 - val accuracy: 0.7500
Epoch 5/20
uracy: 0.7435
Epoch 00005: val accuracy improved from 0.75000 to 0.75250, saving mode
1 to ./best weights/xception
INFO:tensorflow:Assets written to: ./best weights/xception/assets
6 - accuracy: 0.7435 - val_loss: 0.4948 - val_accuracy: 0.7525
Epoch 6/20
uracy: 0.7465
Epoch 00006: val accuracy improved from 0.75250 to 0.75667, saving mode
1 to ./best_weights/xception
INFO:tensorflow:Assets written to: ./best weights/xception/assets
150/150 [============== ] - 40s 265ms/step - loss: 0.519
5 - accuracy: 0.7465 - val_loss: 0.4898 - val accuracy: 0.7567
Epoch 7/20
uracy: 0.7556
Epoch 00007: val_accuracy improved from 0.75667 to 0.76083, saving mode
1 to ./best_weights/xception
INFO:tensorflow:Assets written to: ./best_weights/xception/assets
1 - accuracy: 0.7556 - val loss: 0.4829 - val accuracy: 0.7608
Epoch 8/20
uracy: 0.7556
Epoch 00008: val_accuracy improved from 0.76083 to 0.76250, saving mode
1 to ./best_weights/xception
INFO:tensorflow:Assets written to: ./best weights/xception/assets
```

```
6 - accuracy: 0.7556 - val_loss: 0.4797 - val_accuracy: 0.7625
Epoch 9/20
150/150 [============== ] - ETA: 0s - loss: 0.4944 - acc
uracy: 0.7617
Epoch 00009: val accuracy improved from 0.76250 to 0.76417, saving mode
1 to ./best weights/xception
INFO:tensorflow:Assets written to: ./best weights/xception/assets
4 - accuracy: 0.7617 - val loss: 0.4778 - val accuracy: 0.7642
Epoch 10/20
uracy: 0.7631
Epoch 00010: val accuracy improved from 0.76417 to 0.77500, saving mode
1 to ./best weights/xception
INFO:tensorflow:Assets written to: ./best_weights/xception/assets
150/150 [================] - 37s 249ms/step - loss: 0.491
8 - accuracy: 0.7631 - val loss: 0.4725 - val accuracy: 0.7750
Epoch 11/20
uracy: 0.7631
Epoch 00011: val_accuracy did not improve from 0.77500
9 - accuracy: 0.7631 - val_loss: 0.4727 - val_accuracy: 0.7717
Epoch 12/20
150/150 [============== ] - ETA: 0s - loss: 0.4909 - acc
uracy: 0.7644
Epoch 00012: val accuracy improved from 0.77500 to 0.77833, saving mode
1 to ./best weights/xception
INFO:tensorflow:Assets written to: ./best_weights/xception/assets
150/150 [==============] - 37s 248ms/step - loss: 0.490
9 - accuracy: 0.7644 - val loss: 0.4698 - val accuracy: 0.7783
Epoch 13/20
uracy: 0.7660
Epoch 00013: val_accuracy did not improve from 0.77833
0 - accuracy: 0.7660 - val_loss: 0.4672 - val_accuracy: 0.7733
Epoch 14/20
uracy: 0.7729
Epoch 00014: val accuracy did not improve from 0.77833
3 - accuracy: 0.7729 - val_loss: 0.4655 - val_accuracy: 0.7742
Epoch 15/20
uracy: 0.7727
Epoch 00015: val_accuracy did not improve from 0.77833
150/150 [================] - 21s 140ms/step - loss: 0.482
4 - accuracy: 0.7727 - val_loss: 0.4638 - val_accuracy: 0.7708
Epoch 16/20
uracy: 0.7790
Epoch 00016: val accuracy did not improve from 0.77833
1 - accuracy: 0.7790 - val_loss: 0.4616 - val_accuracy: 0.7733
Epoch 17/20
uracy: 0.7681
Epoch 00017: val_accuracy improved from 0.77833 to 0.77917, saving mode
1 to ./best weights/xception
INFO:tensorflow:Assets written to: ./best_weights/xception/assets
1 - accuracy: 0.7681 - val_loss: 0.4609 - val_accuracy: 0.7792
```

```
Epoch 18/20
uracy: 0.7806
Epoch 00018: val accuracy improved from 0.77917 to 0.78083, saving mode
l to ./best weights/xception
INFO:tensorflow:Assets written to: ./best weights/xception/assets
0 - accuracy: 0.7806 - val_loss: 0.4631 - val_accuracy: 0.7808
Epoch 19/20
uracy: 0.7788
Epoch 00019: val accuracy improved from 0.78083 to 0.78417, saving mode
1 to ./best weights/xception
INFO:tensorflow:Assets written to: ./best weights/xception/assets
150/150 [==============] - 37s 244ms/step - loss: 0.473
7 - accuracy: 0.7788 - val loss: 0.4573 - val accuracy: 0.7842
Epoch 20/20
uracy: 0.7821
Epoch 00020: val accuracy did not improve from 0.78417
150/150 [============== ] - 21s 139ms/step - loss: 0.467
4 - accuracy: 0.7821 - val loss: 0.4563 - val accuracy: 0.7825
training time 673.9549684524536
32/32 [================== ] - 3s 85ms/step - loss: 0.7652 -
accuracy: 0.5860
loss and accuracy [0.7651541829109192, 0.5860000252723694]
50/50 [=======] - 3s 64ms/step
detailed report
          precision
                    recall f1-score
                                  support
              0.65
                     0.37
                             0.47
                                     500
              0.56
        1
                     0.80
                             0.66
                                     500
                             0.59
                                    1000
   accuracy
  macro avg
              0.61
                     0.59
                             0.57
                                    1000
weighted avg
              0.61
                     0.59
                             0.57
                                    1000
```

confusion matrix
[[186 314]
 [100 400]]



In [143]: # xception, 7 trainable layers in base model y_pred_prob_train_xception2, y_train_xception2, y_pred_prob_xception2 = individual_model(xception, Xception, 'xception_2', X_train, y_train, X_t est, y_test, min_delta = 0.00003, k = 7, epochs=20, batch_size=20, thres hold=0.5, drop_rate=0.3, class_weight=None)

12/22/2020

```
Epoch 1/20
uracy: 0.5610
Epoch 00001: val accuracy improved from -inf to 0.58833, saving model t
o ./best weights/xception 2
INFO:tensorflow:Assets written to: ./best_weights/xception_2/assets
150/150 [============= ] - 45s 301ms/step - loss: 0.686
1 - accuracy: 0.5610 - val loss: 0.6622 - val accuracy: 0.5883
Epoch 2/20
uracy: 0.6540
Epoch 00002: val accuracy improved from 0.58833 to 0.73500, saving mode
1 to ./best weights/xception 2
INFO:tensorflow:Assets written to: ./best weights/xception 2/assets
7 - accuracy: 0.6540 - val loss: 0.6008 - val accuracy: 0.7350
Epoch 3/20
uracy: 0.7008
Epoch 00003: val accuracy improved from 0.73500 to 0.74500, saving mode
1 to ./best weights/xception 2
INFO:tensorflow:Assets written to: ./best weights/xception 2/assets
5 - accuracy: 0.7008 - val_loss: 0.5778 - val_accuracy: 0.7450
Epoch 4/20
150/150 [=============== ] - ETA: 0s - loss: 0.6016 - acc
uracy: 0.7210
Epoch 00004: val accuracy improved from 0.74500 to 0.74917, saving mode
1 to ./best weights/xception 2
INFO:tensorflow:Assets written to: ./best_weights/xception_2/assets
6 - accuracy: 0.7210 - val loss: 0.5569 - val accuracy: 0.7492
Epoch 5/20
uracy: 0.7317
Epoch 00005: val accuracy did not improve from 0.74917
4 - accuracy: 0.7317 - val loss: 0.5391 - val accuracy: 0.7492
Epoch 6/20
uracy: 0.7406
Epoch 00006: val accuracy improved from 0.74917 to 0.75167, saving mode
1 to ./best weights/xception 2
INFO:tensorflow:Assets written to: ./best weights/xception 2/assets
150/150 [=============== ] - 45s 297ms/step - loss: 0.554
1 - accuracy: 0.7406 - val loss: 0.5218 - val accuracy: 0.7517
Epoch 7/20
uracy: 0.7481
Epoch 00007: val accuracy improved from 0.75167 to 0.76167, saving mode
1 to ./best weights/xception 2
INFO:tensorflow:Assets written to: ./best_weights/xception_2/assets
2 - accuracy: 0.7481 - val_loss: 0.5087 - val_accuracy: 0.7617
Epoch 8/20
uracy: 0.7519
Epoch 00008: val accuracy improved from 0.76167 to 0.76250, saving mode
1 to ./best weights/xception 2
INFO:tensorflow:Assets written to: ./best_weights/xception_2/assets
5 - accuracy: 0.7519 - val loss: 0.5003 - val accuracy: 0.7625
```

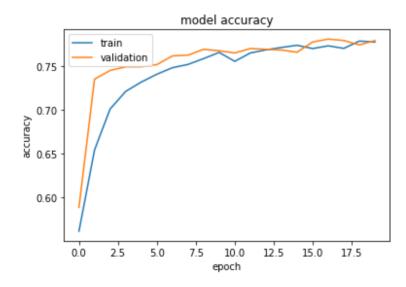
Epoch 9/20

```
uracy: 0.7585
Epoch 00009: val accuracy improved from 0.76250 to 0.76917, saving mode
1 to ./best weights/xception 2
INFO:tensorflow:Assets written to: ./best_weights/xception_2/assets
4 - accuracy: 0.7585 - val loss: 0.4944 - val accuracy: 0.7692
Epoch 10/20
uracy: 0.7656
Epoch 00010: val accuracy did not improve from 0.76917
7 - accuracy: 0.7656 - val loss: 0.4882 - val accuracy: 0.7675
Epoch 11/20
uracy: 0.7554
Epoch 00011: val accuracy did not improve from 0.76917
7 - accuracy: 0.7554 - val_loss: 0.4835 - val_accuracy: 0.7650
Epoch 12/20
uracy: 0.7650
Epoch 00012: val_accuracy improved from 0.76917 to 0.77000, saving mode
1 to ./best weights/xception 2
INFO:tensorflow:Assets written to: ./best_weights/xception_2/assets
2 - accuracy: 0.7650 - val loss: 0.4786 - val accuracy: 0.7700
Epoch 13/20
uracy: 0.7685
Epoch 00013: val accuracy did not improve from 0.77000
9 - accuracy: 0.7685 - val loss: 0.4760 - val accuracy: 0.7692
Epoch 14/20
uracy: 0.7713
Epoch 00014: val_accuracy did not improve from 0.77000
150/150 [================] - 27s 181ms/step - loss: 0.494
3 - accuracy: 0.7713 - val_loss: 0.4723 - val_accuracy: 0.7683
Epoch 15/20
uracy: 0.7738
Epoch 00015: val_accuracy did not improve from 0.77000
150/150 [=============] - 27s 180ms/step - loss: 0.489
1 - accuracy: 0.7738 - val loss: 0.4694 - val accuracy: 0.7658
Epoch 16/20
uracy: 0.7700
Epoch 00016: val_accuracy improved from 0.77000 to 0.77750, saving mode
1 to ./best weights/xception 2
INFO:tensorflow:Assets written to: ./best weights/xception 2/assets
8 - accuracy: 0.7700 - val_loss: 0.4672 - val_accuracy: 0.7775
Epoch 17/20
uracy: 0.7731
Epoch 00017: val_accuracy improved from 0.77750 to 0.78083, saving mode
1 to ./best weights/xception 2
INFO:tensorflow:Assets written to: ./best_weights/xception_2/assets
9 - accuracy: 0.7731 - val_loss: 0.4671 - val_accuracy: 0.7808
Epoch 18/20
```

```
uracy: 0.7702
Epoch 00018: val accuracy did not improve from 0.78083
150/150 [============== ] - 27s 179ms/step - loss: 0.481
4 - accuracy: 0.7702 - val loss: 0.4610 - val accuracy: 0.7792
Epoch 19/20
uracy: 0.7785
Epoch 00019: val accuracy did not improve from 0.78083
150/150 [===========================] - 28s 184ms/step - loss: 0.475
0 - accuracy: 0.7785 - val loss: 0.4587 - val accuracy: 0.7742
Epoch 20/20
uracy: 0.7775
Epoch 00020: val accuracy did not improve from 0.78083
150/150 [=============== ] - 27s 180ms/step - loss: 0.476
1 - accuracy: 0.7775 - val loss: 0.4577 - val accuracy: 0.7792
training time 734.8332531452179
32/32 [============= ] - 3s 83ms/step - loss: 0.7581 -
accuracy: 0.5850
loss and accuracy [0.7581257820129395, 0.5849999785423279]
240/240 [========== ] - 15s 61ms/step
detailed report
           precision
                     recall f1-score
        0
               0.65
                       0.36
                               0.47
                                        500
        1
               0.56
                       0.81
                               0.66
                                        500
   accuracy
                               0.58
                                       1000
  macro avg
               0.61
                       0.58
                               0.56
                                       1000
                                       1000
```

confusion matrix [[181 319] [96 404]]

weighted avg



0.61

0.58

0.56

12/22/2020 CS156 Final-Copy1

In [144]: # densenet, 0 trainable layers in base model y_pred_prob_train_densenet1, y_train_densenet1, y_pred_prob_densenet1 =

individual_model(densenet, DenseNet121, 'densenet', X_train, y_train, X_ test, y_test, min_delta = 0.00003, k = 0, epochs=20, batch_size=20, thre shold=0.5, drop_rate=0.3, class_weight=None)

12/22/2020

```
Epoch 1/20
uracy: 0.6317
Epoch 00001: val accuracy improved from -inf to 0.72833, saving model t
o ./best weights/densenet
INFO:tensorflow:Assets written to: ./best_weights/densenet/assets
150/150 [============= ] - 69s 460ms/step - loss: 0.694
4 - accuracy: 0.6317 - val loss: 0.5319 - val accuracy: 0.7283
Epoch 2/20
uracy: 0.6994
Epoch 00002: val accuracy improved from 0.72833 to 0.73833, saving mode
1 to ./best weights/densenet
INFO:tensorflow:Assets written to: ./best weights/densenet/assets
150/150 [============================] - 67s 446ms/step - loss: 0.584
5 - accuracy: 0.6994 - val loss: 0.5151 - val accuracy: 0.7383
Epoch 3/20
uracy: 0.7217
Epoch 00003: val accuracy improved from 0.73833 to 0.74083, saving mode
1 to ./best weights/densenet
INFO:tensorflow:Assets written to: ./best weights/densenet/assets
8 - accuracy: 0.7217 - val_loss: 0.5046 - val_accuracy: 0.7408
Epoch 4/20
uracy: 0.7221
Epoch 00004: val accuracy improved from 0.74083 to 0.74917, saving mode
1 to ./best weights/densenet
INFO:tensorflow:Assets written to: ./best_weights/densenet/assets
150/150 [============== ] - 71s 476ms/step - loss: 0.551
2 - accuracy: 0.7221 - val loss: 0.4984 - val accuracy: 0.7492
Epoch 5/20
uracy: 0.7467
Epoch 00005: val accuracy improved from 0.74917 to 0.75083, saving mode
1 to ./best weights/densenet
INFO:tensorflow:Assets written to: ./best weights/densenet/assets
150/150 [================] - 66s 441ms/step - loss: 0.528
1 - accuracy: 0.7467 - val_loss: 0.4961 - val_accuracy: 0.7508
Epoch 6/20
uracy: 0.7458
Epoch 00006: val accuracy improved from 0.75083 to 0.75583, saving mode
1 to ./best_weights/densenet
INFO:tensorflow:Assets written to: ./best weights/densenet/assets
1 - accuracy: 0.7458 - val_loss: 0.4875 - val accuracy: 0.7558
Epoch 7/20
uracy: 0.7535
Epoch 00007: val_accuracy improved from 0.75583 to 0.75750, saving mode
1 to ./best_weights/densenet
INFO:tensorflow:Assets written to: ./best_weights/densenet/assets
150/150 [=============== ] - 66s 438ms/step - loss: 0.512
6 - accuracy: 0.7535 - val loss: 0.4857 - val accuracy: 0.7575
Epoch 8/20
uracy: 0.7460
Epoch 00008: val_accuracy improved from 0.75750 to 0.76167, saving mode
1 to ./best weights/densenet
INFO:tensorflow:Assets written to: ./best weights/densenet/assets
```

12/22/2020 CS156 Final-Copy1

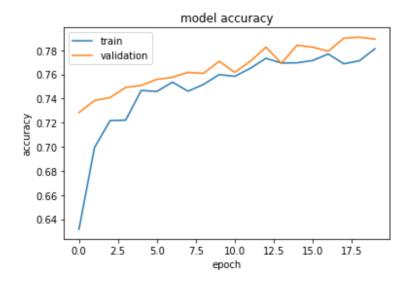
```
8 - accuracy: 0.7460 - val_loss: 0.4799 - val accuracy: 0.7617
Epoch 9/20
150/150 [============== ] - ETA: 0s - loss: 0.5079 - acc
uracy: 0.7517
Epoch 00009: val_accuracy did not improve from 0.76167
9 - accuracy: 0.7517 - val loss: 0.4792 - val accuracy: 0.7608
Epoch 10/20
uracy: 0.7598
Epoch 00010: val accuracy improved from 0.76167 to 0.77083, saving mode
1 to ./best weights/densenet
INFO:tensorflow:Assets written to: ./best weights/densenet/assets
150/150 [=============] - 66s 443ms/step - loss: 0.499
3 - accuracy: 0.7598 - val_loss: 0.4744 - val_accuracy: 0.7708
Epoch 11/20
uracy: 0.7583
Epoch 00011: val accuracy did not improve from 0.77083
150/150 [============== ] - 23s 152ms/step - loss: 0.494
0 - accuracy: 0.7583 - val_loss: 0.4687 - val_accuracy: 0.7617
Epoch 12/20
uracy: 0.7650
Epoch 00012: val_accuracy did not improve from 0.77083
3 - accuracy: 0.7650 - val loss: 0.4678 - val accuracy: 0.7708
Epoch 13/20
uracy: 0.7733
Epoch 00013: val accuracy improved from 0.77083 to 0.78250, saving mode
1 to ./best weights/densenet
INFO:tensorflow:Assets written to: ./best_weights/densenet/assets
150/150 [================] - 71s 470ms/step - loss: 0.482
6 - accuracy: 0.7733 - val_loss: 0.4643 - val_accuracy: 0.7825
Epoch 14/20
uracy: 0.7694
Epoch 00014: val_accuracy did not improve from 0.78250
0 - accuracy: 0.7694 - val loss: 0.4625 - val accuracy: 0.7692
Epoch 15/20
uracy: 0.7696
Epoch 00015: val accuracy improved from 0.78250 to 0.78417, saving mode
1 to ./best weights/densenet
INFO:tensorflow:Assets written to: ./best_weights/densenet/assets
150/150 [=============== ] - 71s 471ms/step - loss: 0.478
6 - accuracy: 0.7696 - val_loss: 0.4616 - val_accuracy: 0.7842
Epoch 16/20
uracy: 0.7715
Epoch 00016: val accuracy did not improve from 0.78417
150/150 [============= ] - 23s 154ms/step - loss: 0.476
4 - accuracy: 0.7715 - val_loss: 0.4625 - val_accuracy: 0.7825
Epoch 17/20
uracy: 0.7769
Epoch 00017: val_accuracy did not improve from 0.78417
150/150 [============== ] - 23s 152ms/step - loss: 0.470
6 - accuracy: 0.7769 - val_loss: 0.4610 - val_accuracy: 0.7792
Epoch 18/20
```

```
CS156 Final-Copy1
uracy: 0.7688
Epoch 00018: val accuracy improved from 0.78417 to 0.79000, saving mode
1 to ./best weights/densenet
INFO:tensorflow:Assets written to: ./best weights/densenet/assets
150/150 [============================] - 69s 462ms/step - loss: 0.476
9 - accuracy: 0.7688 - val loss: 0.4542 - val accuracy: 0.7900
Epoch 19/20
uracy: 0.7713
Epoch 00019: val accuracy improved from 0.79000 to 0.79083, saving mode
1 to ./best weights/densenet
INFO:tensorflow:Assets written to: ./best weights/densenet/assets
8 - accuracy: 0.7713 - val loss: 0.4560 - val accuracy: 0.7908
Epoch 20/20
uracy: 0.7812
Epoch 00020: val accuracy did not improve from 0.79083
150/150 [============================] - 23s 152ms/step - loss: 0.463
6 - accuracy: 0.7812 - val loss: 0.4515 - val accuracy: 0.7892
training time 1065.6567225456238
32/32 [================ ] - 3s 104ms/step - loss: 0.7049 -
accuracy: 0.6060
loss and accuracy [0.7048950791358948, 0.6060000061988831]
240/240 [=========== ] - 21s 88ms/step
50/50 [========= ] - 4s 86ms/step
```

-	precision	recall	f1-score	support
0	0.72	0.34	0.47	500
1	0.57	0.87	0.69	500
accuracy			0.61	1000
macro avg	0.65	0.61	0.58	1000
weighted avg	0.65	0.61	0.58	1000

confusion matrix [[172 328] [66 434]]

detailed report



In [145]:

densenet, 7 trainable layers in base model

y_pred_prob_train_densenet2, y_train_densenet2, y_pred_prob_densenet2 =
individual_model(densenet, DenseNet121, 'densenet_2', X_train, y_train,
X_test, y_test, min_delta = 0.00003, k = 7, epochs=20, batch_size=20, th
reshold=0.5, drop_rate=0.3, class_weight=None)

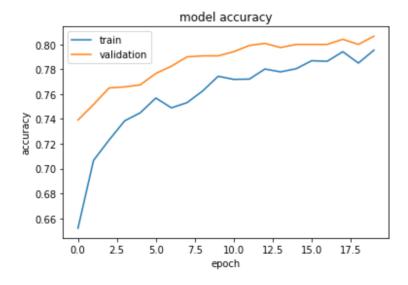
```
12/22/2020
                                  CS156 Final-Copy1
         Epoch 1/20
          uracy: 0.6521
         Epoch 00001: val accuracy improved from -inf to 0.73917, saving model t
          o ./best weights/densenet 2
          INFO:tensorflow:Assets written to: ./best_weights/densenet_2/assets
          150/150 [==============] - 74s 492ms/step - loss: 0.670
          8 - accuracy: 0.6521 - val loss: 0.5260 - val accuracy: 0.7392
         Epoch 2/20
          uracy: 0.7067
         Epoch 00002: val accuracy improved from 0.73917 to 0.75167, saving mode
          1 to ./best weights/densenet 2
          INFO:tensorflow:Assets written to: ./best weights/densenet 2/assets
          150/150 [==============] - 70s 464ms/step - loss: 0.596
          5 - accuracy: 0.7067 - val loss: 0.4930 - val accuracy: 0.7517
         Epoch 3/20
          uracy: 0.7231
         Epoch 00003: val accuracy improved from 0.75167 to 0.76500, saving mode
          1 to ./best weights/densenet 2
          INFO:tensorflow:Assets written to: ./best weights/densenet 2/assets
          4 - accuracy: 0.7231 - val_loss: 0.4852 - val_accuracy: 0.7650
         Epoch 4/20
          uracy: 0.7385
         Epoch 00004: val accuracy improved from 0.76500 to 0.76583, saving mode
          1 to ./best weights/densenet 2
          INFO:tensorflow:Assets written to: ./best_weights/densenet_2/assets
          150/150 [===============] - 74s 490ms/step - loss: 0.537
          0 - accuracy: 0.7385 - val loss: 0.4721 - val accuracy: 0.7658
         Epoch 5/20
          uracy: 0.7450
         Epoch 00005: val accuracy improved from 0.76583 to 0.76750, saving mode
          1 to ./best weights/densenet 2
          INFO:tensorflow:Assets written to: ./best weights/densenet 2/assets
          150/150 [================] - 74s 490ms/step - loss: 0.523
          4 - accuracy: 0.7450 - val_loss: 0.4826 - val_accuracy: 0.7675
         Epoch 6/20
          uracy: 0.7569
         Epoch 00006: val accuracy improved from 0.76750 to 0.77667, saving mode
          1 to ./best_weights/densenet_2
          INFO:tensorflow:Assets written to: ./best weights/densenet 2/assets
          150/150 [============== ] - 69s 462ms/step - loss: 0.509
          5 - accuracy: 0.7569 - val_loss: 0.4632 - val accuracy: 0.7767
         Epoch 7/20
          uracy: 0.7490
         Epoch 00007: val_accuracy improved from 0.77667 to 0.78250, saving mode
          1 to ./best_weights/densenet_2
          INFO:tensorflow:Assets written to: ./best_weights/densenet_2/assets
          150/150 [============== ] - 70s 466ms/step - loss: 0.505
          2 - accuracy: 0.7490 - val loss: 0.4574 - val accuracy: 0.7825
         Epoch 8/20
          uracy: 0.7531
         Epoch 00008: val accuracy improved from 0.78250 to 0.79000, saving mode
```

1 to ./best weights/densenet 2

```
6 - accuracy: 0.7531 - val_loss: 0.4541 - val_accuracy: 0.7900
Epoch 9/20
150/150 [============== ] - ETA: 0s - loss: 0.4925 - acc
uracy: 0.7625
Epoch 00009: val accuracy improved from 0.79000 to 0.79083, saving mode
1 to ./best weights/densenet 2
INFO:tensorflow:Assets written to: ./best weights/densenet 2/assets
5 - accuracy: 0.7625 - val_loss: 0.4550 - val_accuracy: 0.7908
Epoch 10/20
uracy: 0.7744
Epoch 00010: val accuracy did not improve from 0.79083
8 - accuracy: 0.7744 - val_loss: 0.4569 - val_accuracy: 0.7908
Epoch 11/20
uracy: 0.7719
Epoch 00011: val accuracy improved from 0.79083 to 0.79417, saving mode
1 to ./best weights/densenet 2
INFO:tensorflow:Assets written to: ./best_weights/densenet_2/assets
0 - accuracy: 0.7719 - val_loss: 0.4482 - val_accuracy: 0.7942
Epoch 12/20
150/150 [============== ] - ETA: 0s - loss: 0.4746 - acc
uracy: 0.7721
Epoch 00012: val accuracy improved from 0.79417 to 0.79917, saving mode
1 to ./best weights/densenet 2
INFO:tensorflow:Assets written to: ./best_weights/densenet_2/assets
6 - accuracy: 0.7721 - val loss: 0.4422 - val accuracy: 0.7992
Epoch 13/20
uracy: 0.7802
Epoch 00013: val_accuracy improved from 0.79917 to 0.80083, saving mode
1 to ./best weights/densenet 2
INFO:tensorflow:Assets written to: ./best_weights/densenet_2/assets
150/150 [============] - 73s 485ms/step - loss: 0.466
1 - accuracy: 0.7802 - val_loss: 0.4410 - val_accuracy: 0.8008
Epoch 14/20
uracy: 0.7779
Epoch 00014: val_accuracy did not improve from 0.80083
150/150 [=============] - 26s 172ms/step - loss: 0.458
8 - accuracy: 0.7779 - val loss: 0.4375 - val accuracy: 0.7975
Epoch 15/20
uracy: 0.7804
Epoch 00015: val_accuracy did not improve from 0.80083
3 - accuracy: 0.7804 - val loss: 0.4355 - val accuracy: 0.8000
Epoch 16/20
uracy: 0.7869
Epoch 00016: val_accuracy did not improve from 0.80083
5 - accuracy: 0.7869 - val_loss: 0.4357 - val_accuracy: 0.8000
Epoch 17/20
uracy: 0.7865
Epoch 00017: val_accuracy did not improve from 0.80083
150/150 [============================] - 26s 171ms/step - loss: 0.448
0 - accuracy: 0.7865 - val_loss: 0.4331 - val_accuracy: 0.8000
```

```
Epoch 18/20
uracy: 0.7942
Epoch 00018: val accuracy improved from 0.80083 to 0.80417, saving mode
1 to ./best weights/densenet 2
INFO:tensorflow:Assets written to: ./best weights/densenet 2/assets
150/150 [============== ] - 70s 464ms/step - loss: 0.447
7 - accuracy: 0.7942 - val_loss: 0.4298 - val_accuracy: 0.8042
Epoch 19/20
uracy: 0.7850
Epoch 00019: val accuracy did not improve from 0.80417
7 - accuracy: 0.7850 - val loss: 0.4294 - val accuracy: 0.8000
Epoch 20/20
uracy: 0.7954
Epoch 00020: val accuracy improved from 0.80417 to 0.80667, saving mode
1 to ./best weights/densenet 2
INFO:tensorflow:Assets written to: ./best_weights/densenet_2/assets
150/150 [=============] - 71s 475ms/step - loss: 0.445
7 - accuracy: 0.7954 - val loss: 0.4256 - val accuracy: 0.8067
training time 1162.5787949562073
32/32 [================ ] - 3s 107ms/step - loss: 0.6867 -
accuracy: 0.6000
loss and accuracy [0.6867461204528809, 0.6000000238418579]
240/240 [============ ] - 21s 89ms/step
50/50 [========] - 4s 90ms/step
detailed report
          precision
                    recall f1-score
                                   support
              0.66
                      0.41
                              0.50
                                       500
              0.57
                      0.79
        1
                              0.66
                                       500
                              0.60
                                      1000
   accuracy
  macro avg
              0.62
                      0.60
                              0.58
                                      1000
weighted avg
              0.62
                      0.60
                              0.58
                                      1000
```

confusion matrix
[[203 297]
[103 397]]



Traditional ML

```
# reference: https://colab.research.google.com/drive/10PKdqvyUyFQ9gCOto2
In [170]:
          dGqc5z9AdtoVIN
          # transform the data with pca
          def pca(X train, X test, n):
              pca = PCA(n components=n)
              X_train_pca = pca.fit_transform(X_train)
              X_test_pca = pca.transform(X_test)
              return X_train_pca, X_test_pca
          # transform the data with lda
          def lda(X_train, X_test, y_train):
              lda = LinearDiscriminantAnalysis(solver='eigen', shrinkage=0.5)
              X train lda = lda.fit transform(X train, y train)
              X test lda = lda.transform(X test)
              return X_train_lda, X_test_lda
          # logistic regression
          def logistic regression(X train, X test, y train, y test, threshold=0.5
              start = time.time()
              # build the model
              lr = LogisticRegressionCV(solver='liblinear')
              # fit the model
              lr.fit(X train, y train)
              print('running time: ', time.time()-start)
              # predict with train and test data
              y train pred = lr.predict(X train)
              y_test_pred = np.where(lr.predict_proba(X_test)[:,1] > threshold, 1,
          0)
              # print the results
              print('accuracy for train: ', accuracy score(y_train, y_train_pred))
              print('accuracy for test: ', accuracy_score(y_test, y_test_pred))
              print('confusion matrix: ', confusion matrix(y test, y test pred))
              return lr.predict proba(X train), lr.predict proba(X test)
          def rfc(X_train, X_test, y_train, y_test):
              # cross validation for random forest classifier
              param grid = {
                  'n estimators': [200, 500],
                   'max_features': ['auto', 'sqrt', 'log2'],
                  'max_depth' : [4,5,6,7,8],
                  'criterion' :['gini', 'entropy']
              }
              rfc_ = RandomForestClassifier(random state=123)
              grc = GridSearchCV(rfc , param grid=param grid, cv=7)
              grc.fit(X train lda, y train)
              params = grc.best params
              # build random forest classifier
              rfc = RandomForestClassifier(random state=123, max features=params[
          'max features'],\
                                            n estimators= params['n estimators'], m
          ax depth=params['max depth'], criterion=params['criterion'])
              # fit the model
              rfc.fit(X train, y train)
              # predict on train and test
              y_train_pred = rfc.predict(X_train)
              y_test_pred = rfc.predict(X_test)
              print('accuracy for train: ', accuracy score(y_train, y_train_pred))
```

```
print('accuracy for test: ', accuracy_score(y_test, y_test_pred))
return rfc.predict_proba(X_train), rfc.predict_proba(X_test)
```

```
In [110]: # flatten data
          X train flatten = np.array([X train[i].flatten() for i in range(len(X tr
          ain))])
          X test flatten = np.array([X test[i].flatten() for i in range(len(X test
          ))])
In [112]: # transform data with pca and lda
          X_train_pca, X_test_pca = pca(X_train_flatten, X_test_flatten, 50)
          X_train_lda, X_test_lda = lda(X_train_flatten, X_test_flatten, y_train.r
          avel())
In [172]: # linear regression on pca-transformed data
          y pred prob pca lr train, y pred prob pca lr test = logistic regression
          (X_train_pca, X_test_pca, y_train.ravel(), y_test.ravel(), threshold =
          0.5)
          running time: 10.490171194076538
          accuracy for train: 0.7741666666666667
          accuracy for test: 0.598
          confusion matrix: [[198 302]
           [100 400]]
In [151]: # linear regression on lda-transformed data
          y_pred_prob_lda_lr_train, y_pred_prob_lda_lr_test = logistic_regression
          (X train lda, X test lda, y train.ravel(), y test.ravel(), , threshold =
          0.5)
          running time: 0.20975112915039062
          accuracy for train: 0.897
          accuracy for test: 0.59
          confusion matrix: [[237 263]
           [147 353]]
```

accuracy for train: 0.8736666666666667 accuracy for test: 0.588

Combined outputs

12/22/2020 CS156 Final-Copy1

```
# combine the training predicted probabilities of logistic regression an
In [159]:
          d random forest classifier
          train_probs = np.hstack((y_pred_prob_pca_lr_train[:, 1].reshape(-1,1), y
          _pred prob lda lr_train[:, 1].reshape(-1,1), y pred prob pca_rfc_train
          [:, 1].reshape(-1,1)))
          train probs
Out[159]: array([[0.01843482, 0.00233519, 0.05363039],
                 [0.04834813, 0.01931042, 0.07878626],
                 [0.76216741, 0.99451645, 0.61355708],
                 [0.80390993, 0.88128137, 0.77041916],
                 [0.71174946, 0.76396806, 0.84654317],
                 [0.83952279, 0.25583959, 0.6788578911)
In [160]: # combine the test predicted probabilities of logistic regression and ra
          ndom forest classifier
          test_probs = np.hstack((y pred_prob_pca_lr_test[:, 1].reshape(-1,1), y p
          red_prob_lda_lr_test[:, 1].reshape(-1,1), y_pred_prob_pca_rfc_test[:, 1]
          .reshape(-1,1))
          test probs
Out[160]: array([[0.45777932, 0.08109445, 0.69083502],
                 [0.81604671, 0.6485476 , 0.76819172],
                 [0.87916245, 0.97777132, 0.73426752],
                 [0.49003854, 0.9545559, 0.57672728],
                 [0.57983259, 0.03566045, 0.70365834],
                 [0.75967672, 0.90817719, 0.73855011]])
```

```
In [188]: # test the combined model with different classification thresholds
for j in np.linspace(0.5, 0.9, 10):
    print('threshold: ', j)
    logistic_regression(train_probs, test_probs, y_train.ravel(), y_test
    .ravel(), threshold = j)
    print('------')
```

```
threshold: 0.5
running time: 0.293597936630249
accuracy for test: 0.566
confusion matrix: [[175 325]
[109 391]]
_____
threshold: 0.54444444444444444
running time: 0.3421592712402344
accuracy for train: 0.95683333333333333
accuracy for test: 0.566
confusion matrix: [[184 316]
[118 382]]
_____
threshold: 0.5888888888888888
running time: 0.3535337448120117
accuracy for test: 0.564
confusion matrix: [[191 309]
[127 373]]
_____
threshold: 0.63333333333333333
running time: 0.3530902862548828
accuracy for train: 0.95683333333333333
accuracy for test: 0.572
confusion matrix: [[211 289]
[139 361]]
threshold: 0.6777777777778
running time: 0.352189302444458
accuracy for test: 0.573
confusion matrix: [[217 283]
[144 356]]
______
threshold: 0.72222222222222
running time: 0.353330135345459
accuracy for test: 0.576
confusion matrix: [[231 269]
[155 345]]
_____
threshold: 0.766666666666666
running time: 0.35189151763916016
accuracy for test: 0.572
confusion matrix: [[237 263]
[165 335]]
threshold: 0.8111111111111111
running time: 0.35523319244384766
accuracy for test: 0.574
confusion matrix: [[259 241]
[185 315]]
______
threshold: 0.85555555555556
running time: 0.3651437759399414
accuracy for train: 0.95683333333333333
accuracy for test: 0.577
confusion matrix: [[277 223]
[200 300]]
```

12/22/2020 CS156 Final-Copy1

threshold: 0.9

running time: 0.36369895935058594

accuracy for test: 0.576 confusion matrix: [[301 199]

[225 275]]

In [168]: # combine the training predicted probabilities of 3 Deep Learning models train_probs_2 = np.hstack((y_pred_prob_train_vgg2, y_pred_prob_train_xce ption2, y_pred_prob_train_densenet2))

combine the training predicted probabilities of 3 Deep Learning models test_probs_2 = np.hstack((y_pred_prob_vgg2, y_pred_prob_xception2, y_pre d prob densenet2))

```
In [189]: # test the combined model with different classification thresholds
for j in np.linspace(0.5, 0.9, 10):
    print('threshold: ', j)
    logistic_regression(train_probs_2, test_probs_2, y_train_densenet2.r
    avel(), y_test.ravel(), threshold = j)
    print('------')
```

```
threshold: 0.5
running time: 0.21611738204956055
accuracy for train: 0.7970833333333334
accuracy for test: 0.662
confusion matrix: [[249 251]
 [ 87 413]]
_____
threshold: 0.54444444444444444
running time: 0.27676916122436523
accuracy for train: 0.7970833333333334
accuracy for test: 0.677
confusion matrix: [[269 231]
[ 92 408]]
_____
threshold: 0.5888888888888888
running time: 0.2768261432647705
accuracy for train: 0.79708333333333334
accuracy for test: 0.69
confusion matrix: [[286 214]
[ 96 404]]
_____
threshold: 0.6333333333333333
running time: 0.2769310474395752
accuracy for train: 0.7970833333333334
accuracy for test: 0.695
confusion matrix: [[299 201]
[104 396]]
threshold: 0.6777777777778
running time: 0.27771425247192383
accuracy for train: 0.79708333333333334
accuracy for test: 0.704
confusion matrix: [[321 179]
[117 383]]
_____
threshold: 0.72222222222222
running time: 0.2767524719238281
accuracy for train: 0.7970833333333334
accuracy for test: 0.718
confusion matrix: [[338 162]
[120 380]]
_____
threshold: 0.766666666666666
running time: 0.2769191265106201
accuracy for train: 0.79708333333333334
accuracy for test: 0.725
confusion matrix: [[354 146]
 [129 371]]
threshold: 0.8111111111111111
running time: 0.2769310474395752
accuracy for train: 0.7970833333333334
accuracy for test: 0.729
confusion matrix: [[376 124]
[147 353]]
______
threshold: 0.85555555555556
running time: 0.27667689323425293
accuracy for train: 0.7970833333333334
accuracy for test: 0.722
confusion matrix: [[394 106]
 [172 328]]
```

12/22/2020 CS156 Final-Copy1

threshold: 0.9

running time: 0.2767066955566406

accuracy for train: 0.7970833333333333

accuracy for test: 0.712 confusion matrix: [[429 71]

[217 283]]
