

```
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 [ ]:
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```
In [11]: from tensorflow.keras.preprocessing.image import ImageDataGenerator

img_size = 224
batch_size = 16

datagen = ImageDataGenerator(
```

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    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',
```

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

```

In []:

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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 []:

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

```

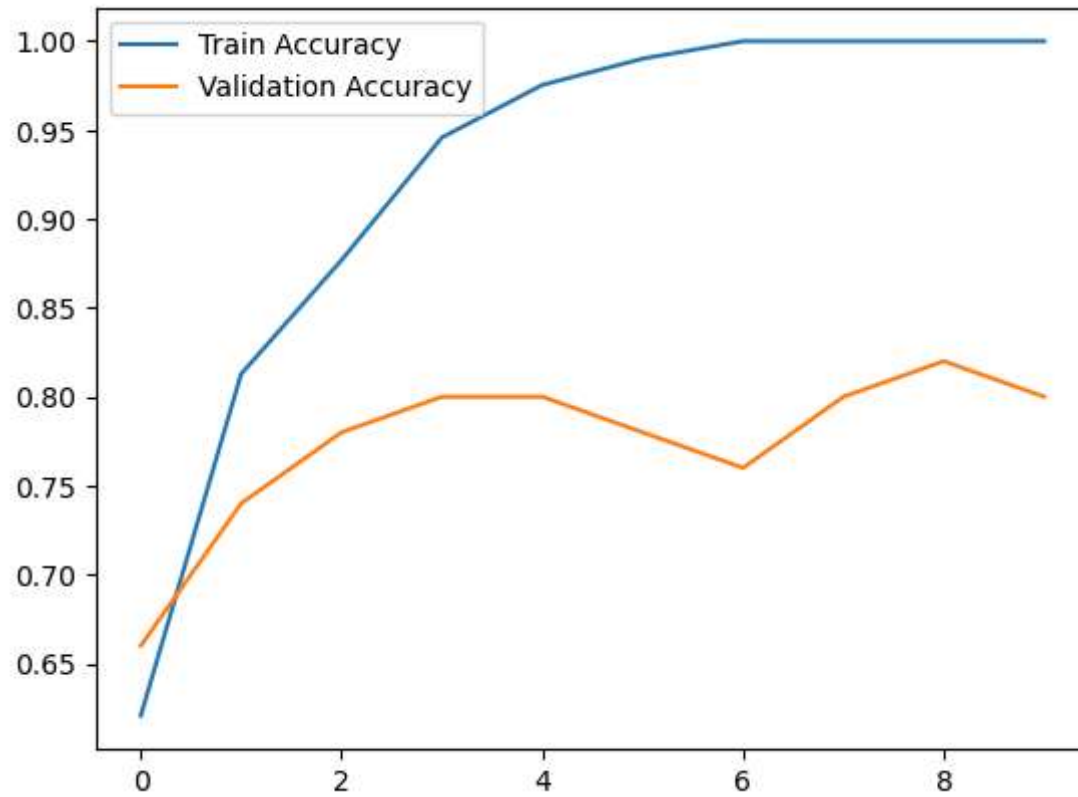
	precision	recall	f1-score	support
No Tumor	0.46	0.32	0.38	19
Tumor	0.65	0.77	0.71	31
accuracy			0.60	50
macro avg	0.56	0.54	0.54	50
weighted avg	0.58	0.60	0.58	50

In []:

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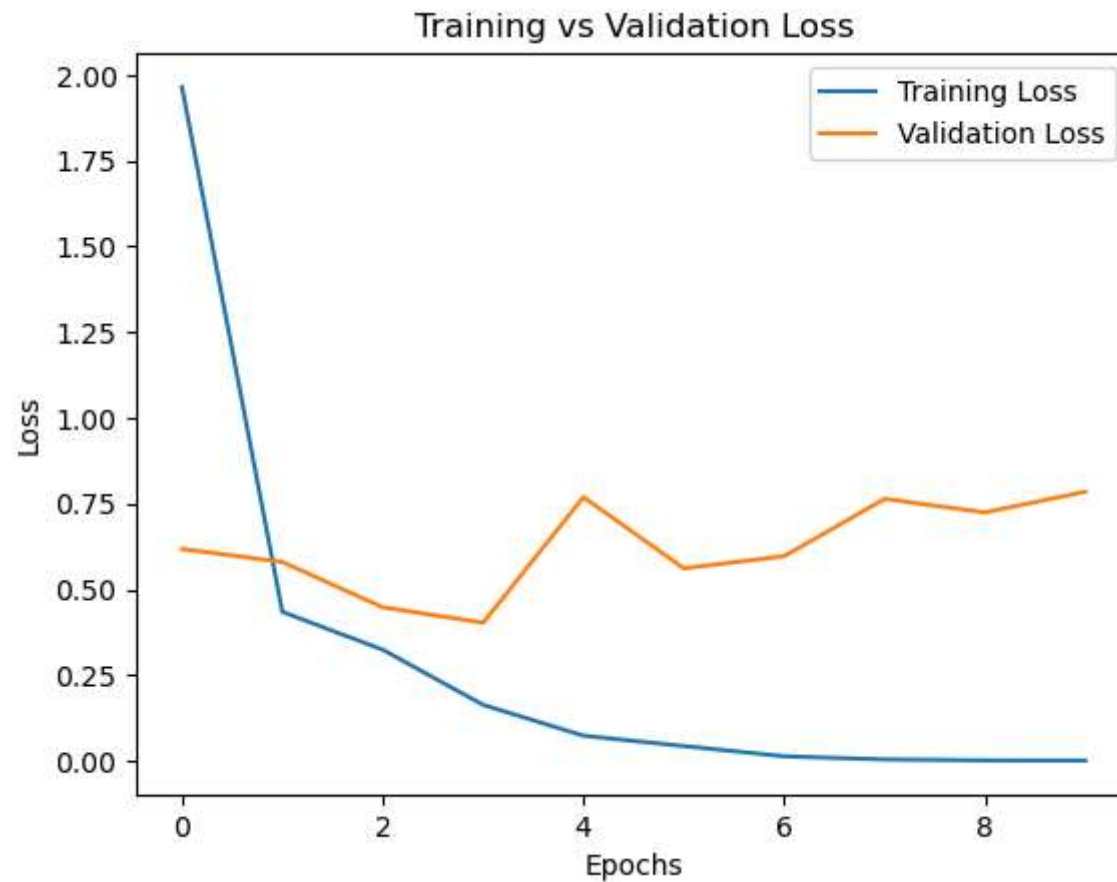
In [7]: plt.plot(history.history['accuracy'], label='Train Accuracy')
plt.plot(history.history['val_accuracy'], label='Validation Accuracy')
plt.legend()
plt.show()

```



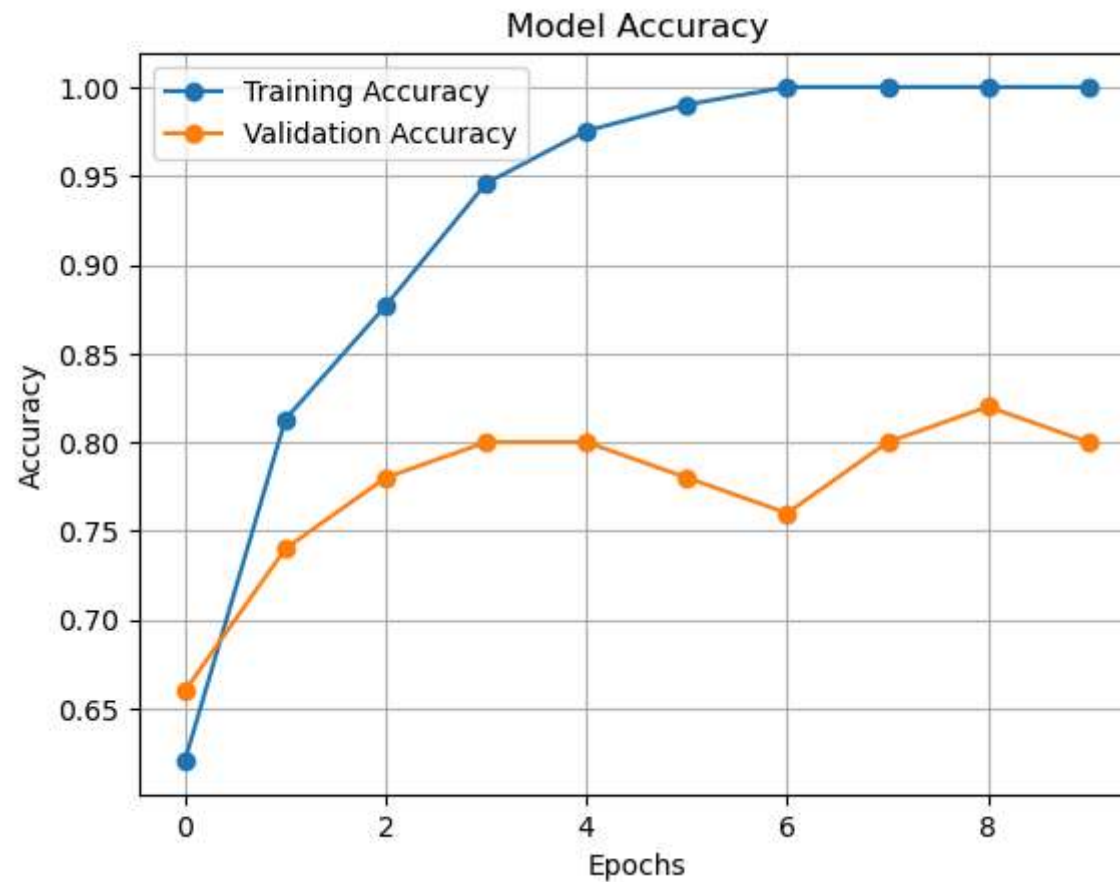
In []:

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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 []:

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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

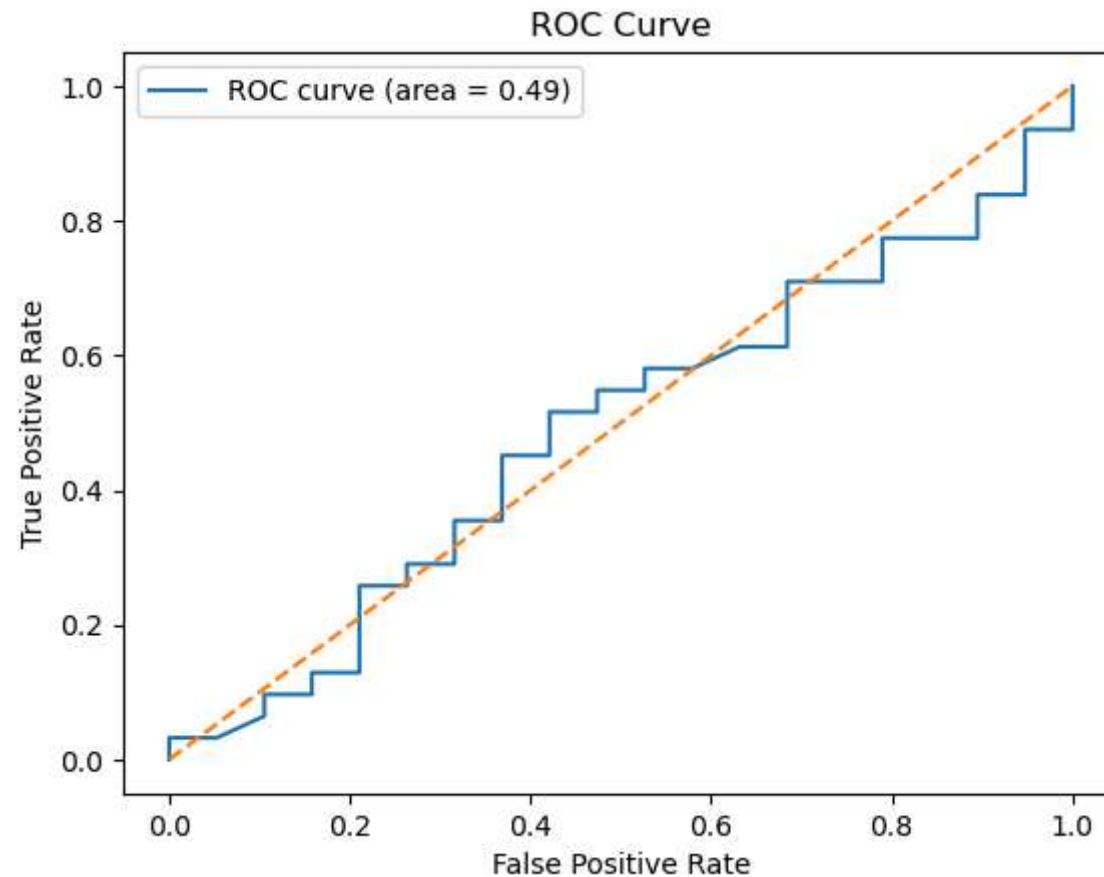
y_prob = model.predict(val_data).ravel()

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

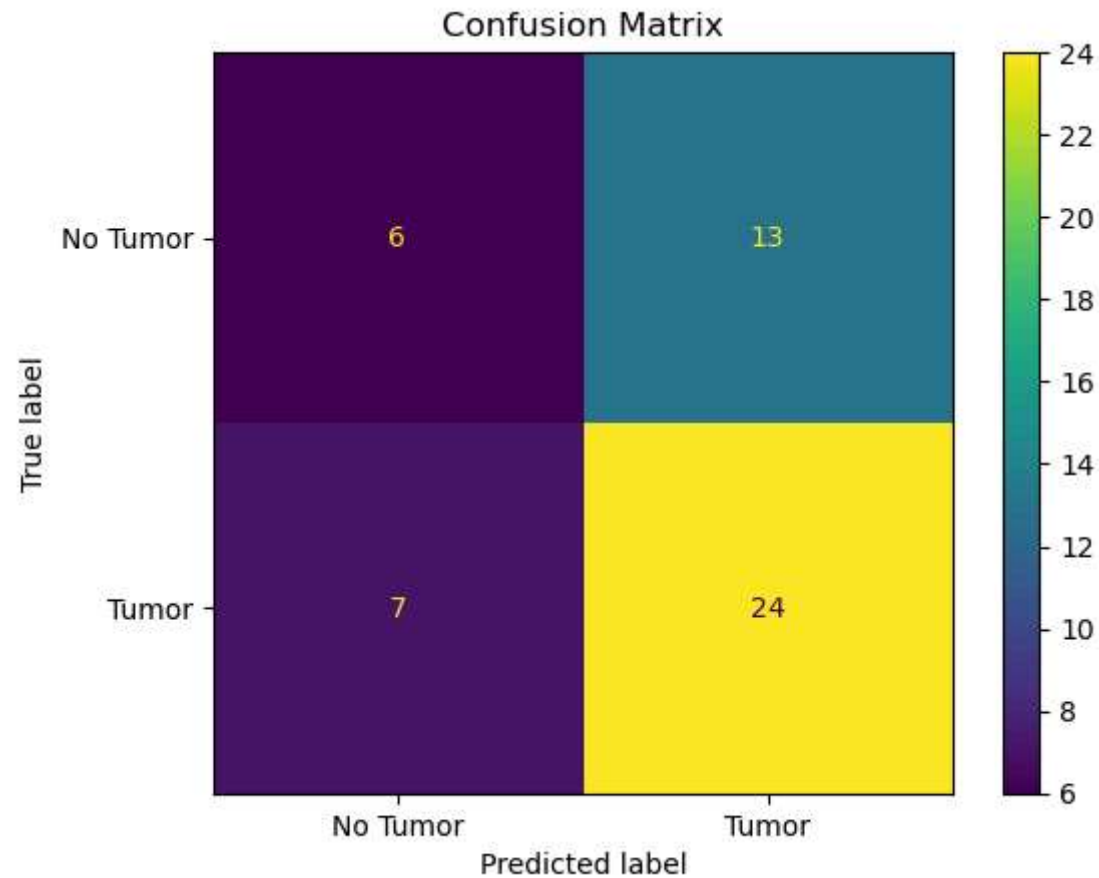


In []:

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

4/4 ————— 1s 107ms/step



In []:

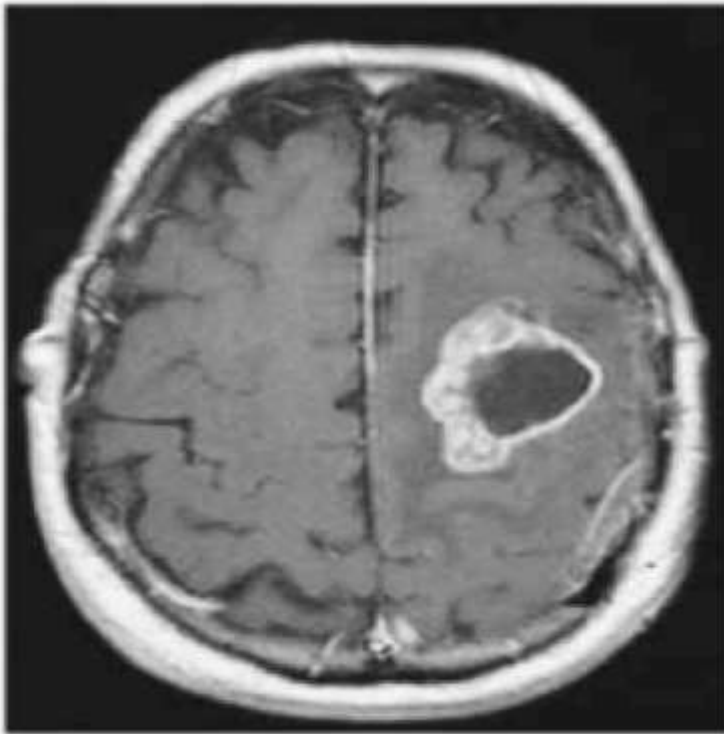
```
In [12]: import random  
import cv2  
  
class_names = ["No Tumor", "Tumor"]
```



```
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()
```

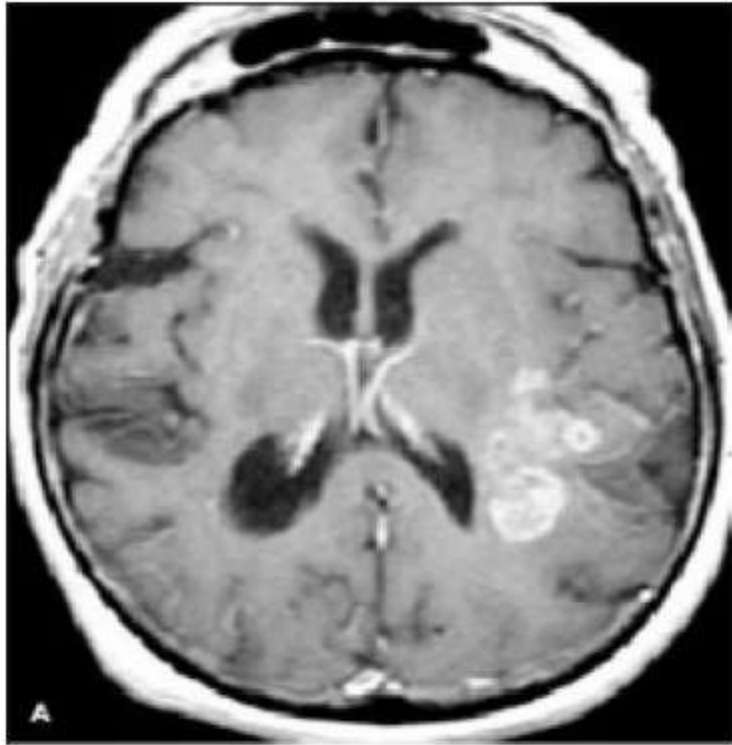
1/1 ————— 0s 174ms/step

Prediction: Tumor



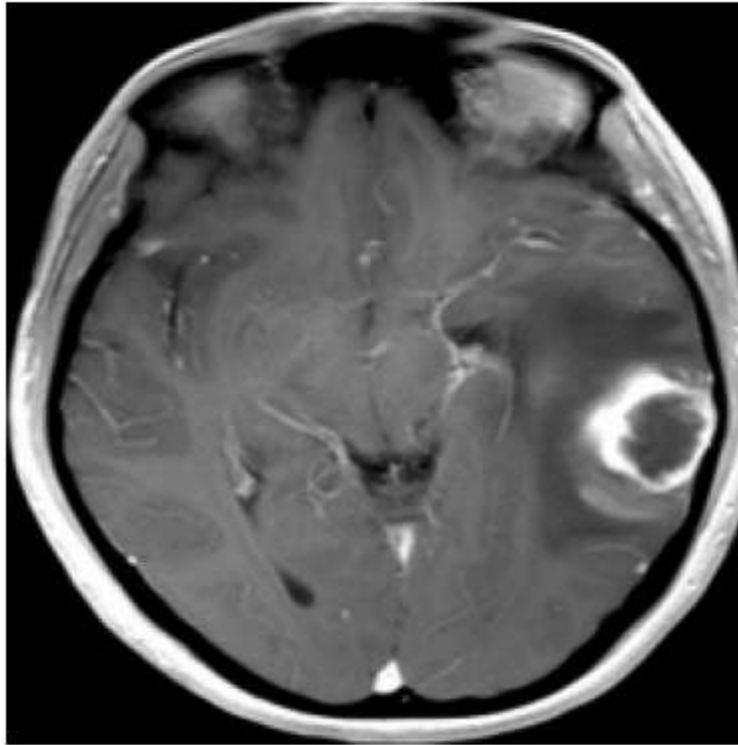
1/1 ————— 0s 82ms/step

Prediction: Tumor



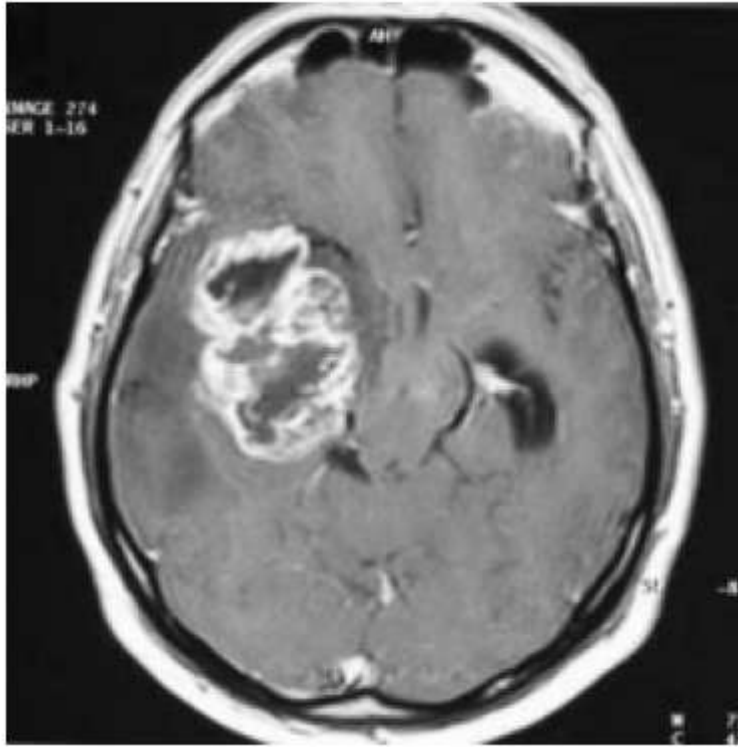
1/1 — 0s 87ms/step

Prediction: Tumor



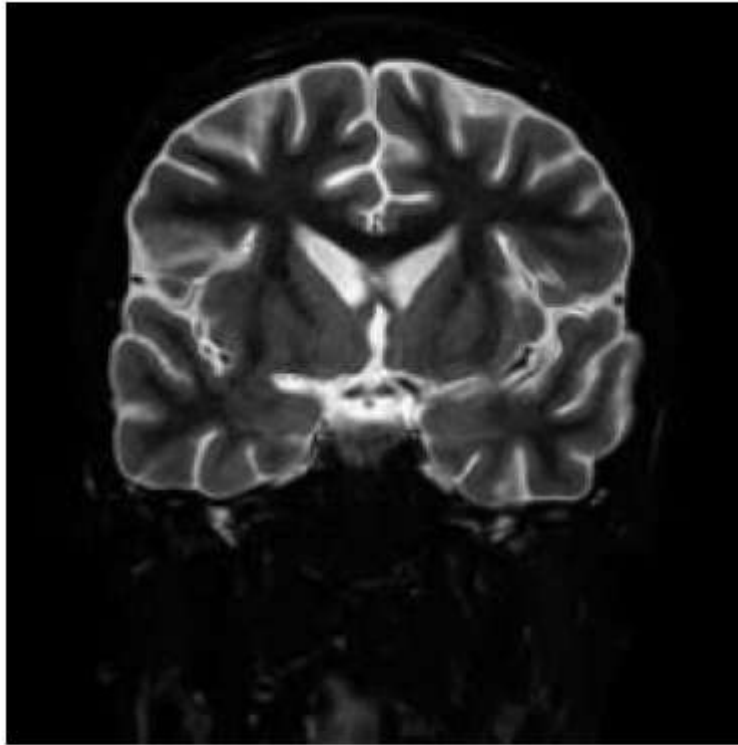
1/1 — 0s 83ms/step

Prediction: Tumor



1/1 — 0s 80ms/step

Prediction: No Tumor



In []:

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

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