

```
In [14]: # import kagglehub  
  
# # Download latest version  
# path = kagglehub.dataset_download("ishans24/brain-tumor-dataset")  
  
# print("Path to dataset files:", path)
```

```
In [15]: # !pip install kagglehub
```

```
In [9]: import tensorflow as tf  
from tensorflow.keras.models import Sequential  
from tensorflow.keras.layers import Conv2D, MaxPooling2D, Flatten, Dense  
from tensorflow.keras.preprocessing.image import ImageDataGenerator  
import matplotlib.pyplot as plt
```

```
In [10]: import os  
  
dataset_path = r"C:\Users\Piyush\Downloads\brain_tumor_dataset"  
  
print("Classes:", os.listdir(dataset_path))  
  
print("No tumor images count:",  
      len(os.listdir(os.path.join(dataset_path, "no"))))  
  
print("Yes tumor images count:",  
      len(os.listdir(os.path.join(dataset_path, "yes"))))
```

```
Classes: ['no', 'yes']  
No tumor images count: 98  
Yes tumor images count: 155
```

```
In [ ]:
```

```
In [11]: from tensorflow.keras.preprocessing.image import ImageDataGenerator  
  
img_size = 224  
batch_size = 16  
  
datagen = ImageDataGenerator(
```

```
        rescale=1./255,
        validation_split=0.2
    )

    train_data = datagen.flow_from_directory(
        r"C:\Users\Piyush\Downloads\brain_tumor_dataset",
        target_size=(img_size, img_size),
        batch_size=batch_size,
        class_mode='binary',
        subset='training'
    )

    val_data = datagen.flow_from_directory(
        r"C:\Users\Piyush\Downloads\brain_tumor_dataset",
        target_size=(img_size, img_size),
        batch_size=batch_size,
        class_mode='binary',
        subset='validation'
    )
```

Found 203 images belonging to 2 classes.

Found 50 images belonging to 2 classes.

In [ ]:

```
In [12]: model = Sequential([
    Conv2D(32, (3,3), activation='relu', input_shape=(img_size, img_size, 3)),
    MaxPooling2D(),

    Conv2D(64, (3,3), activation='relu'),
    MaxPooling2D(),

    Flatten(),
    Dense(128, activation='relu'),
    Dense(1, activation='sigmoid')
])
```

In [ ]:

```
In [13]: model.compile(
    optimizer='adam',
```

```
        loss='binary_crossentropy',
        metrics=['accuracy']
    )
```

In [ ]:

```
In [ ]: history = model.fit(
    train_data,
    validation_data=val_data,
    epochs=10
)
```

C:\Users\Piyush\anaconda3\Lib\site-packages\keras\src\trainers\data\_adapters\py\_dataset\_adapter.py:121: UserWarning:  
Your `PyDataset` class should call `super().\_\_init\_\_(\*\*kwargs)` in its constructor. `\*\*kwargs` can include `workers`,  
`use\_multiprocessing`, `max\_queue\_size`. Do not pass these arguments to `fit()`, as they will be ignored.  
self.\_warn\_if\_super\_not\_called()

Epoch 1/10

13/13 ━━━━━━ 0s 450ms/step - accuracy: 0.6776 - loss: 1.0751

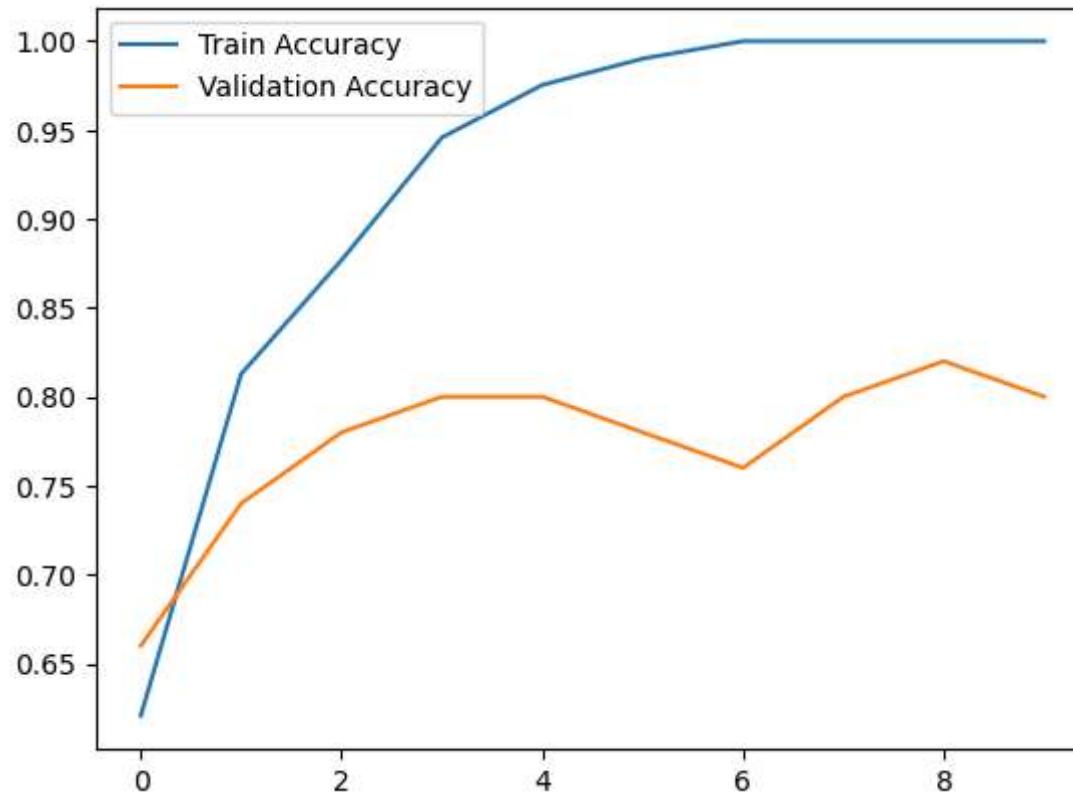
In [ ]:

```
In [11]: from sklearn.metrics import classification_report
print(classification_report(y_true, y_pred, target_names=["No Tumor", "Tumor"]))
```

	precision	recall	f1-score	support
No Tumor	0.46	0.32	0.38	19
Tumor	0.65	0.77	0.71	31
accuracy			0.60	50
macro avg	0.56	0.54	0.54	50
weighted avg	0.58	0.60	0.58	50

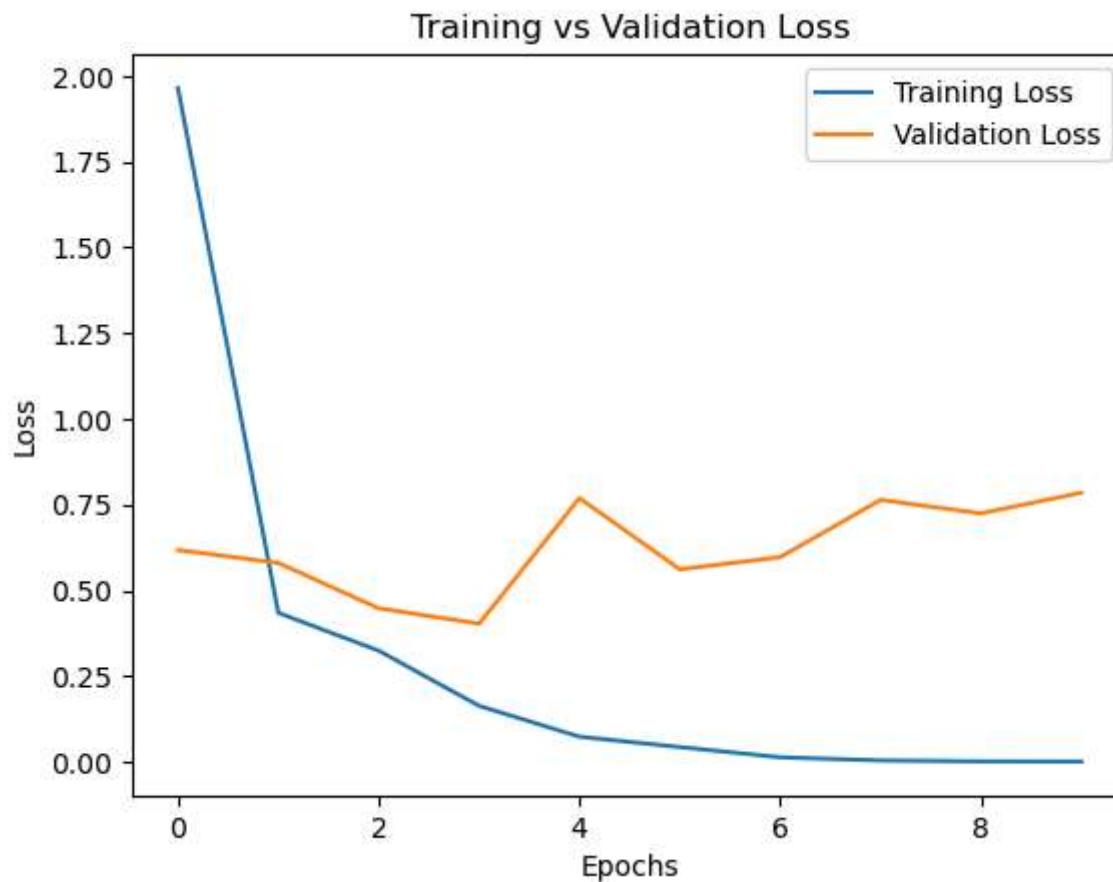
In [ ]:

```
In [7]: plt.plot(history.history['accuracy'], label='Train Accuracy')
plt.plot(history.history['val_accuracy'], label='Validation Accuracy')
plt.legend()
plt.show()
```



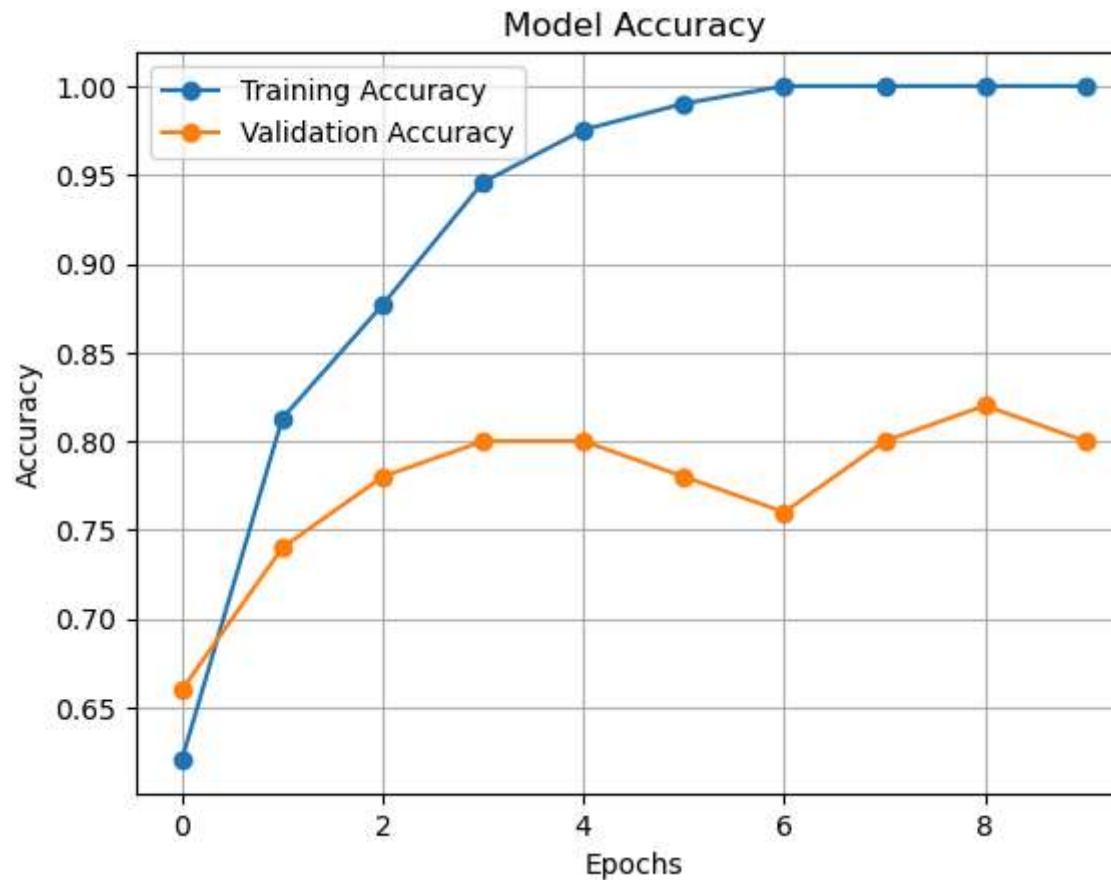
In [ ]:

```
In [8]: plt.figure()
plt.plot(history.history['loss'], label='Training Loss')
plt.plot(history.history['val_loss'], label='Validation Loss')
plt.xlabel('Epochs')
plt.ylabel('Loss')
plt.title('Training vs Validation Loss')
plt.legend()
plt.show()
```



In [ ]:

```
In [9]: plt.figure()
plt.plot(history.history['accuracy'], marker='o', label='Training Accuracy')
plt.plot(history.history['val_accuracy'], marker='o', label='Validation Accuracy')
plt.xlabel('Epochs')
plt.ylabel('Accuracy')
plt.title('Model Accuracy')
plt.legend()
plt.grid()
plt.show()
```



In [ ]:

```
In [13]: from sklearn.metrics import roc_curve, auc

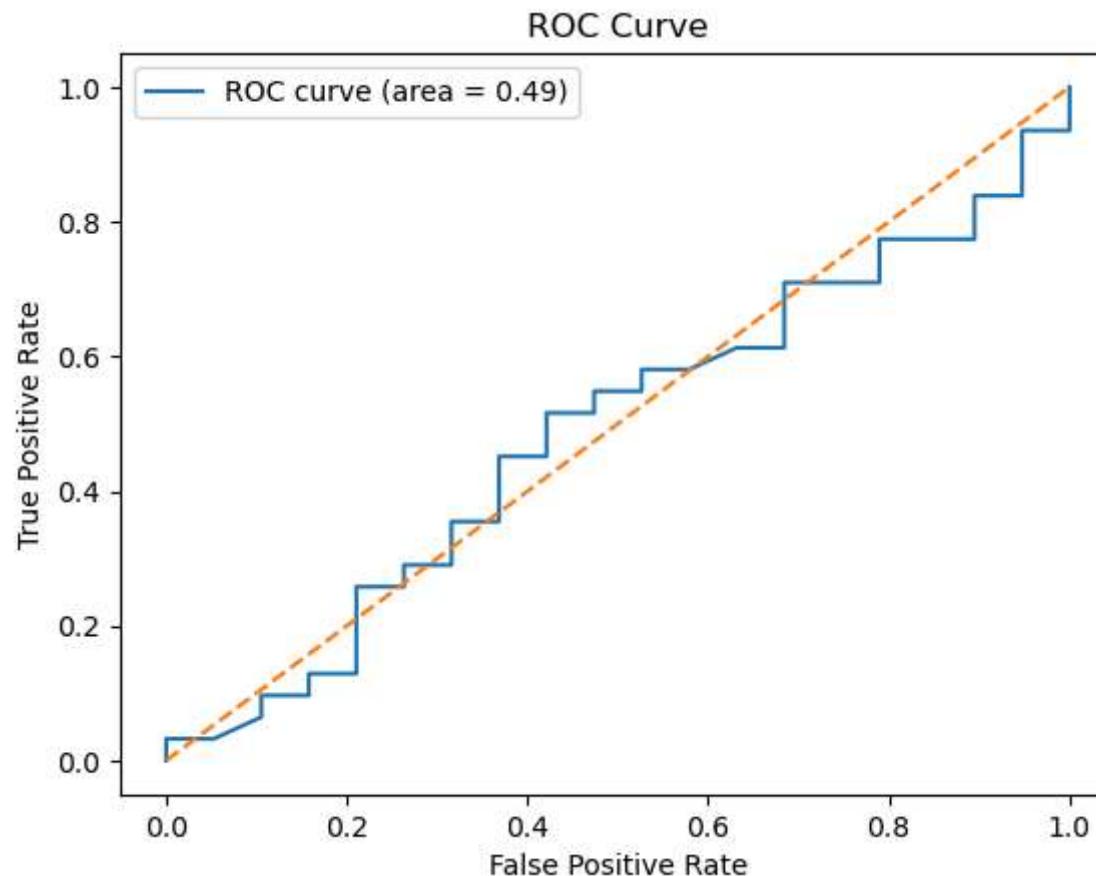
y_prob = model.predict(val_data).ravel()

fpr, tpr, _ = roc_curve(y_true, y_prob)
roc_auc = auc(fpr, tpr)

plt.figure()
plt.plot(fpr, tpr, label='ROC curve (area = %0.2f)' % roc_auc)
plt.plot([0, 1], [0, 1], linestyle='--')
plt.xlabel('False Positive Rate')
plt.ylabel('True Positive Rate')
```

```
plt.title('ROC Curve')
plt.legend()
plt.show()
```

4/4 ————— 0s 84ms/step



In [ ]:

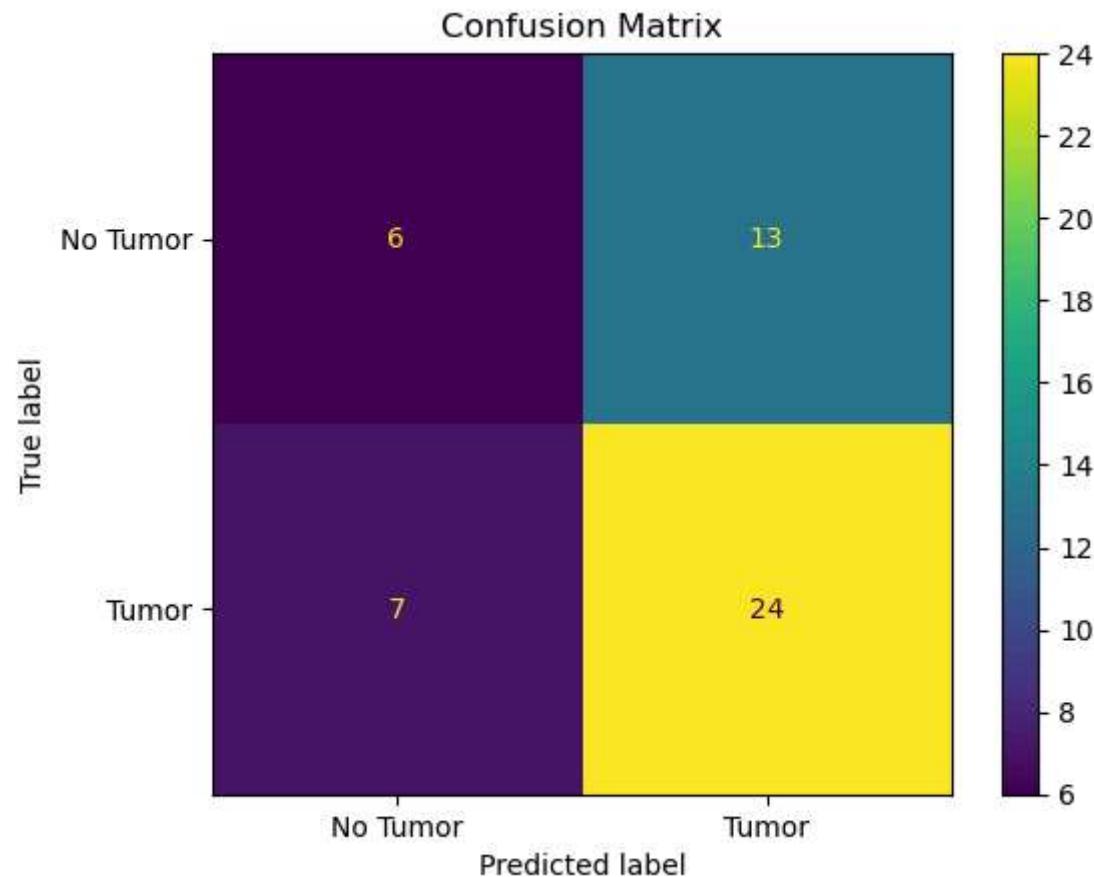
```
In [10]: from sklearn.metrics import confusion_matrix, ConfusionMatrixDisplay
import numpy as np

y_true = val_data.classes
y_pred = (model.predict(val_data) > 0.5).astype("int32").flatten()

cm = confusion_matrix(y_true, y_pred)
```

```
disp = ConfusionMatrixDisplay(confusion_matrix=cm,
                               display_labels=["No Tumor", "Tumor"])
disp.plot()
plt.title("Confusion Matrix")
plt.show()
```

4/4 ━━━━━━━━ 1s 107ms/step



In [ ]:

In [12]:  
import random  
import cv2

```
class_names = ["No Tumor", "Tumor"]
```

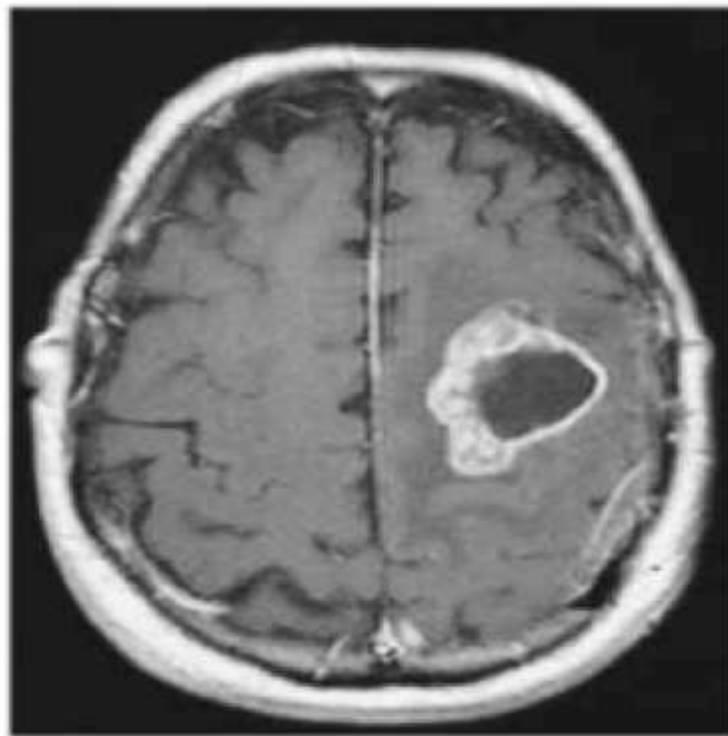
```
for i in range(5):
    img_path, label = random.choice(val_data.filepaths), None
    img = cv2.imread(img_path)
    img = cv2.resize(img, (224,224))
    img_norm = img / 255.0
    img_norm = img_norm.reshape(1,224,224,3)

    pred = model.predict(img_norm)[0][0]
    predicted_label = class_names[int(pred > 0.5)]

    plt.imshow(cv2.cvtColor(img, cv2.COLOR_BGR2RGB))
    plt.title(f"Prediction: {predicted_label}")
    plt.axis("off")
    plt.show()
```

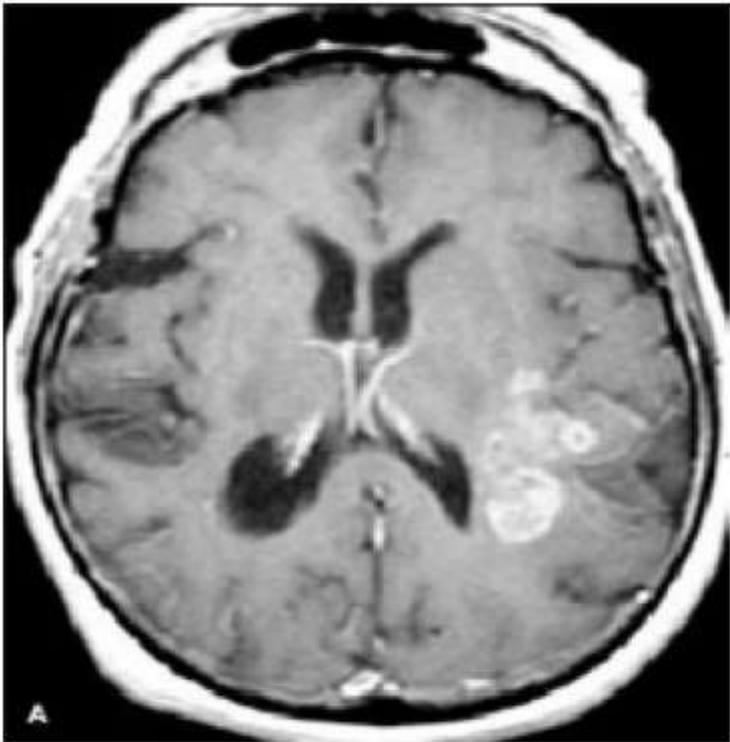
1/1 ————— 0s 174ms/step

Prediction: Tumor



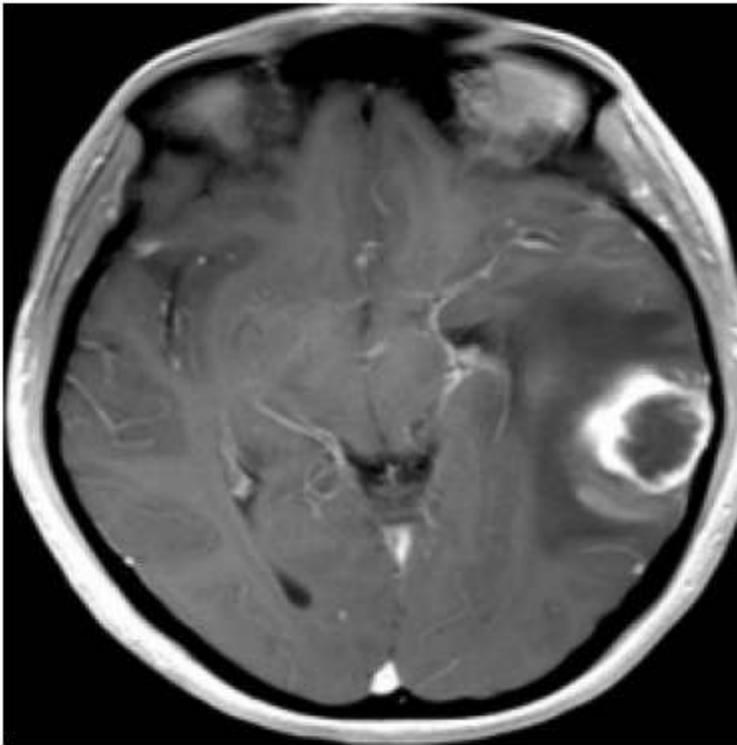
1/1 ————— 0s 82ms/step

Prediction: Tumor



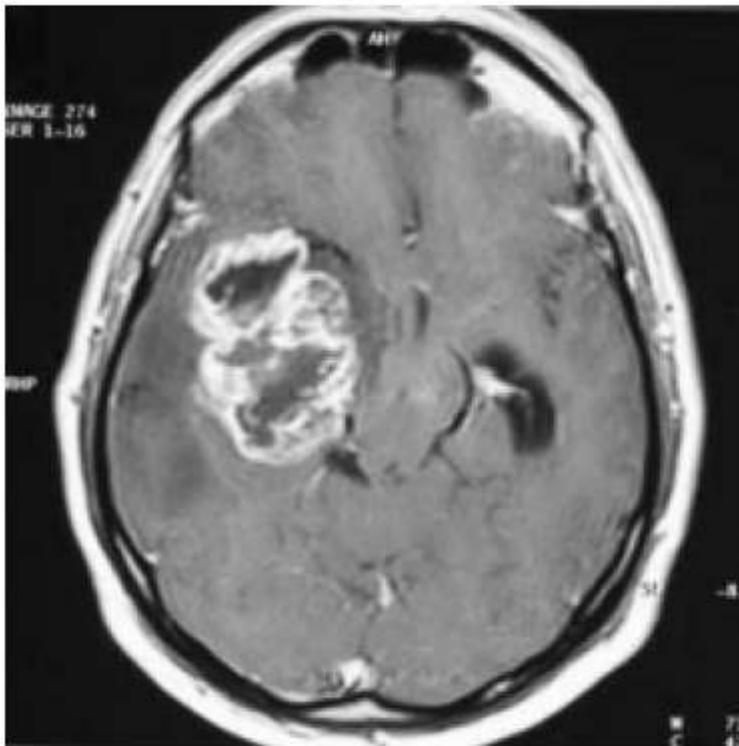
1/1 ————— 0s 87ms/step

Prediction: Tumor



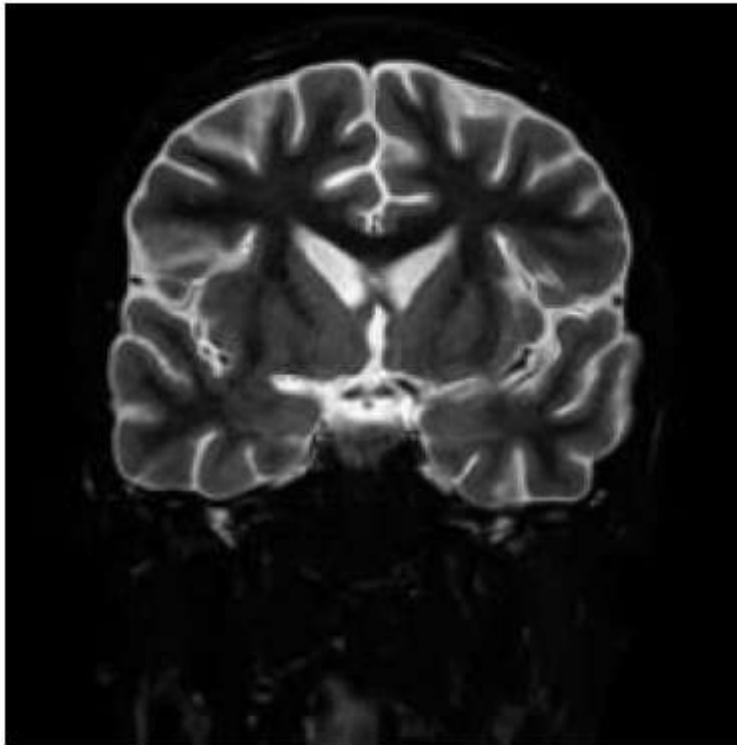
1/1 ————— 0s 83ms/step

Prediction: Tumor



1/1 ————— 0s 80ms/step

Prediction: No Tumor



In [ ]:

In [17]: `model.save("brain_tumor_model.keras")`

In [ ]:

In [ ]: