```
!gcloud config set auth/impersonate_service_account "805566722650-compute@developer.gserviceaccount.com"
```

```
Updated property [auth/impersonate_service_account].
```

```
import sys

if "google.colab" in sys.modules:

    from google.colab import auth

    auth.authenticate_user()
```

Double-click (or enter) to edit

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```
from google.colab import drive
drive.mount('/content/drive', force_remount=True)
```

```
Mounted at /content/drive
```

```
import pandas as pd
import numpy as np
import tensorflow as tf
from tensorflow import keras
from keras import Sequential
import matplotlib.pyplot as plt
from PIL import Image
import os
import pathlib
import random
import glob



# Create data directory in your google MyDrive folder.
# Copy all three folders in your data directory to MyDrive->data folder in your google drive
# Then Execute below code
def get_data_dir_path():
  path = '/content/drive/MyDrive/data/'
  data_dir_path= pathlib.Path(path)
  data_dir_file = str(pathlib.Path(path)) + "/*.*"
  return data_dir_path


# Create filter path to retive all the png files
def filter_path():
  data_dir_path_benign = str(pathlib.Path(os.path.join(get_data_dir_path(),'benign'))) + "/*.png"
  data_dir_path_normal = str(pathlib.Path(os.path.join(get_data_dir_path(),'normal'))) + "/*.png"
  data_dir_path_malignant = str(pathlib.Path(os.path.join(get_data_dir_path(),'malignant'))) + "/*.png"
  return data_dir_path_benign,data_dir_path_normal,data_dir_path_malignant


# Use the filter to retrieve the corresponsing image counts
def image_count():
  data_dir_path_benign,data_dir_path_normal,data_dir_path_malignant = filter_path()
  img_count_benign = glob.glob(data_dir_path_benign,recursive=True)
  img_count_normal = glob.glob(data_dir_path_normal,recursive=True)
  img_count_malignant = glob.glob(data_dir_path_malignant,recursive=True)
  img_total_count = len(img_count_benign) + len(img_count_normal) + len(img_count_malignant)
  print("Total  Images: ", img_total_count)
  print("Benign (non-dangerous) Images: {}({})".format(len(img_count_benign), round(len(img_count_benign)*100/img_total_count, 2)))
  print("Malignant (dangerous) Images: {}({})".format(len(img_count_normal), round(len(img_count_normal)*100/img_total_count, 2)))
  print("Normal (No Traces) Images: {}({})".format(len(img_count_malignant), round(len(img_count_malignant)*100/img_total_count, 2)))
  #return img_total_count,len(img_count_benign),len(img_count_normal),len(img_count_malignant)



#Display the image totals and respective class image counts
image_count()
```

```
Total  Images:  1587
    Benign (non-dangerous) Images: 900(56.71)
    Malignant (dangerous) Images: 266(16.76)
```

```
Normal (No Traces) Images: 421(26.53)
```

```
# Configure batch size,img size and create train, validation split use function call configure_for_performance to fine tune performance
def create_data_sets():
  batch_size = 40
  img_height = 224
  img_width = 224
  train_ds = tf.keras.utils.image_dataset_from_directory(get_data_dir_path(),validation_split=0.3,subset="training",seed=123,image_size=(img
  val_ds = tf.keras.utils.image_dataset_from_directory(get_data_dir_path(),validation_split=0.3,subset="validation",seed=123,image_size=(img
  val_ds.class_names
  return train_ds,val_ds


train_ds,val_ds = create_data_sets()
```

```
Found 1587 files belonging to 3 classes.
Using 1111 files for training.
Found 1587 files belonging to 3 classes.
Using 476 files for validation.
```

```
def create_model(img_height=224,img_width=224):
  model = tf.keras.Sequential([
    tf.keras.layers.Rescaling(1./255, input_shape=(img_height, img_width, 3)),

    tf.keras.layers.Conv2D(16, 3, padding='same', activation='relu'),
    tf.keras.layers.MaxPooling2D(),

    tf.keras.layers.Conv2D(32, 3, padding='same', activation='relu'),
    tf.keras.layers.MaxPooling2D(),

    tf.keras.layers.Conv2D(64, 3, padding='same', activation='relu'),
    tf.keras.layers.MaxPooling2D(),

    tf.keras.layers.Conv2D(128, 3, padding='same', activation='relu'),
    tf.keras.layers.MaxPooling2D(),

    tf.keras.layers.Dropout(0.3),
    tf.keras.layers.Flatten(),
    tf.keras.layers.Dense(64, activation='relu'),
    tf.keras.layers.Dense(3,activation="softmax")
  ])
  return model



model = create_model()
model.summary()
```

```
/usr/local/lib/python3.11/dist-packages/keras/src/layers/preprocessing/tf_data_layer.py:19: UserWarning: Do not pass an `input_shape`/`i
  super().__init__(**kwargs)
```

Model: "sequential"

Layer (type)	Output Shape	Param #
rescaling (Rescaling)	(None, 224, 224, 3)	0
conv2d (Conv2D)	(None, 224, 224, 16)	448
max_pooling2d (MaxPooling2D)	(None, 112, 112, 16)	0
conv2d_1 (Conv2D)	(None, 112, 112, 32)	4,640
max_pooling2d_1 (MaxPooling2D)	(None, 56, 56, 32)	0
conv2d_2 (Conv2D)	(None, 56, 56, 64)	18,496
max_pooling2d_2 (MaxPooling2D)	(None, 28, 28, 64)	0
conv2d_3 (Conv2D)	(None, 28, 28, 128)	73,856
max_pooling2d_3 (MaxPooling2D)	(None, 14, 14, 128)	0
dropout (Dropout)	(None, 14, 14, 128)	0
flatten (Flatten)	(None, 25088)	0
dense (Dense)	(None, 64)	1,605,696
dense_1 (Dense)	(None, 3)	195

Total params: 1,703,331 (6.50 MB)
Trainable params: 1,703,331 (6.50 MB)

```
def train_model(batch_size = 32,epochs=4):
  train_ds,val_ds = create_data_sets()
  model.compile(optimizer="Adam",
            loss=tf.keras.losses.SparseCategoricalCrossentropy(from_logits=True),
            metrics=["accuracy"])

  history = model.fit(train_ds,
                    epochs=epochs,
                    validation_data=val_ds,
                    batch_size=batch_size)
  return history


#Train the model
history=train_model(batch_size = 32,epochs=4)
history.history.keys()
```

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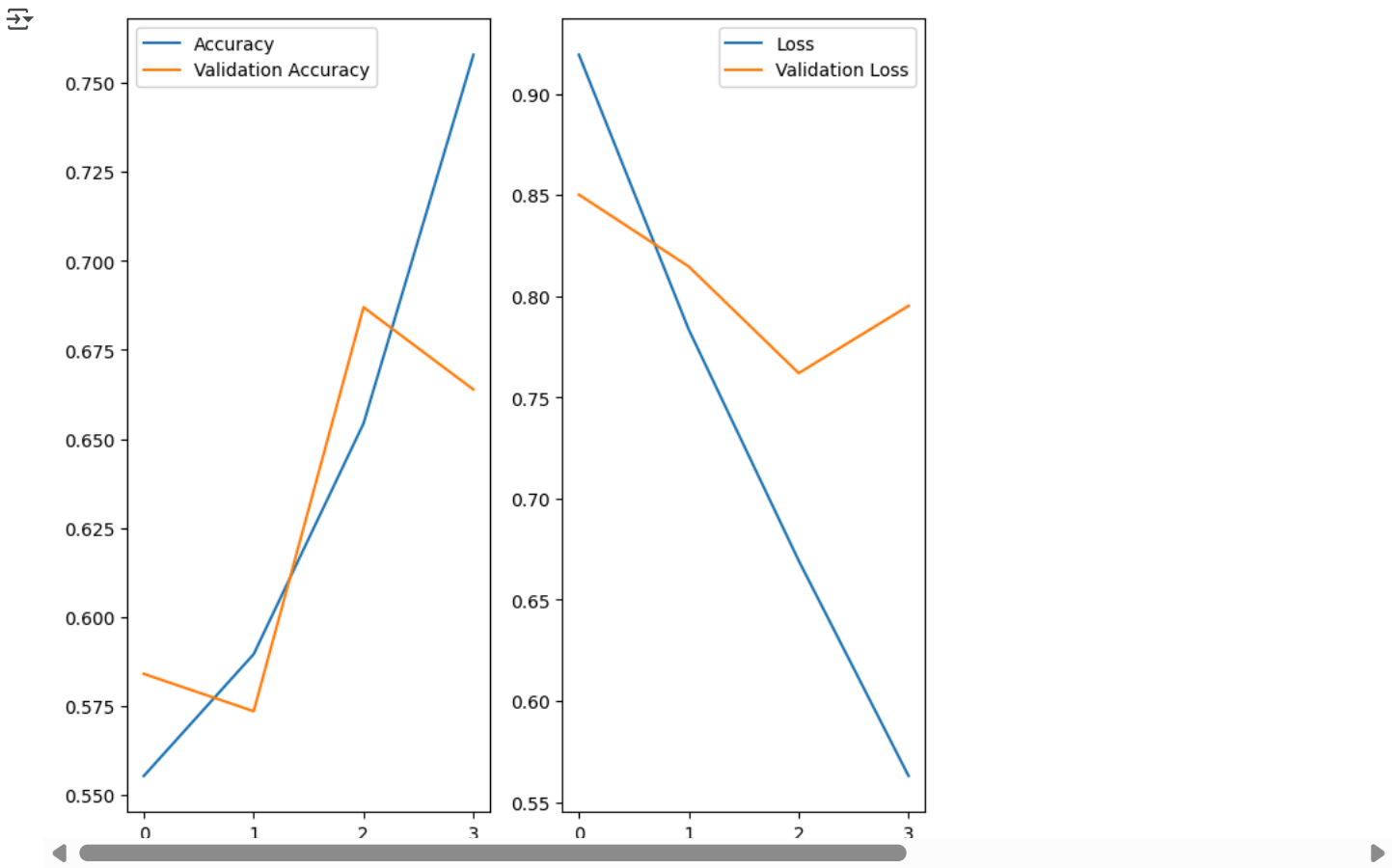
Epoch 1/4
/usr/local/lib/python3.11/dist-packages/keras/src/backend/tensorflow/nn.py:708: UserWarning: "`sparse_categorical_crossentropy` received
  output, from_logits = _get_logits(
28/28 ━━━━━━━━━━━━━━━━━━━━ 186s 6s/step - accuracy: 0.5087 - loss: 0.9661 - val_accuracy: 0.5840 - val_loss: 0.8502
Epoch 2/4
28/28 ━━━━━━━━━━━━━━━━━━━━ 104s 3s/step - accuracy: 0.6001 - loss: 0.7965 - val_accuracy: 0.5735 - val_loss: 0.8146
Epoch 3/4
28/28 ━━━━━━━━━━━━━━━━━━━━ 150s 3s/step - accuracy: 0.6356 - loss: 0.6609 - val_accuracy: 0.6870 - val_loss: 0.7620
Epoch 4/4
28/28 ━━━━━━━━━━━━━━━━━━━━ 143s 3s/step - accuracy: 0.7522 - loss: 0.5743 - val_accuracy: 0.6639 - val_loss: 0.7952
dict_keys(['accuracy', 'loss', 'val_accuracy', 'val_loss'])
```

```
#Plot the graph acc/val accuracy versus loss/validation loss
epochs=4
acc = history.history['accuracy']
val_acc =  history.history['val_accuracy']
loss = history.history['loss']
val_loss = history.history['val_loss']
ep_range = range(epochs)
plt.figure(figsize=(8,8))
plt.subplot(1,2,1)
plt.plot(ep_range,acc,label='Accuracy')
plt.plot(ep_range,val_acc,label="Validation Accuracy")
```

```
plt.legend()

plt.subplot(1,2,2)
plt.plot(ep_range,loss,label='Loss')
plt.plot(ep_range,val_loss,label="Validation Loss")
plt.legend()
plt.show()
```




```
#Evaluate model with validation data set
train_ds,val_ds = create_data_sets()
loss, acc = model.evaluate(val_ds)
print('model, accuracy: {:5.2f}%'.format(100 * acc))
```

```
Found 1587 files belonging to 3 classes.
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Using 476 files for validation.
12/12 ━━━━━━━━━━━━━━━━━━━━ 21s 1s/step - accuracy: 0.6545 - loss: 0.8383
model, accuracy: 66.39%
```

```
#Pedict the model using Validation data
plt.figure(figsize=(15, 15))
class_names = val_ds.class_names
result = ' | False'
for images, labels in val_ds.take(1):
    for i in range(25):

        ax = plt.subplot(5, 5, i + 1)
        img = images[i].numpy().astype("uint8")
        img = tf.expand_dims(img, axis=0)
        predication=""
        predictions = model.predict(img)
        predicted_class = np.argmax(predictions)
        if class_names[predicted_class] == class_names[labels[i]]:
            result = ' | TRUE'

        plt.imshow(images[i].numpy().astype("uint8"))
        plt.title('Predicated:'+ class_names[predicted_class]+result)
```

```
            plt.axis("off")
```

```
1/1 ━━━━━━━━━━━━━━━━━━━━ 0s 160ms/step
1/1 ━━━━━━━━━━━━━━━━━━━━ 0s 67ms/step
1/1 ━━━━━━━━━━━━━━━━━━━━ 0s 59ms/step
1/1 ━━━━━━━━━━━━━━━━━━━━ 0s 63ms/step
1/1 ━━━━━━━━━━━━━━━━━━━━ 0s 54ms/step
1/1 ━━━━━━━━━━━━━━━━━━━━ 0s 53ms/step
1/1 ━━━━━━━━━━━━━━━━━━━━ 0s 62ms/step
1/1 ━━━━━━━━━━━━━━━━━━━━ 0s 53ms/step
1/1 ━━━━━━━━━━━━━━━━━━━━ 0s 55ms/step
1/1 ━━━━━━━━━━━━━━━━━━━━ 0s 54ms/step
1/1 ━━━━━━━━━━━━━━━━━━━━ 0s 57ms/step
1/1 ━━━━━━━━━━━━━━━━━━━━ 0s 64ms/step
1/1 ━━━━━━━━━━━━━━━━━━━━ 0s 57ms/step
1/1 ━━━━━━━━━━━━━━━━━━━━ 0s 54ms/step
1/1 ━━━━━━━━━━━━━━━━━━━━ 0s 55ms/step
1/1 ━━━━━━━━━━━━━━━━━━━━ 0s 55ms/step
1/1 ━━━━━━━━━━━━━━━━━━━━ 0s 53ms/step
1/1 ━━━━━━━━━━━━━━━━━━━━ 0s 57ms/step
1/1 ━━━━━━━━━━━━━━━━━━━━ 0s 56ms/step
1/1 ━━━━━━━━━━━━━━━━━━━━ 0s 57ms/step
1/1 ━━━━━━━━━━━━━━━━━━━━ 0s 56ms/step
1/1 ━━━━━━━━━━━━━━━━━━━━ 0s 60ms/step
1/1 ━━━━━━━━━━━━━━━━━━━━ 0s 61ms/step
1/1 ━━━━━━━━━━━━━━━━━━━━ 0s 58ms/step
1/1 ━━━━━━━━━━━━━━━━━━━━ 0s 95ms/step
```