In [1]: !pip install keras

Requirement already satisfied: keras in c:\users\shadm\anaconda3\lib\site-packages (2.8.0)

In [2]: !pip install tensorflow

```
Requirement already satisfied: tensorflow in c:\users\shadm\anaconda3\lib\site-pac
kages (2.8.0)
Requirement already satisfied: termcolor>=1.1.0 in c:\users\shadm\anaconda3\lib\si
te-packages (from tensorflow) (1.1.0)
Requirement already satisfied: typing-extensions>=3.6.6 in c:\users\shadm\anaconda
3\lib\site-packages (from tensorflow) (4.1.1)
Requirement already satisfied: wrapt>=1.11.0 in c:\users\shadm\anaconda3\lib\site-
packages (from tensorflow) (1.12.1)
Requirement already satisfied: tf-estimator-nightly==2.8.0.dev2021122109 in c:\use
rs\shadm\anaconda3\lib\site-packages (from tensorflow) (2.8.0.dev2021122109)
Requirement already satisfied: six>=1.12.0 in c:\users\shadm\anaconda3\lib\site-pa
ckages (from tensorflow) (1.16.0)
Requirement already satisfied: protobuf>=3.9.2 in c:\users\shadm\anaconda3\lib\sit
e-packages (from tensorflow) (3.19.1)
Requirement already satisfied: opt-einsum>=2.3.2 in c:\users\shadm\anaconda3\lib\s
ite-packages (from tensorflow) (3.3.0)
Requirement already satisfied: numpy>=1.20 in c:\users\shadm\anaconda3\lib\site-pa
ckages (from tensorflow) (1.21.5)
Requirement already satisfied: setuptools in c:\users\shadm\anaconda3\lib\site-pac
kages (from tensorflow) (61.2.0)
Requirement already satisfied: keras-preprocessing>=1.1.1 in c:\users\shadm\anacon
da3\lib\site-packages (from tensorflow) (1.1.2)
Requirement already satisfied: grpcio<2.0,>=1.24.3 in c:\users\shadm\anaconda3\lib
\site-packages (from tensorflow) (1.42.0)
Requirement already satisfied: astunparse>=1.6.0 in c:\users\shadm\anaconda3\lib\s
ite-packages (from tensorflow) (1.6.3)
Requirement already satisfied: keras<2.9,>=2.8.0rc0 in c:\users\shadm\anaconda3\li
b\site-packages (from tensorflow) (2.8.0)
Requirement already satisfied: google-pasta>=0.1.1 in c:\users\shadm\anaconda3\lib
\site-packages (from tensorflow) (0.2.0)
Requirement already satisfied: tensorflow-io-gcs-filesystem>=0.23.1 in c:\users\sh
adm\anaconda3\lib\site-packages (from tensorflow) (0.25.0)
Requirement already satisfied: flatbuffers>=1.12 in c:\users\shadm\anaconda3\lib\s
ite-packages (from tensorflow) (2.0)
Requirement already satisfied: tensorboard<2.9,>=2.8 in c:\users\shadm\anaconda3\l
ib\site-packages (from tensorflow) (2.8.0)
Requirement already satisfied: gast>=0.2.1 in c:\users\shadm\anaconda3\lib\site-pa
ckages (from tensorflow) (0.5.3)
Requirement already satisfied: h5py>=2.9.0 in c:\users\shadm\anaconda3\lib\site-pa
ckages (from tensorflow) (3.6.0)
Requirement already satisfied: absl-py>=0.4.0 in c:\users\shadm\anaconda3\lib\site
-packages (from tensorflow) (1.0.0)
Requirement already satisfied: libclang>=9.0.1 in c:\users\shadm\anaconda3\lib\sit
e-packages (from tensorflow) (14.0.1)
Requirement already satisfied: wheel<1.0,>=0.23.0 in c:\users\shadm\anaconda3\lib
\site-packages (from astunparse>=1.6.0->tensorflow) (0.37.1)
Requirement already satisfied: google-auth-oauthlib<0.5,>=0.4.1 in c:\users\shadm
\anaconda3\lib\site-packages (from tensorboard<2.9,>=2.8->tensorflow) (0.4.6)
Requirement already satisfied: markdown>=2.6.8 in c:\users\shadm\anaconda3\lib\sit
e-packages (from tensorboard<2.9,>=2.8->tensorflow) (3.3.4)
Requirement already satisfied: tensorboard-plugin-wit>=1.6.0 in c:\users\shadm\ana
conda3\lib\site-packages (from tensorboard<2.9,>=2.8->tensorflow) (1.8.1)
Requirement already satisfied: google-auth<3,>=1.6.3 in c:\users\shadm\anaconda3\l
ib\site-packages (from tensorboard<2.9,>=2.8->tensorflow) (1.33.0)
Requirement already satisfied: requests<3,>=2.21.0 in c:\users\shadm\anaconda3\lib
\site-packages (from tensorboard<2.9,>=2.8->tensorflow) (2.27.1)
Requirement already satisfied: werkzeug>=0.11.15 in c:\users\shadm\anaconda3\lib\s
ite-packages (from tensorboard<2.9,>=2.8->tensorflow) (2.0.3)
Requirement already satisfied: tensorboard-data-server<0.7.0,>=0.6.0 in c:\users\s
hadm\anaconda3\lib\site-packages (from tensorboard<2.9,>=2.8->tensorflow) (0.6.1)
Requirement already satisfied: pyasn1-modules>=0.2.1 in c:\users\shadm\anaconda3\l
ib\site-packages (from google-auth<3,>=1.6.3->tensorboard<2.9,>=2.8->tensorflow)
(0.2.8)
```

Requirement already satisfied: rsa<5,>=3.1.4 in c:\users\shadm\anaconda3\lib\site-

packages (from google-auth<3,>=1.6.3->tensorboard<2.9,>=2.8->tensorflow) (4.7.2)
Requirement already satisfied: cachetools<5.0,>=2.0.0 in c:\users\shadm\anaconda3
\lib\site-packages (from google-auth<3,>=1.6.3->tensorboard<2.9,>=2.8->tensorflow)
(4.2.2)

Requirement already satisfied: requests-oauthlib>=0.7.0 in c:\users\shadm\anaconda 3\lib\site-packages (from google-auth-oauthlib<0.5,>=0.4.1->tensorboard<2.9,>=2.8->tensorflow) (1.3.1)

Requirement already satisfied: pyasn1<0.5.0,>=0.4.6 in c:\users\shadm\anaconda3\li b\site-packages (from pyasn1-modules>=0.2.1->google-auth<3,>=1.6.3->tensorboard<2.9,>=2.8->tensorflow) (0.4.8)

Requirement already satisfied: idna<4,>=2.5 in c:\users\shadm\anaconda3\lib\site-p ackages (from requests<3,>=2.21.0->tensorboard<2.9,>=2.8->tensorflow) (3.3)
Requirement already satisfied: urllib3<1.27,>=1.21.1 in c:\users\shadm\anaconda3\lib\site-packages (from requests<3,>=2.21.0->tensorboard<2.9,>=2.8->tensorflow) (1.26.9)

Requirement already satisfied: charset-normalizer~=2.0.0 in c:\users\shadm\anacond a3\lib\site-packages (from requests<3,>=2.21.0->tensorboard<2.9,>=2.8->tensorflow) (2.0.4)

Requirement already satisfied: certifi>=2017.4.17 in c:\users\shadm\anaconda3\lib\site-packages (from requests<3,>=2.21.0->tensorboard<2.9,>=2.8->tensorflow) (202 1.10.8)

Requirement already satisfied: oauthlib>=3.0.0 in c:\users\shadm\anaconda3\lib\sit e-packages (from requests-oauthlib>=0.7.0->google-auth-oauthlib<0.5,>=0.4.1->tenso rboard<2.9,>=2.8->tensorflow) (3.2.0)

```
In [3]: import os
        %matplotlib inline
        import matplotlib.pyplot as plt
        import numpy as np
        import pandas as pd
        import os
        from glob import glob
        import seaborn as sns
        from PIL import Image
        np.random.seed(11)
        from sklearn.preprocessing import StandardScaler
        from sklearn.model_selection import train_test_split, KFold, cross_val_score, Grid
        from sklearn.metrics import accuracy_score
        import itertools
        import keras
        from keras.utils.np utils import to categorical
        from keras.models import Sequential, Model
        from keras.layers import Dense, Dropout, Flatten, Conv2D, MaxPool2D
        from keras import backend as K
        from tensorflow.keras.layers import BatchNormalization
        from keras.utils.np utils import to categorical
        from tensorflow.keras.optimizers import Adam, RMSprop
        from keras.preprocessing.image import ImageDataGenerator
        from keras.callbacks import ReduceLROnPlateau
        from keras.wrappers.scikit_learn import KerasClassifier
        from tensorflow.keras.applications.resnet50 import ResNet50
        from keras import backend as K
```

```
In [4]: folder_benign_train = r'C:\Users\shadm\Desktop\Project\Biomedical_Image_Processing_
folder_malignant_train = r'C:\Users\shadm\Desktop\Project\Biomedical_Image_Process:

folder_benign_test = r'C:\Users\shadm\Desktop\Project\Biomedical_Image_Processing_u
folder_malignant_test = r'C:\Users\shadm\Desktop\Project\Biomedical_Image_Processing_u
read = lambda imname: np.asarray(Image.open(imname).convert("RGB"))
```

```
ims_benign = [read(os.path.join(folder_benign_train, filename)) for filename in os
        X_benign = np.array(ims_benign, dtype='uint8')
        ims_malignant = [read(os.path.join(folder_malignant_train, filename)) for filename
        X_malignant = np.array(ims_malignant, dtype='uint8')
        ims_benign = [read(os.path.join(folder_benign_test, filename)) for filename in os.]
        X_benign_test = np.array(ims_benign, dtype='uint8')
        ims_malignant = [read(os.path.join(folder_malignant_test, filename)) for filename
        X_malignant_test = np.array(ims_malignant, dtype='uint8')
        y_benign = np.zeros(X_benign.shape[0])
        y_malignant = np.ones(X_malignant.shape[0])
        y_benign_test = np.zeros(X_benign_test.shape[0])
        y_malignant_test = np.ones(X_malignant_test.shape[0])
        X_train = np.concatenate((X_benign, X_malignant), axis = 0)
        y_train = np.concatenate((y_benign, y_malignant), axis = 0)
        X_test = np.concatenate((X_benign_test, X_malignant_test), axis = 0)
        y_test = np.concatenate((y_benign_test, y_malignant_test), axis = 0)
        s = np.arange(X_train.shape[0])
        np.random.shuffle(s)
        X_train = X_train[s]
        y_train = y_train[s]
        s = np.arange(X_test.shape[0])
        np.random.shuffle(s)
        X_{\text{test}} = X_{\text{test}}[s]
        y_test = y_test[s]
In [5]: w=40
        h=30
        fig=plt.figure(figsize=(12, 8))
        columns = 5
        rows = 3
        for i in range(1, columns*rows +1):
             ax = fig.add_subplot(rows, columns, i)
            if y train[i] == 0:
                 ax.title.set_text('Benign')
                 ax.title.set text('Malignant')
             plt.imshow(X_train[i], interpolation='nearest')
        plt.show()
```

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```

```
y_train = to_categorical(y_train, num_classes= 2)
In [6]:
         y_test = to_categorical(y_test, num_classes= 2)
In [7]:
        X_{train} = X_{train}/255.
         X_{\text{test}} = X_{\text{test}}/255.
        def build(input_shape= (224,224,3), lr = 1e-3, num_classes= 2,
In [8]:
                   init= 'normal', activ= 'relu', optim= 'adam'):
             model = Sequential()
             model.add(Conv2D(64, kernel_size=(3, 3),padding = 'Same',input_shape=input_shape
                               activation= activ, kernel_initializer='glorot_uniform'))
             model.add(MaxPool2D(pool_size = (2, 2)))
             model.add(Dropout(0.25))
             model.add(Conv2D(64, kernel_size=(3, 3),padding = 'Same',
                               activation =activ, kernel_initializer = 'glorot_uniform'))
             model.add(MaxPool2D(pool_size = (2, 2)))
             model.add(Dropout(0.25))
             model.add(Flatten())
             model.add(Dense(128, activation='relu', kernel_initializer=init))
             model.add(Dense(num_classes, activation='softmax'))
             model.summary()
             if optim == 'rmsprop':
                 optimizer = RMSprop(lr=lr)
             else:
                 optimizer = Adam(lr=lr)
             model.compile(optimizer = optimizer ,loss = "binary_crossentropy", metrics=["ad
             return model
         learning_rate_reduction = ReduceLROnPlateau(monitor='val_accuracy',
                                                      patience=5,
                                                      verbose=1,
                                                      factor=0.5,
                                                      min lr=1e-7)
```

```
input\_shape = (224, 224, 3)
In [10]:
         lr = 1e-5
         init = 'normal'
         activ = 'relu'
         optim = 'adam'
         epochs = 40
         batch_size = 64
         model = build(lr=lr, init= init, activ= activ, optim=optim, input_shape= input_shape
        Model: "sequential"
                                   Output Shape
                                                            Param #
         Layer (type)
         ______
         conv2d (Conv2D)
                                    (None, 224, 224, 64)
                                                           1792
         max_pooling2d (MaxPooling2D (None, 112, 112, 64)
         )
         dropout (Dropout)
                                   (None, 112, 112, 64)
                                  (None, 112, 112, 64)
         conv2d_1 (Conv2D)
                                                            36928
         max_pooling2d_1 (MaxPooling (None, 56, 56, 64)
                                                            0
         2D)
                                (None, 56, 56, 64)
         dropout_1 (Dropout)
         flatten (Flatten)
                             (None, 200704)
         dense (Dense)
                                   (None, 128)
                                                            25690240
         dense_1 (Dense)
                                   (None, 2)
                                                            258
         Total params: 25,729,218
        Trainable params: 25,729,218
        Non-trainable params: 0
        C:\Users\shadm\anaconda3\lib\site-packages\keras\optimizer_v2\adam.py:105: UserWar
        ning: The `lr` argument is deprecated, use `learning_rate` instead.
          super(Adam, self).__init__(name, **kwargs)
In [11]: history = model.fit(X_train, y_train, validation_split=0.2,
                           epochs= epochs, batch size= batch size, verbose=1,
                           callbacks=[learning rate reduction]
                           )
```

```
Epoch 1/40
0.5462 - val loss: 0.6564 - val accuracy: 0.7121 - lr: 1.0000e-05
Epoch 2/40
0.5391 - val_loss: 0.6518 - val_accuracy: 0.5322 - lr: 1.0000e-05
0.5799 - val_loss: 0.6402 - val_accuracy: 0.5341 - lr: 1.0000e-05
Epoch 4/40
33/33 [===============] - 129s 4s/step - loss: 0.7645 - accuracy:
0.6083 - val_loss: 0.6640 - val_accuracy: 0.5341 - lr: 1.0000e-05
Epoch 5/40
33/33 [=============== ] - 131s 4s/step - loss: 0.7490 - accuracy:
0.5951 - val loss: 0.6505 - val accuracy: 0.5341 - lr: 1.0000e-05
Epoch 6/40
33/33 [==============] - ETA: 0s - loss: 0.7083 - accuracy: 0.6126
Epoch 6: ReduceLROnPlateau reducing learning rate to 4.999999873689376e-06.
33/33 [================] - 133s 4s/step - loss: 0.7083 - accuracy:
0.6126 - val_loss: 0.6288 - val_accuracy: 0.5436 - lr: 1.0000e-05
Epoch 7/40
33/33 [================ ] - 135s 4s/step - loss: 0.6984 - accuracy:
0.6259 - val_loss: 0.6555 - val_accuracy: 0.5341 - lr: 5.0000e-06
Epoch 8/40
33/33 [==============] - 136s 4s/step - loss: 0.6754 - accuracy:
0.6321 - val loss: 0.6514 - val accuracy: 0.5341 - lr: 5.0000e-06
Epoch 9/40
33/33 [================] - 131s 4s/step - loss: 0.6714 - accuracy:
0.6358 - val_loss: 0.6841 - val_accuracy: 0.5341 - lr: 5.0000e-06
Epoch 10/40
33/33 [==============] - 137s 4s/step - loss: 0.6488 - accuracy:
0.6543 - val_loss: 0.6506 - val_accuracy: 0.5341 - lr: 5.0000e-06
Epoch 11/40
Epoch 11: ReduceLROnPlateau reducing learning rate to 2.499999936844688e-06.
33/33 [================] - 131s 4s/step - loss: 0.6574 - accuracy:
0.6486 - val_loss: 0.6644 - val_accuracy: 0.5341 - lr: 5.0000e-06
Epoch 12/40
33/33 [=============== ] - 130s 4s/step - loss: 0.6523 - accuracy:
0.6520 - val_loss: 0.6558 - val_accuracy: 0.5341 - lr: 2.5000e-06
Epoch 13/40
33/33 [================ ] - 134s 4s/step - loss: 0.6253 - accuracy:
0.6847 - val_loss: 0.6636 - val_accuracy: 0.5341 - lr: 2.5000e-06
Epoch 14/40
33/33 [==================] - 129s 4s/step - loss: 0.6216 - accuracy:
0.6728 - val_loss: 0.6584 - val_accuracy: 0.5341 - lr: 2.5000e-06
Epoch 15/40
33/33 [=============== ] - 131s 4s/step - loss: 0.6362 - accuracy:
0.6724 - val loss: 0.6620 - val accuracy: 0.5341 - lr: 2.5000e-06
Epoch 16/40
Epoch 16: ReduceLROnPlateau reducing learning rate to 1.249999968422344e-06.
33/33 [================] - 128s 4s/step - loss: 0.6295 - accuracy:
0.6648 - val_loss: 0.6646 - val_accuracy: 0.5341 - lr: 2.5000e-06
Epoch 17/40
33/33 [=============== ] - 136s 4s/step - loss: 0.6188 - accuracy:
0.6757 - val_loss: 0.6528 - val_accuracy: 0.5379 - lr: 1.2500e-06
Epoch 18/40
33/33 [================ ] - 196s 6s/step - loss: 0.6230 - accuracy:
0.6671 - val loss: 0.6614 - val accuracy: 0.5341 - lr: 1.2500e-06
Epoch 19/40
33/33 [================] - 164s 5s/step - loss: 0.6379 - accuracy:
0.6534 - val_loss: 0.6568 - val_accuracy: 0.5341 - lr: 1.2500e-06
Epoch 20/40
```

```
33/33 [=============] - 163s 5s/step - loss: 0.6094 - accuracy:
0.6700 - val loss: 0.6652 - val accuracy: 0.5341 - lr: 1.2500e-06
Epoch 21/40
Epoch 21: ReduceLROnPlateau reducing learning rate to 6.24999984211172e-07.
0.6752 - val_loss: 0.6637 - val_accuracy: 0.5341 - lr: 1.2500e-06
Epoch 22/40
33/33 [==============] - 162s 5s/step - loss: 0.6227 - accuracy:
0.6728 - val_loss: 0.6624 - val_accuracy: 0.5341 - lr: 6.2500e-07
Epoch 23/40
33/33 [==============] - 163s 5s/step - loss: 0.6178 - accuracy:
0.6686 - val loss: 0.6633 - val accuracy: 0.5341 - lr: 6.2500e-07
Epoch 24/40
33/33 [================] - 187s 6s/step - loss: 0.6235 - accuracy:
0.6667 - val_loss: 0.6616 - val_accuracy: 0.5341 - lr: 6.2500e-07
Epoch 25/40
0.6671 - val_loss: 0.6590 - val_accuracy: 0.5341 - lr: 6.2500e-07
Epoch 26/40
Epoch 26: ReduceLROnPlateau reducing learning rate to 3.12499992105586e-07.
33/33 [================] - 165s 5s/step - loss: 0.6050 - accuracy:
0.6761 - val_loss: 0.6620 - val_accuracy: 0.5341 - lr: 6.2500e-07
Epoch 27/40
0.6567 - val_loss: 0.6602 - val_accuracy: 0.5341 - lr: 3.1250e-07
Epoch 28/40
33/33 [=================] - 164s 5s/step - loss: 0.6053 - accuracy:
0.6818 - val_loss: 0.6588 - val_accuracy: 0.5341 - lr: 3.1250e-07
Epoch 29/40
0.6709 - val_loss: 0.6591 - val_accuracy: 0.5341 - lr: 3.1250e-07
Epoch 30/40
33/33 [==============] - 165s 5s/step - loss: 0.5994 - accuracy:
0.6818 - val_loss: 0.6578 - val_accuracy: 0.5341 - lr: 3.1250e-07
Epoch 31/40
33/33 [================= ] - ETA: 0s - loss: 0.6232 - accuracy: 0.6690
Epoch 31: ReduceLROnPlateau reducing learning rate to 1.56249996052793e-07.
33/33 [=================] - 167s 5s/step - loss: 0.6232 - accuracy:
0.6690 - val_loss: 0.6578 - val_accuracy: 0.5341 - lr: 3.1250e-07
Epoch 32/40
33/33 [================ ] - 165s 5s/step - loss: 0.6138 - accuracy:
0.6724 - val_loss: 0.6584 - val_accuracy: 0.5341 - lr: 1.5625e-07
Epoch 33/40
33/33 [=================] - 164s 5s/step - loss: 0.6156 - accuracy:
0.6780 - val_loss: 0.6582 - val_accuracy: 0.5341 - lr: 1.5625e-07
Epoch 34/40
0.6871 - val loss: 0.6607 - val accuracy: 0.5341 - lr: 1.5625e-07
33/33 [================] - 163s 5s/step - loss: 0.5976 - accuracy:
0.6965 - val_loss: 0.6604 - val_accuracy: 0.5341 - lr: 1.5625e-07
Epoch 36/40
Epoch 36: ReduceLROnPlateau reducing learning rate to 1e-07.
33/33 [=============] - 165s 5s/step - loss: 0.6138 - accuracy:
0.6719 - val_loss: 0.6597 - val_accuracy: 0.5341 - lr: 1.5625e-07
Epoch 37/40
0.6780 - val loss: 0.6604 - val accuracy: 0.5341 - lr: 1.0000e-07
Epoch 38/40
0.6866 - val_loss: 0.6609 - val_accuracy: 0.5341 - lr: 1.0000e-07
```

```
Epoch 39/40
         33/33 [============== ] - 127s 4s/step - loss: 0.6050 - accuracy:
         0.6795 - val_loss: 0.6614 - val_accuracy: 0.5341 - lr: 1.0000e-07
         Epoch 40/40
         33/33 [============== ] - 138s 4s/step - loss: 0.6055 - accuracy:
         0.6776 - val_loss: 0.6604 - val_accuracy: 0.5341 - lr: 1.0000e-07
In [12]: print(history.history.keys())
          plt.plot(history.history['accuracy'])
          plt.plot(history.history['val_accuracy'])
          plt.title('model accuracy')
          plt.ylabel('accuracy')
          plt.xlabel('epoch')
          plt.legend(['train', 'test'], loc='upper left')
          plt.show()
          plt.plot(history.history['loss'])
          plt.plot(history.history['val_loss'])
          plt.title('model loss')
          plt.ylabel('loss')
          plt.xlabel('epoch')
          plt.legend(['train', 'test'], loc='upper left')
          plt.show()
         dict_keys(['loss', 'accuracy', 'val_loss', 'val_accuracy', 'lr'])
                                   model accuracy
                      train
            0.700
                      test
            0.675
            0.650
          accuracy
            0.625
            0.600
            0.575
            0.550
            0.525
                              10
                                   15
                                         20
                                               25
                                                     30
                                                          35
                                       epoch
                                   model loss
                    train
            1.0
                    test
            0.9
         0.8
            0.7
            0.6
                       5
                                  15
                                             25
                            10
                                       20
                                                         35
                                                              40
                                     epoch
```