

```
In [2]: import pandas as pd
from tensorflow.keras.preprocessing.image import ImageDataGenerator
import os
import numpy as np
from PIL import Image
import os
from IPython.display import display
import zipfile
import matplotlib.pyplot as plt
from keras.preprocessing.image import ImageDataGenerator
from keras.models import Sequential
from keras.layers import Conv2D, MaxPooling2D, Dense, Flatten, Dropout
```

```
In [3]: from google.colab import drive

drive.mount('/content/drive')

Mounted at /content/drive
```

```
In [ ]: # Loading data and Image Augmentation
source_zip_path = '/content/archive (2).zip'

destination_folder_path = '/content/drive/MyDrive/<drive>'

!mv "/content/archive (2).zip" "/content/drive/MyDrive"
```

```
In [ ]: # Loading data and Image Augmentation
zip_file_path = '/content/drive/MyDrive/<drive>/archive (2).zip'

destination_folder_path = '/content/drive/MyDrive/<drive>/extracted_folder'

with zipfile.ZipFile('/content/drive/MyDrive/archive (2).zip', 'r') as zip_ref:
    zip_ref.extractall(destination_folder_path)
```

```
In [ ]: # Loading data and Image Augmentation
image_folder_path = '/content/drive/MyDrive/<drive>/extracted_folder'

image_files = [file for file in os.listdir('/content/drive/MyDrive/<drive>/extracted_folder/Training') if file.]

for file in image_files:
    file_path = os.path.join('/content/drive/MyDrive/<drive>/extracted_folder/Training', file)

    image = Image.open(file_path)
    resized_image = image.resize((256, 256))
    resized_image.show()
```

```
In [ ]: # Loading data and Image Augmentation
image_folder_path = '/content/drive/MyDrive/<drive>/extracted_folder'

image_files = [file for file in os.listdir(image_folder_path) if file.endswith('.jpg') or file.endswith('.png')]

for file in image_files:
    file_path = os.path.join(image_folder_path, file)
    image = Image.open(file_path)
    display(image)
```

```
In [4]: # Loading data and Image Augmentation
train_datagen = ImageDataGenerator(rescale=1.0/255)
train_generator = train_datagen.flow_from_directory('/content/drive/MyDrive/<drive>/extracted_folder/Training',
                                                    batch_size=30,
                                                    target_size=(150, 150))

validation_datagen = ImageDataGenerator(rescale=1.0/255)
validation_generator = validation_datagen.flow_from_directory('/content/drive/MyDrive/<drive>/extracted_folder/',
                                                            batch_size=30,
                                                            target_size=(150, 150))

Found 5712 images belonging to 4 classes.
Found 1311 images belonging to 4 classes.
```

```
In [5]: labels_2=['notumor', 'pituitary', 'glioma', 'meningioma']
```

```
In [6]: # Display the count of each unique classes
labels = validation_generator.labels
labels_series = pd.Series(labels)

value_counts = labels_series.value_counts()

print(value_counts)
```

```
2    405
1    306
0    300
3    300
dtype: int64
```

```
In [7]: # Display the count of each unique classes
labels = train_generator.labels
labels_series = pd.Series(labels)

value_counts = labels_series.value_counts()

print(value_counts)
```

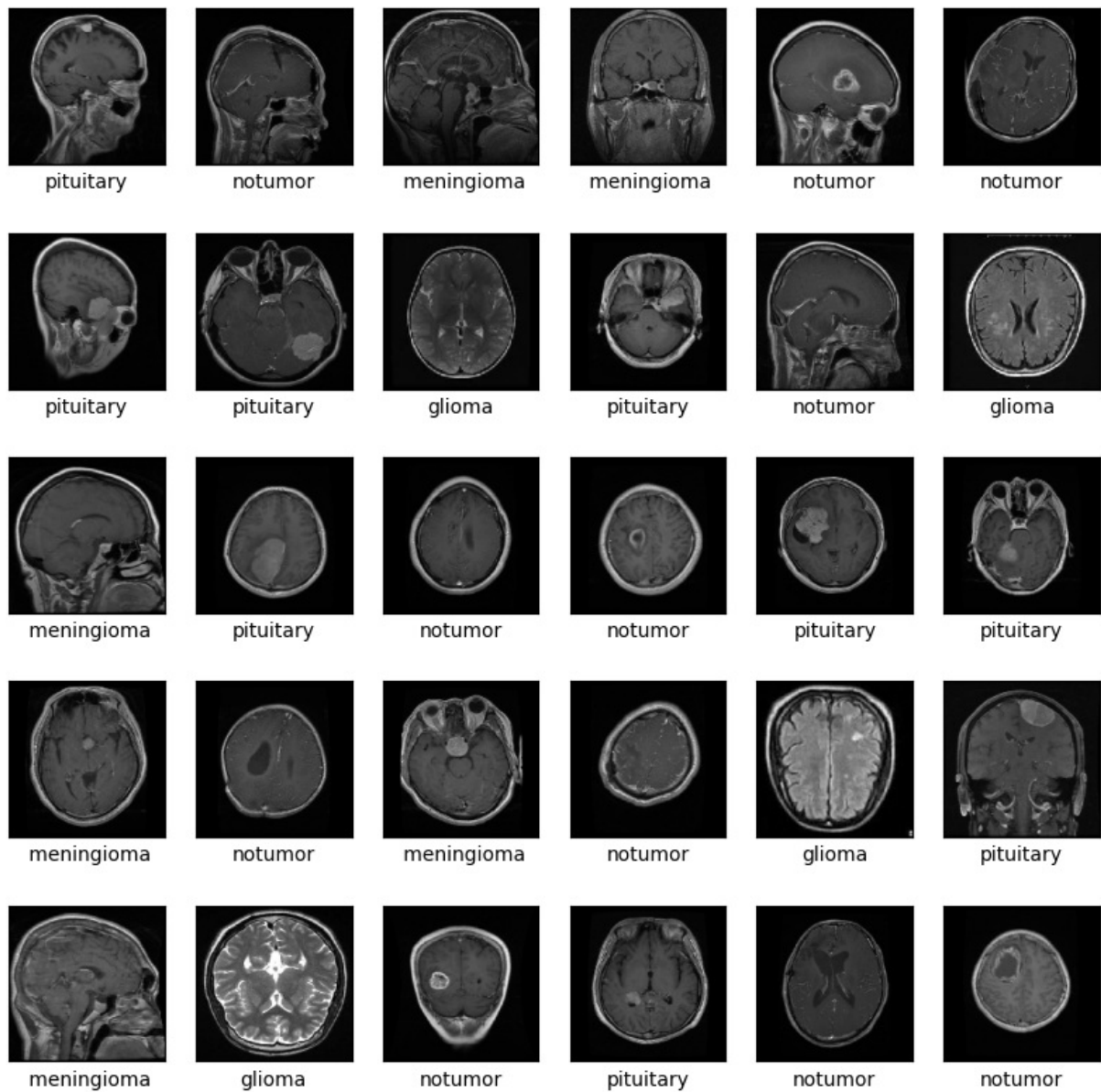
```
2    1595
3    1457
1    1339
0    1321
dtype: int64
```

```
In [8]: # Display some images
i = 0

batch_size = validation_generator.batch_size
if i < len(validation_generator):
    batch_images, batch_labels = validation_generator[i]
    plt.figure(figsize=(10,10))
    for idx in range(batch_size):
        plt.subplot(5, 6, idx + 1)
        plt.xticks([])
        plt.yticks([])
        plt.grid(False)

        plt.imshow(batch_images[idx])
        class_index = np.argmax(batch_labels[idx])
        class_name = labels_2[class_index]
        plt.xlabel(class_name)

        if idx >= batch_size - 1:
            break
    else:
        print("Invalid batch index!")
```



```
In [9]: # Building model
model=Sequential()

model.add(Conv2D(filters=50,
                  kernel_size=5,
                  strides=1,
                  activation='relu',
                  input_shape=(150,150,3)))

model.add(MaxPooling2D(pool_size=2,strides=2))

model.add(Conv2D(filters=100,
                  kernel_size=5,
                  strides=1,
                  activation='relu'))

model.add(MaxPooling2D(pool_size=2,strides=2))

model.add(Flatten())
model.add(Dropout(0.5))

model.add(Dense(150, activation='relu'))
model.add(Dense(4,activation='softmax'))

# Compiling the model
model.compile(optimizer='adam',loss='categorical_crossentropy',metrics=['accuracy'])

In [10]: # Model Summary
model.summary()
```

Model: "sequential"

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 146, 146, 50)	3800
max_pooling2d (MaxPooling2D)	(None, 73, 73, 50)	0
conv2d_1 (Conv2D)	(None, 69, 69, 100)	125100
max_pooling2d_1 (MaxPooling2D)	(None, 34, 34, 100)	0
flatten (Flatten)	(None, 115600)	0
dropout (Dropout)	(None, 115600)	0
dense (Dense)	(None, 150)	17340150
dense_1 (Dense)	(None, 4)	604

=====  
Total params: 17469654 (66.64 MB)  
Trainable params: 17469654 (66.64 MB)  
Non-trainable params: 0 (0.00 Byte)

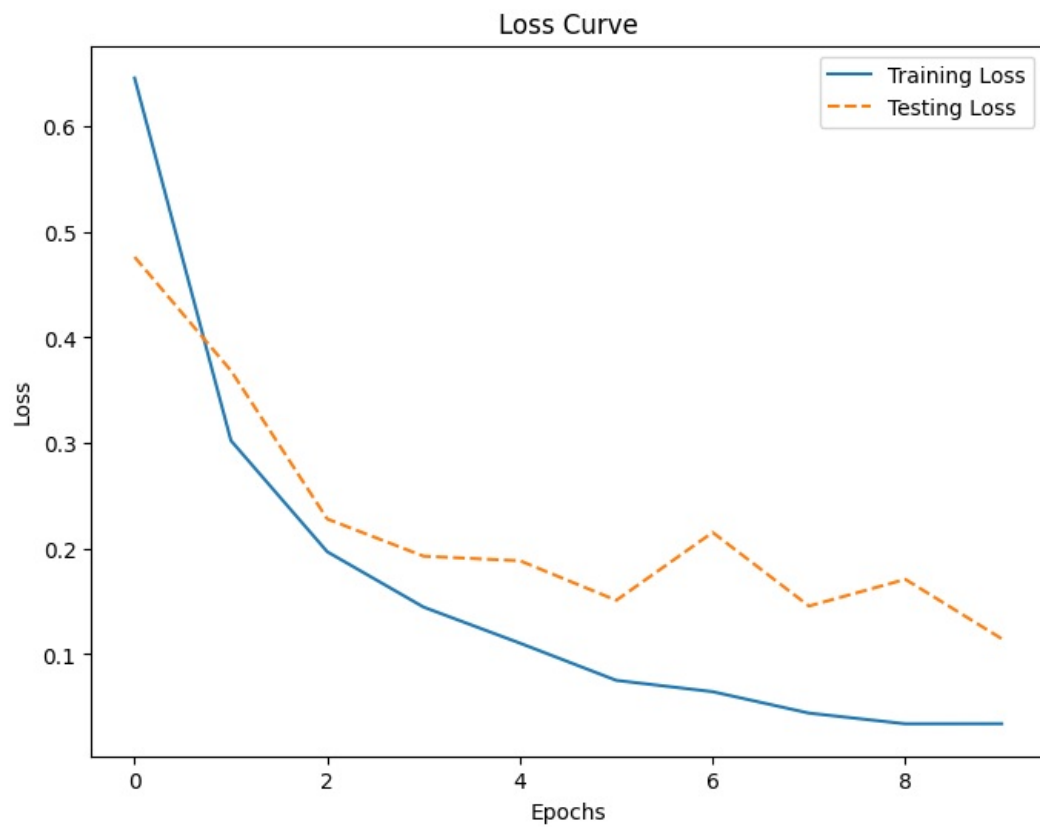
```
In [11]: # Training Model
history=model.fit(train_generator, epochs=10,validation_data=validation_generator)

Epoch 1/10
191/191 [=====] - 1476s 8s/step - loss: 0.6457 - accuracy: 0.7468 - val_loss: 0.4762 -
val_accuracy: 0.8009
Epoch 2/10
191/191 [=====] - 24s 126ms/step - loss: 0.3023 - accuracy: 0.8922 - val_loss: 0.3686
- val_accuracy: 0.8391
Epoch 3/10
191/191 [=====] - 24s 124ms/step - loss: 0.1970 - accuracy: 0.9261 - val_loss: 0.2281
- val_accuracy: 0.9100
Epoch 4/10
191/191 [=====] - 22s 116ms/step - loss: 0.1446 - accuracy: 0.9503 - val_loss: 0.1929
- val_accuracy: 0.9283
Epoch 5/10
191/191 [=====] - 24s 123ms/step - loss: 0.1105 - accuracy: 0.9597 - val_loss: 0.1885
- val_accuracy: 0.9367
Epoch 6/10
191/191 [=====] - 23s 118ms/step - loss: 0.0753 - accuracy: 0.9718 - val_loss: 0.1511
- val_accuracy: 0.9466
Epoch 7/10
191/191 [=====] - 23s 122ms/step - loss: 0.0645 - accuracy: 0.9772 - val_loss: 0.2154
- val_accuracy: 0.9359
Epoch 8/10
191/191 [=====] - 22s 116ms/step - loss: 0.0443 - accuracy: 0.9860 - val_loss: 0.1457
- val_accuracy: 0.9565
Epoch 9/10
191/191 [=====] - 23s 122ms/step - loss: 0.0341 - accuracy: 0.9879 - val_loss: 0.1710
- val_accuracy: 0.9489
Epoch 10/10
191/191 [=====] - 23s 118ms/step - loss: 0.0342 - accuracy: 0.9898 - val_loss: 0.1147
- val_accuracy: 0.9680
```

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In [12]: # Loss Curves

plt.figure(figsize=(8,6))
plt.plot(history.history['loss'])
plt.plot(history.history['val_loss'],ls='--')
plt.legend(['Training Loss','Testing Loss'])
plt.xlabel('Epochs')
plt.ylabel('Loss')
plt.title("Loss Curve")

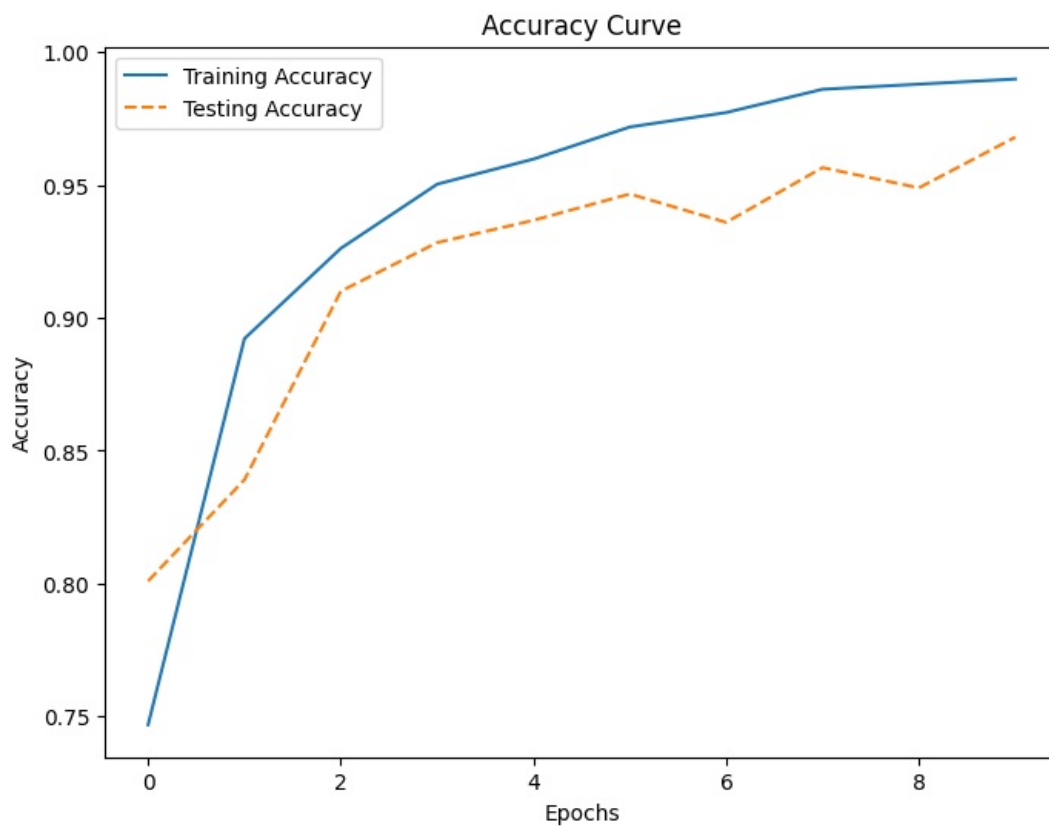
plt.show()
```



```
In [13]: # Accuracy Curves

plt.figure(figsize=(8,6))
plt.plot(history.history['accuracy'])
plt.plot(history.history['val_accuracy'],ls='--')
plt.legend(['Training Accuracy','Testing Accuracy'])
plt.xlabel('Epochs')
plt.ylabel('Accuracy')
plt.title("Accuracy Curve")

plt.show()
```



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In [14]: # Model Evaluation
model.evaluate(validation_generator)

44/44 [=====] - 4s 85ms/step - loss: 0.1147 - accuracy: 0.9680
Out[14]: [0.11472178995609283, 0.9679633975028992]
```

```

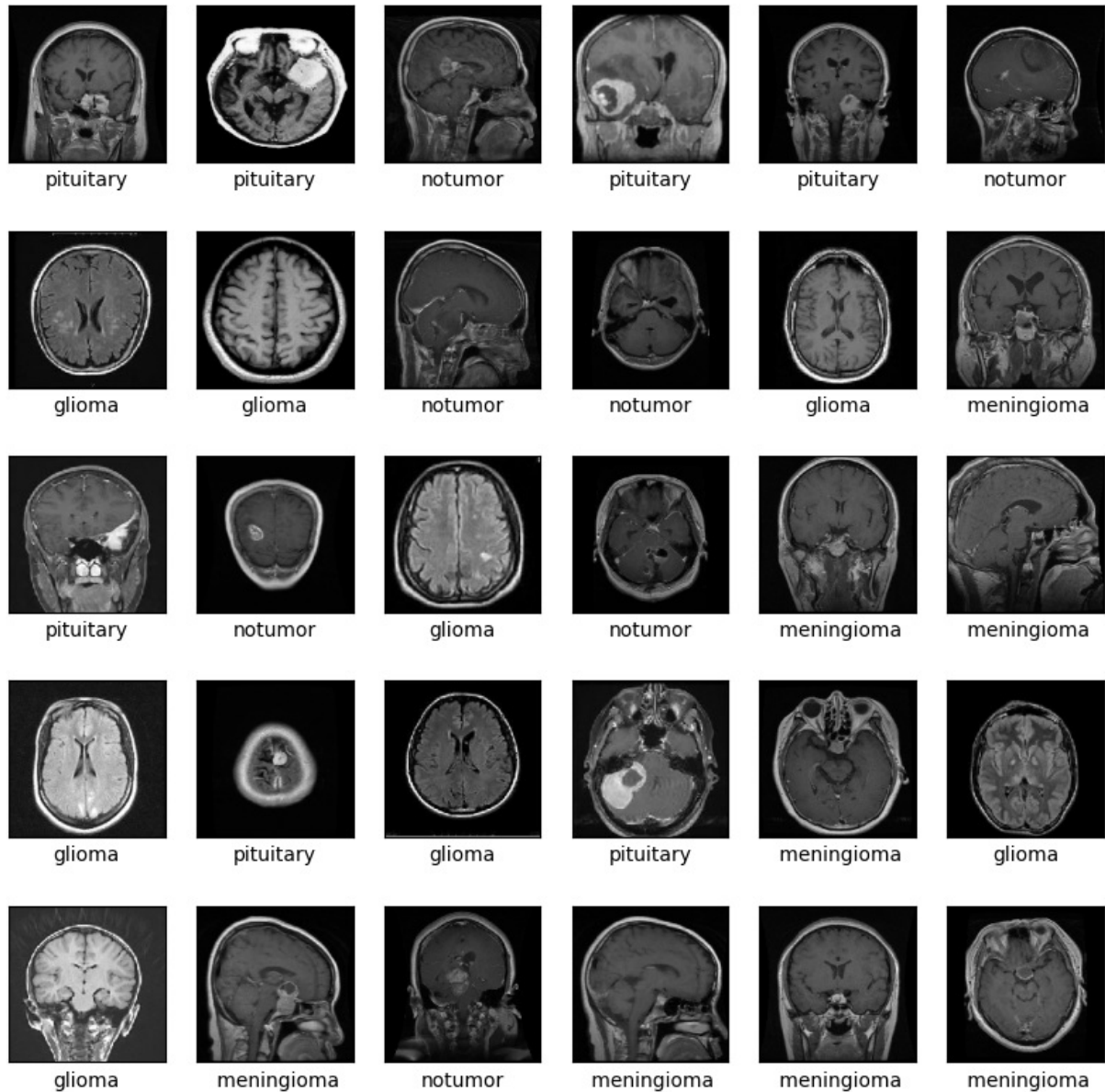
In [15]: # predict some records
i = 0

batch_size = validation_generator.batch_size

if i < len(validation_generator):
    batch_images, batch_labels = validation_generator[i]
    plt.figure(figsize=(10,10))
    for idx in range(batch_size):
        plt.subplot(5, 6, idx + 1)
        plt.xticks([])
        plt.yticks([])
        plt.grid(False)
        plt.imshow(batch_images[idx])
        class_index = np.argmax(batch_labels[idx])
        class_name = labels_2[class_index]
        plt.xlabel(class_name)

        if idx >= batch_size - 1:
            break
    else:
        print("Invalid batch index!")

```



In [ ]:

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