

In [1]:

```
"""
Tumor Classssification using SqueezeNet
Author: Amruth Karun M V
Date: 12-Nov-2021
"""

import os
import cv2
import numpy as np
import tensorflow as tf
from tensorflow import keras
from tensorflow.keras.optimizers import Adam
from tensorflow.keras.models import Model
from tensorflow.keras.layers import (
    Dense, Input, Dropout, Flatten,
    Conv2D, MaxPooling2D, concatenate,
    GlobalAveragePooling2D, Activation
)

from sklearn import metrics
import matplotlib.pyplot as plt
%matplotlib inline

TRAIN_PATH = "../input/brain-tumor-mri-dataset/Training"
TEST_PATH = "../input/brain-tumor-mri-dataset/Testing"
CLASS_NAMES = ['Glioma', 'Meningioma', 'No-tumor', 'Pituitary']
EPOCHS = 100
BATCH_SIZE = 128
LEARNING_RATE = 0.001
```

```

def plot_sample_images():
    """
    Plots sample images for each class
    Arguments: None
    Returns: Plots sample data
    """

    plt.figure(figsize=(10, 10))
    sample_image_path = ['/glioma/Tr-gl_0010.jpg', '/meningioma/Tr-me_0010.jpg',
                        '/notumor/Tr-no_0010.jpg', '/pituitary/Tr-pi_0010.jpg']
    for i in range(len(CLASS_NAMES)):
        ax = plt.subplot(2, 2, i + 1)
        img = cv2.imread(TRAIN_PATH + sample_image_path[i])
        img = cv2.resize(img, (128, 128))
        plt.imshow(img)
        plt.title(CLASS_NAMES[i])

def load_data(input_path, shuffle=False):
    """
    Loads input data fro directory
    Arguments:
        input_path -- input data path
        shuffle     -- whether data needs to be shuffled or not
    Returns: Data generator
    """

    data_generator = keras.preprocessing.image.ImageDataGenerator()
    data_generator = data_generator.flow_from_directory(directory=input_path, target_size=(224,224),
                                                    shuffle=shuffle, class_mode="categorical")

```

```
return data_generator
```

```
def fire_module(x, fire_id, squeeze=16, expand=64):
```

```
    """
```

```
    Fire module for SqueezeNet
```

```
    Arguments:
```

```
        x          -- input
```

```
        fire_id    -- id for the module
```

```
        squeeze    -- No. of squeeze layer filters
```

```
        expand     -- No. of expand layer filters
```

```
    Returns: Concatenated output of fire_module
```

```
    """
```

```
s_id = 'fire' + str(fire_id) + '/'
```

```
x = Conv2D(squeeze, (1, 1), padding='valid', name=s_id + 'sq1x1')(x)
```

```
x = Activation('relu', name=s_id + 'relu_sq1x1')(x)
```

```
left = Conv2D(expand, (1, 1), padding='valid', name=s_id + 'exp1x1')(x)
```

```
left = Activation('relu', name=s_id + 'relu_exp1x1')(left)
```

```
right = Conv2D(expand, (3, 3), padding='same', name=s_id + 'exp3x3')(x)
```

```
right = Activation('relu', name=s_id + 'relu_exp3x3')(right)
```

```
x = concatenate([left, right], axis=3, name=s_id + 'concat')
```

```
return x
```

```
def load_model():
```

```
"""
```

Creates a keras SqueezeNet model

Arguments: None

Returns: SqueezeNet Model

```
"""
```

```
img_input = Input(shape=(224, 224, 3))
```

```
x = Conv2D(64, (3, 3), strides=(2, 2), padding='valid', name='conv1')(img_input)
```

```
x = Activation('relu', name='relu_conv1')(x)
```

```
x = MaxPooling2D(pool_size=(3, 3), strides=(2, 2), name='pool1')(x)
```

```
x = fire_module(x, fire_id=2, squeeze=16, expand=64)
```

```
x = fire_module(x, fire_id=3, squeeze=16, expand=64)
```

```
x = MaxPooling2D(pool_size=(3, 3), strides=(2, 2), name='pool3')(x)
```

```
x = fire_module(x, fire_id=4, squeeze=32, expand=128)
```

```
x = fire_module(x, fire_id=5, squeeze=32, expand=128)
```

```
x = MaxPooling2D(pool_size=(3, 3), strides=(2, 2), name='pool5')(x)
```

```
x = fire_module(x, fire_id=6, squeeze=48, expand=192)
```

```
x = fire_module(x, fire_id=7, squeeze=48, expand=192)
```

```
x = fire_module(x, fire_id=8, squeeze=64, expand=256)
```

```
x = fire_module(x, fire_id=9, squeeze=64, expand=256)
```

```
# Classification block
```

```
x = Dropout(0.5, name='drop9')(x)
```

```
x = Conv2D(4, (1, 1), padding='valid', name='conv10')(x)
```

```
x = Activation('relu', name='relu_conv10')(x)
```

```
x = GlobalAveragePooling2D()(x)
```

```
x = Activation('softmax', name='loss')(x)
```

```
model = Model(img_input, x, name='SqueezeNet')

model.summary()

opt = Adam(learning_rate=LEARNING_RATE)
model.compile(loss = keras.losses.categorical_crossentropy, optimizer=opt, metrics=['accuracy'])

return model
```

```
def plot_curves(history):
```

```
    """
```

```
    Plots loss and accuracy and loss plots for  
    training and validation datasets
```

```
    Arguments:
```

```
        history -- training history
```

```
    Returns: None
```

```
    """
```

```
    plt.plot(history.history['loss'], label="Training loss")
    plt.plot(history.history['val_loss'], label="Validation loss")
    plt.legend()
    plt.title('Training Loss VS Validation Loss')
    plt.show()
```

```
    plt.plot(history.history['accuracy'], label="Training accuracy")
    plt.plot(history.history['val_accuracy'], label="Validation accuracy")
    plt.title('Training Accuracy VS Validation Accuracy')
    plt.legend()
    plt.show()
```

```

def evaluate_model(model, input_path):
    """
    Evaluates the model and displays
    the confusion matrix
    Arguments:
        model          -- trained model
        input_path     -- input data path
    Returns: Model score and confusion matrix
    """

    data_generator = load_data(input_path)
    predictions = model.predict(data_generator, BATCH_SIZE)
    y_pred = np.argmax(predictions, axis=1)
    y_true = data_generator.classes

    print("Score = ", model.evaluate(data_generator))
    print("Accuracy = ", metrics.accuracy_score(y_true, y_pred))
    cm = metrics.confusion_matrix(y_true, y_pred)
    metrics.ConfusionMatrixDisplay(cm, display_labels=CLASS_NAMES).plot(cmap=plt.cm.Blues,
                                                                           xticks_rotation='vertical')

    plt.show()

def train_model():
    """
    Trains SqueezeNet model and saves the
    trained weights to an H5 file.
    Arguments: None
    Returns: None
    """

```

```
"""
```

```
train_generator = load_data(TRAIN_PATH, True)
```

```
val_generator = load_data(TEST_PATH, True)
```

```
# Loads SqueezeNet model
```

```
model = load_model()
```

```
earlystop = keras.callbacks.EarlyStopping(monitor='loss', min_delta=1e-11, patience=10)
```

```
reduce_lr = keras.callbacks.ReduceLROnPlateau(monitor='val_loss', factor=0.2,  
                                                patience=6, verbose=1)
```

```
model_callbacks = [earlystop, reduce_lr]
```

```
history = model.fit(  
    train_generator,  
    batch_size=BATCH_SIZE,  
    epochs=EPOCHS,  
    validation_data=val_generator,  
    validation_steps=val_generator.samples//BATCH_SIZE,  
    steps_per_epoch=train_generator.samples//BATCH_SIZE,  
    callbacks=model_callbacks)
```

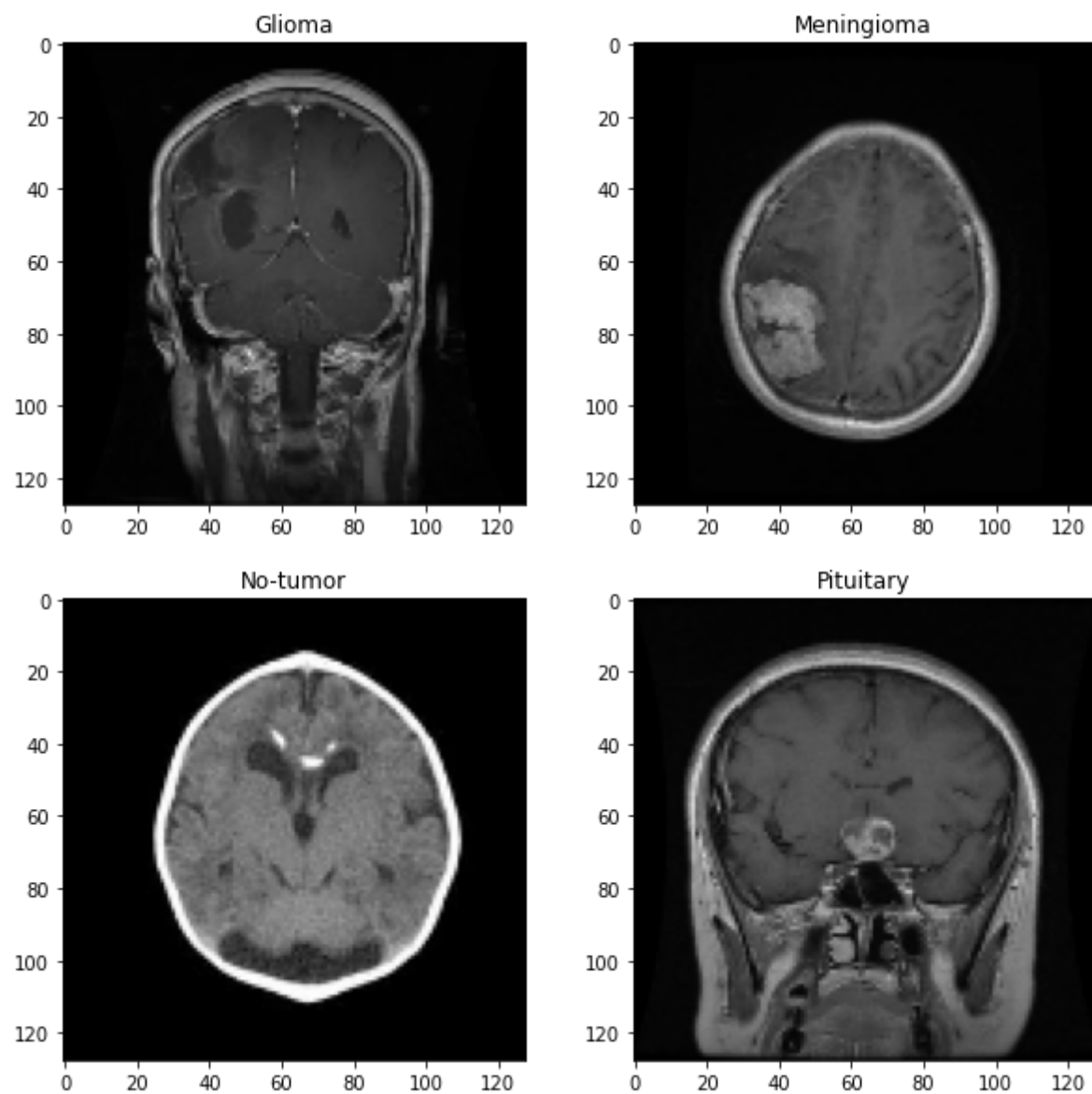
```
plot_curves(history)
```

```
model.save_weights("model_squeezenet.h5")
```

```
print("Model saved successfully!")
```

```
return model
```

```
In [2]: plot_sample_images()
```



In [3]:

```
# Train the model
model = train_model()
print("Confusion matrix for train data: ")
evaluate_model(model, TRAIN_PATH)
print("Confusion matrix for val data: ")
evaluate_model(model, TEST_PATH)
```

Found 5712 images belonging to 4 classes.

Found 1311 images belonging to 4 classes.

Model: "SqueezeNet"

Layer (type)	Output Shape	Param #	Connected to
=====			
input_1 (InputLayer)	[(None, 224, 224, 3)]	0	

conv1 (Conv2D)	(None, 111, 111, 64)	1792	input_1[0][0]

relu_conv1 (Activation)	(None, 111, 111, 64)	0	conv1[0][0]

pool1 (MaxPooling2D)	(None, 55, 55, 64)	0	relu_conv1[0][0]

fire2/sq1x1 (Conv2D)	(None, 55, 55, 16)	1040	pool1[0][0]

fire2/relu_sq1x1 (Activation)	(None, 55, 55, 16)	0	fire2/sq1x1[0][0]

fire2/exp1x1 (Conv2D)	(None, 55, 55, 64)	1088	fire2/relu_sq1x1[0][0]

fire2/exp3x3 (Conv2D)	(None, 55, 55, 64)	9280	fire2/relu_sq1x1[0][0]

fire2/relu_exp1x1 (Activation)	(None, 55, 55, 64)	0	fire2/exp1x1[0][0]

fire2/relu_exp3x3 (Activation)	(None, 55, 55, 64)	0	fire2/exp3x3[0][0]

fire2/concat (Concatenate)	(None, 55, 55, 128)	0	fire2/relu_exp1x1[0][0] fire2/relu_exp3x3[0][0]

fire3/sq1x1 (Conv2D)	(None, 55, 55, 16)	2064	fire2/concat[0][0]

fire3/relu_sq1x1 (Activation)	(None, 55, 55, 16)	0	fire3/sq1x1[0][0]

fire3/exp1x1 (Conv2D)	(None, 55, 55, 64)	1088	fire3/relu_sq1x1[0][0]
fire3/exp3x3 (Conv2D)	(None, 55, 55, 64)	9280	fire3/relu_sq1x1[0][0]
fire3/relu_exp1x1 (Activation)	(None, 55, 55, 64)	0	fire3/exp1x1[0][0]
fire3/relu_exp3x3 (Activation)	(None, 55, 55, 64)	0	fire3/exp3x3[0][0]
fire3/concat (Concatenate)	(None, 55, 55, 128)	0	fire3/relu_exp1x1[0][0] fire3/relu_exp3x3[0][0]
pool3 (MaxPooling2D)	(None, 27, 27, 128)	0	fire3/concat[0][0]
fire4/sq1x1 (Conv2D)	(None, 27, 27, 32)	4128	pool3[0][0]
fire4/relu_sq1x1 (Activation)	(None, 27, 27, 32)	0	fire4/sq1x1[0][0]
fire4/exp1x1 (Conv2D)	(None, 27, 27, 128)	4224	fire4/relu_sq1x1[0][0]
fire4/exp3x3 (Conv2D)	(None, 27, 27, 128)	36992	fire4/relu_sq1x1[0][0]
fire4/relu_exp1x1 (Activation)	(None, 27, 27, 128)	0	fire4/exp1x1[0][0]
fire4/relu_exp3x3 (Activation)	(None, 27, 27, 128)	0	fire4/exp3x3[0][0]
fire4/concat (Concatenate)	(None, 27, 27, 256)	0	fire4/relu_exp1x1[0][0] fire4/relu_exp3x3[0][0]
fire5/sq1x1 (Conv2D)	(None, 27, 27, 32)	8224	fire4/concat[0][0]

fire5/relu_sq1x1 (Activation)	(None, 27, 27, 32)	0	fire5/sq1x1[0][0]
fire5/exp1x1 (Conv2D)	(None, 27, 27, 128)	4224	fire5/relu_sq1x1[0][0]
fire5/exp3x3 (Conv2D)	(None, 27, 27, 128)	36992	fire5/relu_sq1x1[0][0]
fire5/relu_exp1x1 (Activation)	(None, 27, 27, 128)	0	fire5/exp1x1[0][0]
fire5/relu_exp3x3 (Activation)	(None, 27, 27, 128)	0	fire5/exp3x3[0][0]
fire5/concat (Concatenate)	(None, 27, 27, 256)	0	fire5/relu_exp1x1[0][0] fire5/relu_exp3x3[0][0]
pool5 (MaxPooling2D)	(None, 13, 13, 256)	0	fire5/concat[0][0]
fire6/sq1x1 (Conv2D)	(None, 13, 13, 48)	12336	pool5[0][0]
fire6/relu_sq1x1 (Activation)	(None, 13, 13, 48)	0	fire6/sq1x1[0][0]
fire6/exp1x1 (Conv2D)	(None, 13, 13, 192)	9408	fire6/relu_sq1x1[0][0]
fire6/exp3x3 (Conv2D)	(None, 13, 13, 192)	83136	fire6/relu_sq1x1[0][0]
fire6/relu_exp1x1 (Activation)	(None, 13, 13, 192)	0	fire6/exp1x1[0][0]
fire6/relu_exp3x3 (Activation)	(None, 13, 13, 192)	0	fire6/exp3x3[0][0]
fire6/concat (Concatenate)	(None, 13, 13, 384)	0	fire6/relu_exp1x1[0][0] fire6/relu_exp3x3[0][0]
fire7/sq1x1 (Conv2D)	(None, 13, 13, 48)	18480	fire6/concat[0][0]

fire7/relu_sq1x1 (Activation)	(None, 13, 13, 48)	0	fire7/sq1x1[0][0]
fire7/exp1x1 (Conv2D)	(None, 13, 13, 192)	9408	fire7/relu_sq1x1[0][0]
fire7/exp3x3 (Conv2D)	(None, 13, 13, 192)	83136	fire7/relu_sq1x1[0][0]
fire7/relu_exp1x1 (Activation)	(None, 13, 13, 192)	0	fire7/exp1x1[0][0]
fire7/relu_exp3x3 (Activation)	(None, 13, 13, 192)	0	fire7/exp3x3[0][0]
fire7/concat (Concatenate)	(None, 13, 13, 384)	0	fire7/relu_exp1x1[0][0] fire7/relu_exp3x3[0][0]
fire8/sq1x1 (Conv2D)	(None, 13, 13, 64)	24640	fire7/concat[0][0]
fire8/relu_sq1x1 (Activation)	(None, 13, 13, 64)	0	fire8/sq1x1[0][0]
fire8/exp1x1 (Conv2D)	(None, 13, 13, 256)	16640	fire8/relu_sq1x1[0][0]
fire8/exp3x3 (Conv2D)	(None, 13, 13, 256)	147712	fire8/relu_sq1x1[0][0]
fire8/relu_exp1x1 (Activation)	(None, 13, 13, 256)	0	fire8/exp1x1[0][0]
fire8/relu_exp3x3 (Activation)	(None, 13, 13, 256)	0	fire8/exp3x3[0][0]
fire8/concat (Concatenate)	(None, 13, 13, 512)	0	fire8/relu_exp1x1[0][0] fire8/relu_exp3x3[0][0]
fire9/sq1x1 (Conv2D)	(None, 13, 13, 64)	32832	fire8/concat[0][0]

fire9/relu_sq1x1 (Activation)	(None, 13, 13, 64)	0	fire9/sq1x1[0][0]
fire9/exp1x1 (Conv2D)	(None, 13, 13, 256)	16640	fire9/relu_sq1x1[0][0]
fire9/exp3x3 (Conv2D)	(None, 13, 13, 256)	147712	fire9/relu_sq1x1[0][0]
fire9/relu_exp1x1 (Activation)	(None, 13, 13, 256)	0	fire9/exp1x1[0][0]
fire9/relu_exp3x3 (Activation)	(None, 13, 13, 256)	0	fire9/exp3x3[0][0]
fire9/concat (Concatenate)	(None, 13, 13, 512)	0	fire9/relu_exp1x1[0][0] fire9/relu_exp3x3[0][0]
drop9 (Dropout)	(None, 13, 13, 512)	0	fire9/concat[0][0]
conv10 (Conv2D)	(None, 13, 13, 4)	2052	drop9[0][0]
relu_conv10 (Activation)	(None, 13, 13, 4)	0	conv10[0][0]
global_average_pooling2d (GlobalAveragePooling2D)	(None, 4)	0	relu_conv10[0][0]
loss (Activation)	(None, 4)	0	global_average_pooling2d[0][0]
=====			
Total params: 724,548			
Trainable params: 724,548			
Non-trainable params: 0			

2021-11-12 08:48:55.882065: I tensorflow/compiler/mlir/mlir_graph_optimization_pass.cc:185] None of the ML IR Optimization Passes are enabled (registered 2)

Epoch 1/100

2021-11-12 08:48:58.326635: I tensorflow/stream_executor/cuda/cuda_dnn.cc:369] Loaded cuDNN version 8005

44/44 [=====] - 21s 296ms/step - loss: 1.3810 - accuracy: 0.2699 - val_loss: 1.3854 - val_accuracy: 0.2438
Epoch 2/100
44/44 [=====] - 11s 250ms/step - loss: 1.2917 - accuracy: 0.4009 - val_loss: 1.3228 - val_accuracy: 0.3844
Epoch 3/100
44/44 [=====] - 10s 222ms/step - loss: 1.0541 - accuracy: 0.5431 - val_loss: 1.1311 - val_accuracy: 0.5125
Epoch 4/100
44/44 [=====] - 9s 201ms/step - loss: 0.8827 - accuracy: 0.6009 - val_loss: 0.9056 - val_accuracy: 0.6562
Epoch 5/100
44/44 [=====] - 8s 180ms/step - loss: 0.8622 - accuracy: 0.6200 - val_loss: 0.8727 - val_accuracy: 0.5781
Epoch 6/100
44/44 [=====] - 7s 165ms/step - loss: 0.7619 - accuracy: 0.6435 - val_loss: 0.8873 - val_accuracy: 0.6375
Epoch 7/100
44/44 [=====] - 7s 156ms/step - loss: 0.7379 - accuracy: 0.6903 - val_loss: 0.8637 - val_accuracy: 0.6562
Epoch 8/100
44/44 [=====] - 7s 162ms/step - loss: 0.7837 - accuracy: 0.6626 - val_loss: 0.7444 - val_accuracy: 0.6875
Epoch 9/100
44/44 [=====] - 6s 147ms/step - loss: 0.6979 - accuracy: 0.7003 - val_loss: 0.8722 - val_accuracy: 0.6562
Epoch 10/100
44/44 [=====] - 6s 147ms/step - loss: 0.7253 - accuracy: 0.7138 - val_loss: 0.8780 - val_accuracy: 0.6594
Epoch 11/100

44/44 [=====] - 6s 140ms/step - loss: 0.6598 - accuracy: 0.7457 - val_loss: 1.118
4 - val_accuracy: 0.6250
Epoch 12/100
44/44 [=====] - 6s 135ms/step - loss: 0.7004 - accuracy: 0.7422 - val_loss: 0.682
3 - val_accuracy: 0.7344
Epoch 13/100
44/44 [=====] - 6s 138ms/step - loss: 0.5146 - accuracy: 0.8068 - val_loss: 0.662
0 - val_accuracy: 0.7563
Epoch 14/100
44/44 [=====] - 6s 129ms/step - loss: 0.5036 - accuracy: 0.8132 - val_loss: 0.755
7 - val_accuracy: 0.7031
Epoch 15/100
44/44 [=====] - 6s 142ms/step - loss: 0.5222 - accuracy: 0.8075 - val_loss: 0.939
8 - val_accuracy: 0.6500
Epoch 16/100
44/44 [=====] - 6s 134ms/step - loss: 0.5174 - accuracy: 0.8082 - val_loss: 0.514
5 - val_accuracy: 0.7969
Epoch 17/100
44/44 [=====] - 6s 126ms/step - loss: 0.4715 - accuracy: 0.8362 - val_loss: 0.704
9 - val_accuracy: 0.7312
Epoch 18/100
44/44 [=====] - 6s 137ms/step - loss: 0.4626 - accuracy: 0.8068 - val_loss: 0.620
7 - val_accuracy: 0.7188
Epoch 19/100
44/44 [=====] - 6s 126ms/step - loss: 0.4277 - accuracy: 0.8501 - val_loss: 0.598
2 - val_accuracy: 0.7625
Epoch 20/100
44/44 [=====] - 6s 143ms/step - loss: 0.4899 - accuracy: 0.8253 - val_loss: 0.524
2 - val_accuracy: 0.7625
Epoch 21/100
44/44 [=====] - 6s 129ms/step - loss: 0.4082 - accuracy: 0.8580 - val_loss: 0.519

2 - val_accuracy: 0.7906

Epoch 22/100

44/44 [=====] - 6s 131ms/step - loss: 0.3886 - accuracy: 0.8587 - val_loss: 0.469

3 - val_accuracy: 0.7969

Epoch 23/100

44/44 [=====] - 6s 132ms/step - loss: 0.4036 - accuracy: 0.8578 - val_loss: 0.509

9 - val_accuracy: 0.7969

Epoch 24/100

44/44 [=====] - 6s 132ms/step - loss: 0.3820 - accuracy: 0.8750 - val_loss: 0.615

5 - val_accuracy: 0.7625

Epoch 25/100

44/44 [=====] - 6s 142ms/step - loss: 0.3605 - accuracy: 0.8679 - val_loss: 0.564

1 - val_accuracy: 0.7844

Epoch 26/100

44/44 [=====] - 6s 132ms/step - loss: 0.3695 - accuracy: 0.8628 - val_loss: 0.607

3 - val_accuracy: 0.7563

Epoch 27/100

44/44 [=====] - 6s 132ms/step - loss: 0.4224 - accuracy: 0.8544 - val_loss: 0.571

3 - val_accuracy: 0.7656

Epoch 28/100

44/44 [=====] - 6s 133ms/step - loss: 0.3548 - accuracy: 0.8778 - val_loss: 0.502

1 - val_accuracy: 0.8344

Epoch 00028: ReduceLROnPlateau reducing learning rate to 0.00020000000949949026.

Epoch 29/100

44/44 [=====] - 6s 130ms/step - loss: 0.2946 - accuracy: 0.9027 - val_loss: 0.414

3 - val_accuracy: 0.8250

Epoch 30/100

44/44 [=====] - 6s 128ms/step - loss: 0.2901 - accuracy: 0.8991 - val_loss: 0.461

8 - val_accuracy: 0.8219

Epoch 31/100

44/44 [=====] - 6s 126ms/step - loss: 0.2685 - accuracy: 0.9009 - val_loss: 0.4028 - val_accuracy: 0.8469
Epoch 32/100
44/44 [=====] - 6s 126ms/step - loss: 0.2239 - accuracy: 0.9212 - val_loss: 0.3870 - val_accuracy: 0.8406
Epoch 33/100
44/44 [=====] - 6s 129ms/step - loss: 0.2366 - accuracy: 0.9155 - val_loss: 0.3796 - val_accuracy: 0.8656
Epoch 34/100
44/44 [=====] - 6s 137ms/step - loss: 0.2360 - accuracy: 0.9148 - val_loss: 0.3351 - val_accuracy: 0.8719
Epoch 35/100
44/44 [=====] - 6s 130ms/step - loss: 0.2469 - accuracy: 0.9048 - val_loss: 0.3281 - val_accuracy: 0.8562
Epoch 36/100
44/44 [=====] - 6s 136ms/step - loss: 0.2122 - accuracy: 0.9219 - val_loss: 0.3684 - val_accuracy: 0.8469
Epoch 37/100
44/44 [=====] - 5s 123ms/step - loss: 0.2508 - accuracy: 0.9077 - val_loss: 0.3942 - val_accuracy: 0.8250
Epoch 38/100
44/44 [=====] - 6s 128ms/step - loss: 0.2298 - accuracy: 0.9155 - val_loss: 0.2819 - val_accuracy: 0.8781
Epoch 39/100
44/44 [=====] - 6s 128ms/step - loss: 0.2008 - accuracy: 0.9233 - val_loss: 0.3285 - val_accuracy: 0.8594
Epoch 40/100
44/44 [=====] - 6s 143ms/step - loss: 0.2015 - accuracy: 0.9240 - val_loss: 0.3728 - val_accuracy: 0.8469
Epoch 41/100
44/44 [=====] - 6s 134ms/step - loss: 0.1845 - accuracy: 0.9297 - val_loss: 0.336

5 - val_accuracy: 0.8750

Epoch 42/100

44/44 [=====] - 6s 135ms/step - loss: 0.1929 - accuracy: 0.9247 - val_loss: 0.334

5 - val_accuracy: 0.8687

Epoch 43/100

44/44 [=====] - 6s 127ms/step - loss: 0.1962 - accuracy: 0.9205 - val_loss: 0.260

9 - val_accuracy: 0.8844

Epoch 44/100

44/44 [=====] - 6s 136ms/step - loss: 0.1720 - accuracy: 0.9361 - val_loss: 0.302

0 - val_accuracy: 0.8813

Epoch 45/100

44/44 [=====] - 6s 130ms/step - loss: 0.1562 - accuracy: 0.9283 - val_loss: 0.435

9 - val_accuracy: 0.8562

Epoch 46/100

44/44 [=====] - 6s 139ms/step - loss: 0.1672 - accuracy: 0.9375 - val_loss: 0.355

3 - val_accuracy: 0.8656

Epoch 47/100

44/44 [=====] - 6s 138ms/step - loss: 0.1644 - accuracy: 0.9339 - val_loss: 0.325

3 - val_accuracy: 0.8938

Epoch 48/100

44/44 [=====] - 6s 139ms/step - loss: 0.1664 - accuracy: 0.9425 - val_loss: 0.341

3 - val_accuracy: 0.8781

Epoch 49/100

44/44 [=====] - 6s 126ms/step - loss: 0.1648 - accuracy: 0.9347 - val_loss: 0.317

7 - val_accuracy: 0.8781

Epoch 00049: ReduceLROnPlateau reducing learning rate to 4.0000001899898055e-05.

Epoch 50/100

44/44 [=====] - 6s 128ms/step - loss: 0.1382 - accuracy: 0.9483 - val_loss: 0.288

3 - val_accuracy: 0.9000

Epoch 51/100

44/44 [=====] - 6s 129ms/step - loss: 0.1304 - accuracy: 0.9567 - val_loss: 0.318
5 - val_accuracy: 0.8906
Epoch 52/100
44/44 [=====] - 6s 137ms/step - loss: 0.1150 - accuracy: 0.9602 - val_loss: 0.307
2 - val_accuracy: 0.9000
Epoch 53/100
44/44 [=====] - 6s 128ms/step - loss: 0.1253 - accuracy: 0.9538 - val_loss: 0.244
7 - val_accuracy: 0.9031
Epoch 54/100
44/44 [=====] - 6s 141ms/step - loss: 0.1199 - accuracy: 0.9609 - val_loss: 0.290
5 - val_accuracy: 0.9062
Epoch 55/100
44/44 [=====] - 5s 124ms/step - loss: 0.1230 - accuracy: 0.9553 - val_loss: 0.245
0 - val_accuracy: 0.9156
Epoch 56/100
44/44 [=====] - 6s 140ms/step - loss: 0.1355 - accuracy: 0.9517 - val_loss: 0.250
0 - val_accuracy: 0.8969
Epoch 57/100
44/44 [=====] - 6s 130ms/step - loss: 0.1043 - accuracy: 0.9602 - val_loss: 0.346
1 - val_accuracy: 0.8531
Epoch 58/100
44/44 [=====] - 6s 139ms/step - loss: 0.1266 - accuracy: 0.9517 - val_loss: 0.238
5 - val_accuracy: 0.9094
Epoch 59/100
44/44 [=====] - 6s 126ms/step - loss: 0.1091 - accuracy: 0.9659 - val_loss: 0.284
5 - val_accuracy: 0.8969
Epoch 60/100
44/44 [=====] - 6s 136ms/step - loss: 0.1179 - accuracy: 0.9581 - val_loss: 0.267
9 - val_accuracy: 0.9187
Epoch 61/100
44/44 [=====] - 5s 123ms/step - loss: 0.1010 - accuracy: 0.9680 - val_loss: 0.242

0 - val_accuracy: 0.9094

Epoch 62/100

44/44 [=====] - 6s 140ms/step - loss: 0.1029 - accuracy: 0.9655 - val_loss: 0.339

6 - val_accuracy: 0.8781

Epoch 63/100

44/44 [=====] - 6s 131ms/step - loss: 0.0944 - accuracy: 0.9680 - val_loss: 0.265

2 - val_accuracy: 0.8969

Epoch 64/100

44/44 [=====] - 6s 130ms/step - loss: 0.0871 - accuracy: 0.9705 - val_loss: 0.291

1 - val_accuracy: 0.9219

Epoch 00064: ReduceLROnPlateau reducing learning rate to 8.000000525498762e-06.

Epoch 65/100

44/44 [=====] - 6s 132ms/step - loss: 0.0915 - accuracy: 0.9716 - val_loss: 0.263

4 - val_accuracy: 0.9062

Epoch 66/100

44/44 [=====] - 6s 135ms/step - loss: 0.1110 - accuracy: 0.9609 - val_loss: 0.274

8 - val_accuracy: 0.9062

Epoch 67/100

44/44 [=====] - 7s 148ms/step - loss: 0.1016 - accuracy: 0.9666 - val_loss: 0.254

1 - val_accuracy: 0.9062

Epoch 68/100

44/44 [=====] - 6s 127ms/step - loss: 0.1099 - accuracy: 0.9638 - val_loss: 0.271

7 - val_accuracy: 0.9375

Epoch 69/100

44/44 [=====] - 6s 126ms/step - loss: 0.1095 - accuracy: 0.9631 - val_loss: 0.285

9 - val_accuracy: 0.8938

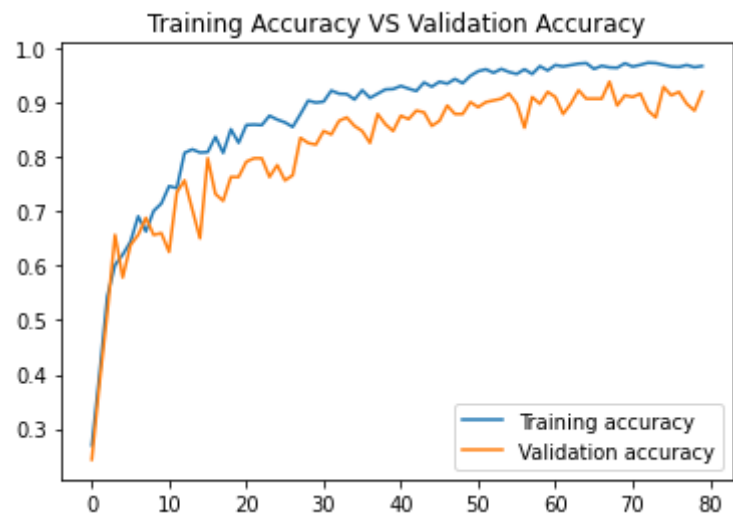
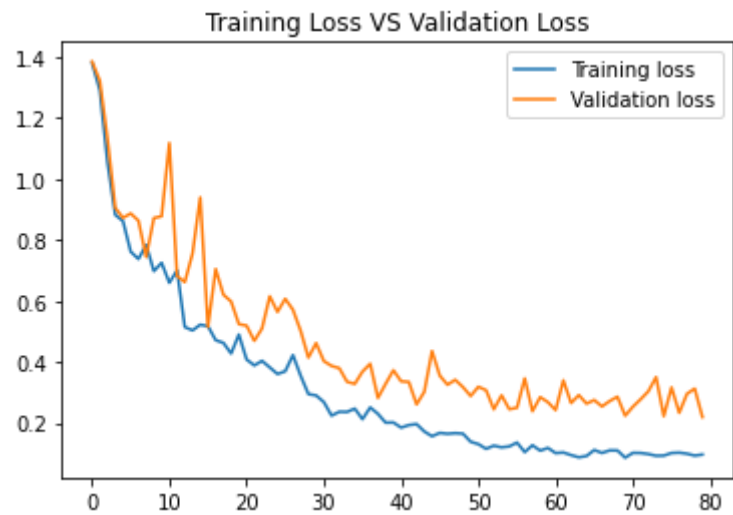
Epoch 70/100

44/44 [=====] - 6s 135ms/step - loss: 0.0859 - accuracy: 0.9713 - val_loss: 0.223

6 - val_accuracy: 0.9125

Epoch 71/100

44/44 [=====] - 6s 126ms/step - loss: 0.1017 - accuracy: 0.9652 - val_loss: 0.253
1 - val_accuracy: 0.9094
Epoch 72/100
44/44 [=====] - 6s 127ms/step - loss: 0.1013 - accuracy: 0.9684 - val_loss: 0.277
4 - val_accuracy: 0.9156
Epoch 73/100
44/44 [=====] - 6s 140ms/step - loss: 0.0978 - accuracy: 0.9723 - val_loss: 0.303
7 - val_accuracy: 0.8844
Epoch 74/100
44/44 [=====] - 6s 132ms/step - loss: 0.0922 - accuracy: 0.9716 - val_loss: 0.350
2 - val_accuracy: 0.8719
Epoch 75/100
44/44 [=====] - 6s 145ms/step - loss: 0.0927 - accuracy: 0.9684 - val_loss: 0.222
3 - val_accuracy: 0.9281
Epoch 76/100
44/44 [=====] - 6s 129ms/step - loss: 0.1012 - accuracy: 0.9652 - val_loss: 0.316
4 - val_accuracy: 0.9125
Epoch 77/100
44/44 [=====] - 6s 132ms/step - loss: 0.1024 - accuracy: 0.9645 - val_loss: 0.233
2 - val_accuracy: 0.9187
Epoch 78/100
44/44 [=====] - 6s 131ms/step - loss: 0.0990 - accuracy: 0.9680 - val_loss: 0.295
4 - val_accuracy: 0.8969
Epoch 79/100
44/44 [=====] - 5s 124ms/step - loss: 0.0927 - accuracy: 0.9641 - val_loss: 0.312
5 - val_accuracy: 0.8844
Epoch 80/100
44/44 [=====] - 6s 127ms/step - loss: 0.0962 - accuracy: 0.9662 - val_loss: 0.219
2 - val_accuracy: 0.9187



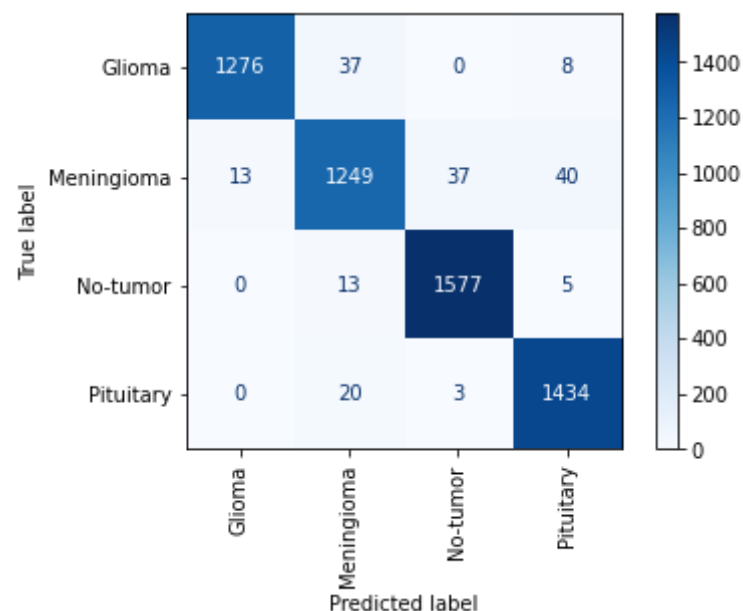
Model saved successfully!
Confusion matrix for train data:
Found 5712 images belonging to 4 classes.

```
2021-11-12 09:00:18.579558: W tensorflow/core/grappler/costs/op_level_cost_estimator.cc:689] Error in PredictCost() for the op: op: "Softmax" attr { key: "T" value { type: DT_FLOAT } } inputs { dtype: DT_FLOAT shape { unknown_rank: true } } device { type: "GPU" vendor: "NVIDIA" model: "Tesla P100-PCIE-16GB" frequency: 1328 num_cores: 56 environment { key: "architecture" value: "6.0" } environment { key: "cuda" value: "11000" } environment { key: "cudnn" value: "8005" } num_registers: 65536 l1_cache_size: 24576 l2_cache_size: 4194304 shared_memory_size_per_multiprocessor: 65536 memory_size: 16152002560 bandwidth: 732160000 } outputs { dtype: DT_FLOAT shape { unknown_rank: true } }
```

179/179 [=====] - 19s 103ms/step - loss: 0.0927 - accuracy: 0.9692

Score = [0.09273112565279007, 0.9691876769065857]

Accuracy = 0.969187675070028



Confusion matrix for val data:

Found 1311 images belonging to 4 classes.

41/41 [=====] - 4s 94ms/step - loss: 0.2816 - accuracy: 0.9054

Score = [0.2815778851509094, 0.9054157137870789]

Accuracy = 0.9054157131960335

