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In [1]: # This Python 3 environment comes with many helpful analytics libraries instal
        # It is defined by the kaggle/python Docker image: https://github.com/kaggle/d
        ocker-python
        # For example, here's several helpful packages to load
        import numpy as np # linear algebra
        import pandas as pd # data processing, CSV file I/O (e.g. pd.read csv)
        # Input data files are available in the read-only "../input/" directory
        # For example, running this (by clicking run or pressing Shift+Enter) will lis
        t all files under the input directory
        import os
        for dirname, , filenames in os.walk('/kaggle/input'):
            for filename in filenames:
                print(os.path.join(dirname, filename))
        # You can write up to 20GB to the current directory (/kaggle/working/) that ge
        ts preserved as output when you create a version using "Save & Run All"
        # You can also write temporary files to /kaqqle/temp/, but they won't be saved
        outside of the current session
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In [2]: # importing libraries
        import tensorflow as tf
        from tensorflow.keras.layers import Input, Lambda, Dense, Flatten
        from tensorflow.keras.models import Model
        from tensorflow.keras.applications.resnet50 import ResNet50
        from tensorflow.keras.applications.resnet50 import preprocess input
        from tensorflow.keras.preprocessing import image
        from tensorflow.keras.preprocessing.image import ImageDataGenerator,load img
        from tensorflow.keras.models import Sequential
        import numpy as np
        from glob import glob
        import matplotlib.pyplot as plt
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In [3]: image_set = "../input/lung-and-colon-cancer-histopathological-images/lung_col
        on image set/lung image sets"
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In [4]: SIZE X = SIZE Y = 224
         datagen = tf.keras.preprocessing.image.ImageDataGenerator(validation split =
         0.2)
         train_set = datagen.flow_from_directory(image_set,
                                                 class mode = "categorical",
                                                 target size = (SIZE X,SIZE Y),
                                                 color mode="rgb",
                                                 batch_size = 128,
                                                 shuffle = False,
                                                 subset='training',
                                                 seed = 42)
         validate set = datagen.flow from directory(image set,
                                                 class_mode = "categorical",
                                                 target size = (SIZE X, SIZE Y),
                                                 color mode="rgb",
                                                 batch_size = 128,
                                                 shuffle = False,
                                                 subset='validation',
                                                 seed = 42)
In [5]: IMAGE_SIZE = [224, 224]
 In [6]: resnet = ResNet50(input shape=IMAGE SIZE + [3], weights='imagenet', include to
         p=False)
In [7]: # don't train existing weights
         for layer in resnet.layers:
             layer.trainable = False
In [8]: | flatten = Flatten()(resnet.output)
         dense = Dense(256, activation = 'relu')(flatten)
         dense = Dense(128, activation = 'relu')(dense)
         prediction = Dense(3, activation = 'softmax')(dense)
In [9]: #creating a model
         model = Model(inputs = resnet.input, outputs = prediction )
In [10]: model.summary()
In [11]: # COMPILING THE MODEL
         model.compile(loss = 'categorical crossentropy', optimizer = 'adam', metrics =
         ['accuracy'])
In [12]: #executing the model
         history = model.fit generator(train set, validation data = (validate set), epo
         chs = 5, verbose = 1)
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In [13]: | # plotting the loss
         plt.plot(history.history['loss'],label = 'train_loss')
         plt.plot(history.history['val_loss'], label = 'testing_loss')
         plt.title('loss')
         plt.legend()
         plt.show()
In [14]: # Both Validation and Training accuracy is shown here
         plt.plot(history.history['accuracy'], label='training_accuracy')
         plt.plot(history.history['val accuracy'], label='validation accuracy')
         plt.title('Accuracy')
         plt.legend()
         plt.show()
In [15]: # CHECKING THE CONFUSION MATRIX
         from sklearn.metrics import classification report
         from sklearn.metrics import confusion_matrix
         from sklearn.metrics import f1 score
         Y_pred = model.predict_generator(validate_set)
         y_pred = np.argmax(Y_pred ,axis =1)
         print('Confusion Matrix')
         confusion matrix = confusion matrix(validate set.classes, y pred)
         print(confusion_matrix)
         print('Classification Report')
         target_names = ['aca','n', 'scc']
         print(classification_report(validate_set.classes, y_pred, target_names=target_
         names))
In [16]: result = model.evaluate(validate_set,batch_size=128)
         print("test loss, test accuracy",result)
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In [ ]:
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