Building a Brain Tumor Detection Model on Google Colab

A Step-by-Step Guide from Dataset Download to Model Training

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Objective

- Download the dataset from Kaggle.
- Upload files to Google Drive and access them in Google Colab.
- Create and train a convolutional neural network (CNN) on the dataset.

Prerequisites

1. Kaggle (for downloading the dataset).
2. Google account (for Google Colab and Drive).
3. Basic understanding of Python and TensorFlow

Download Dataset from Kaggle

- **Steps 1**:

1. Go to the Kaggle dataset link: **Brain Tumor Dataset - Yes/No Class.**

https://www.kaggle.com/datasets/princelv84/brain-tumor-dataset-yesno-class

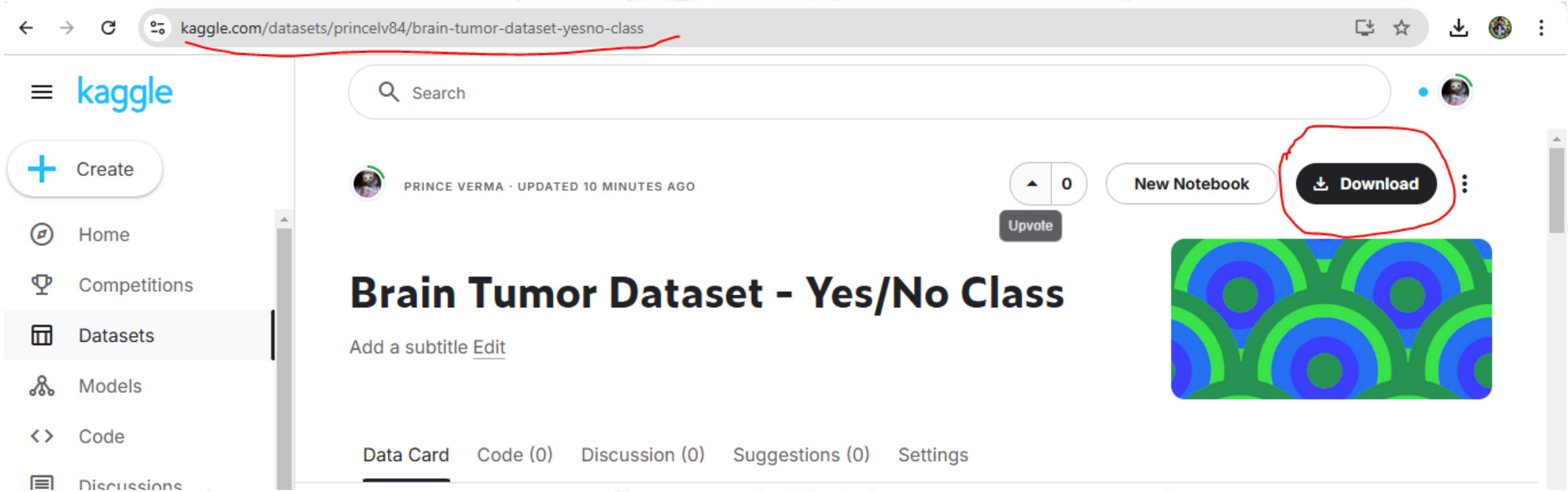
- **Step 2**:

1. Click on the **Download** button.
2. Save the .zip file to your local machine.

Download Dataset from Kaggle

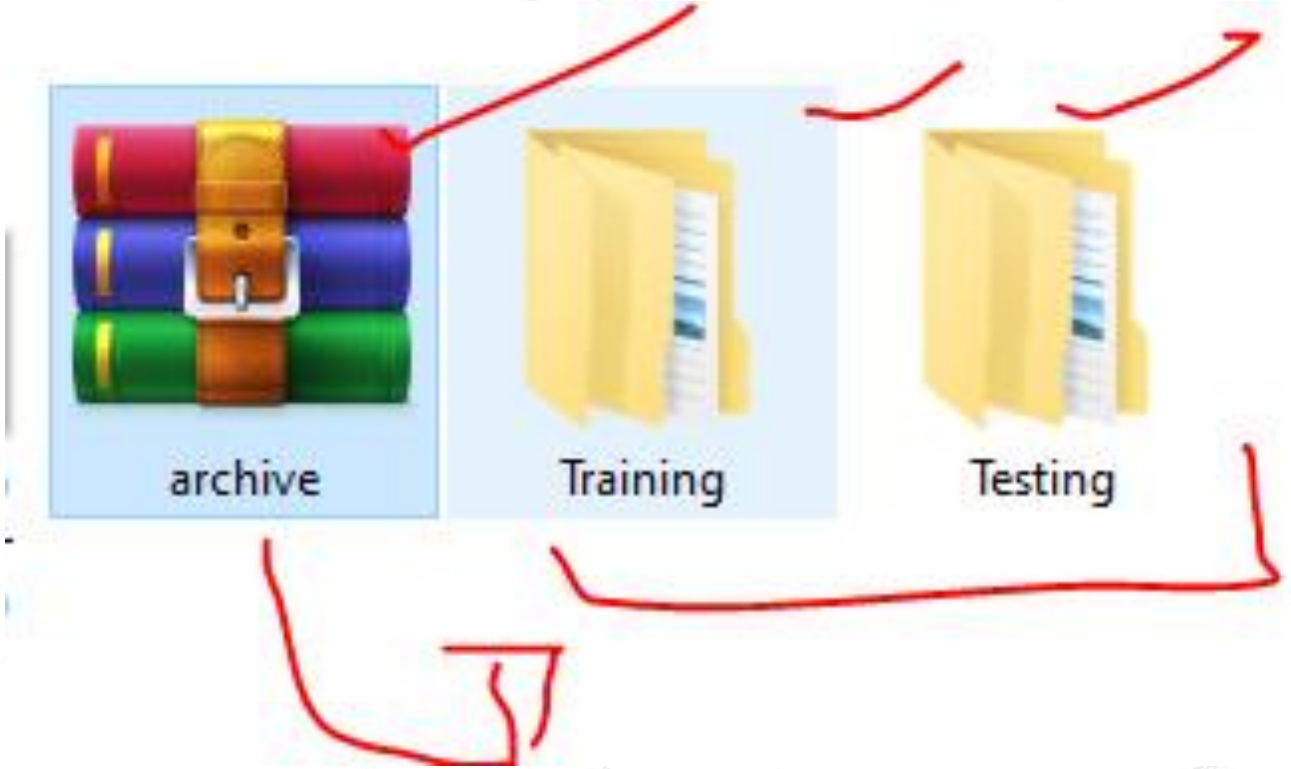
Screenshot:

https://www.kaggle.com/datasets/princelv84/brain-tumor-dataset-yesno-class



Extract the Dataset

- Use winrar or 7zip to extract that zip we just downloaded from Kaggle.

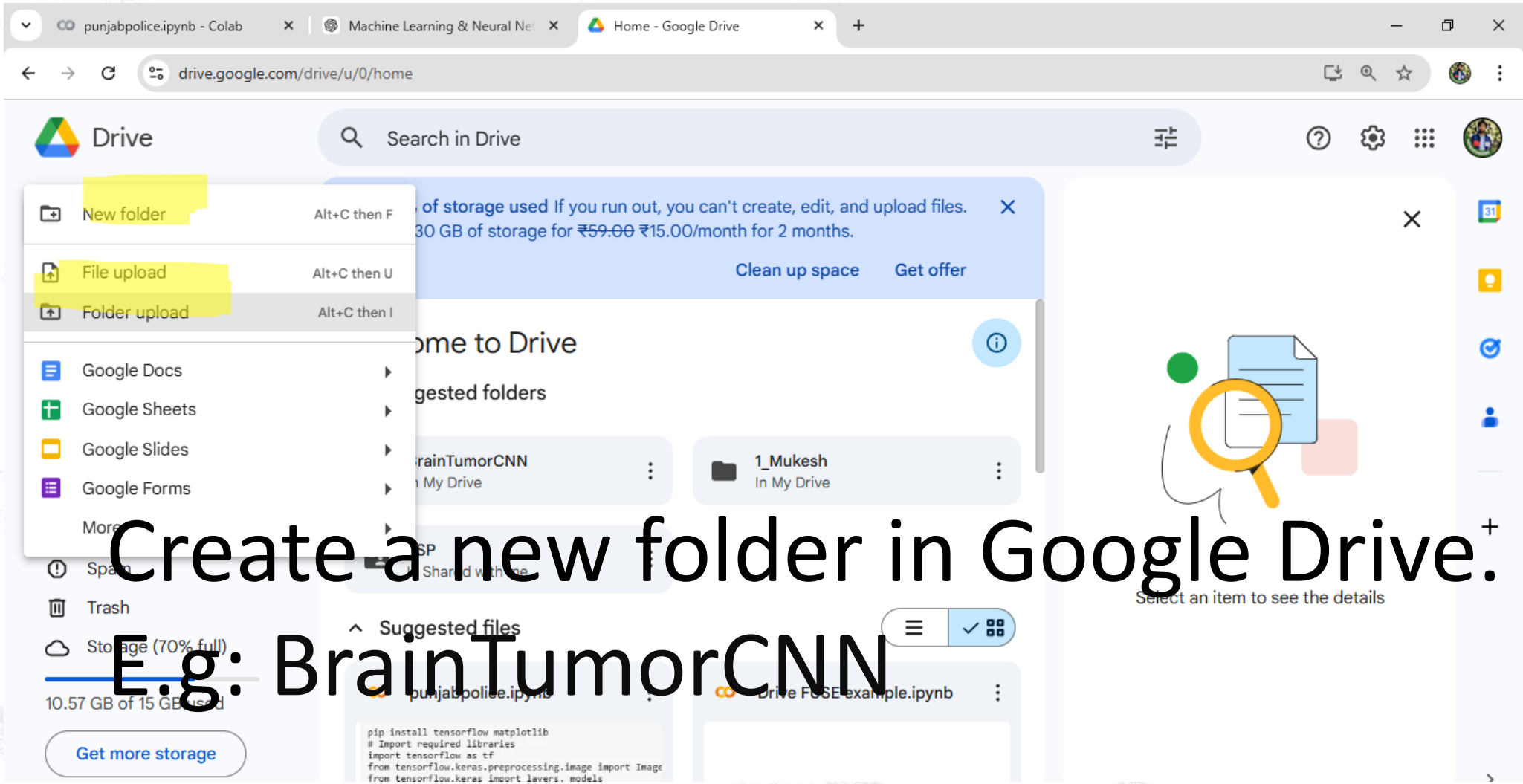


Upload Dataset to Google Drive

- **Steps**:

1. Open Google Drive.

Create a folder named BrainTumorCNN.



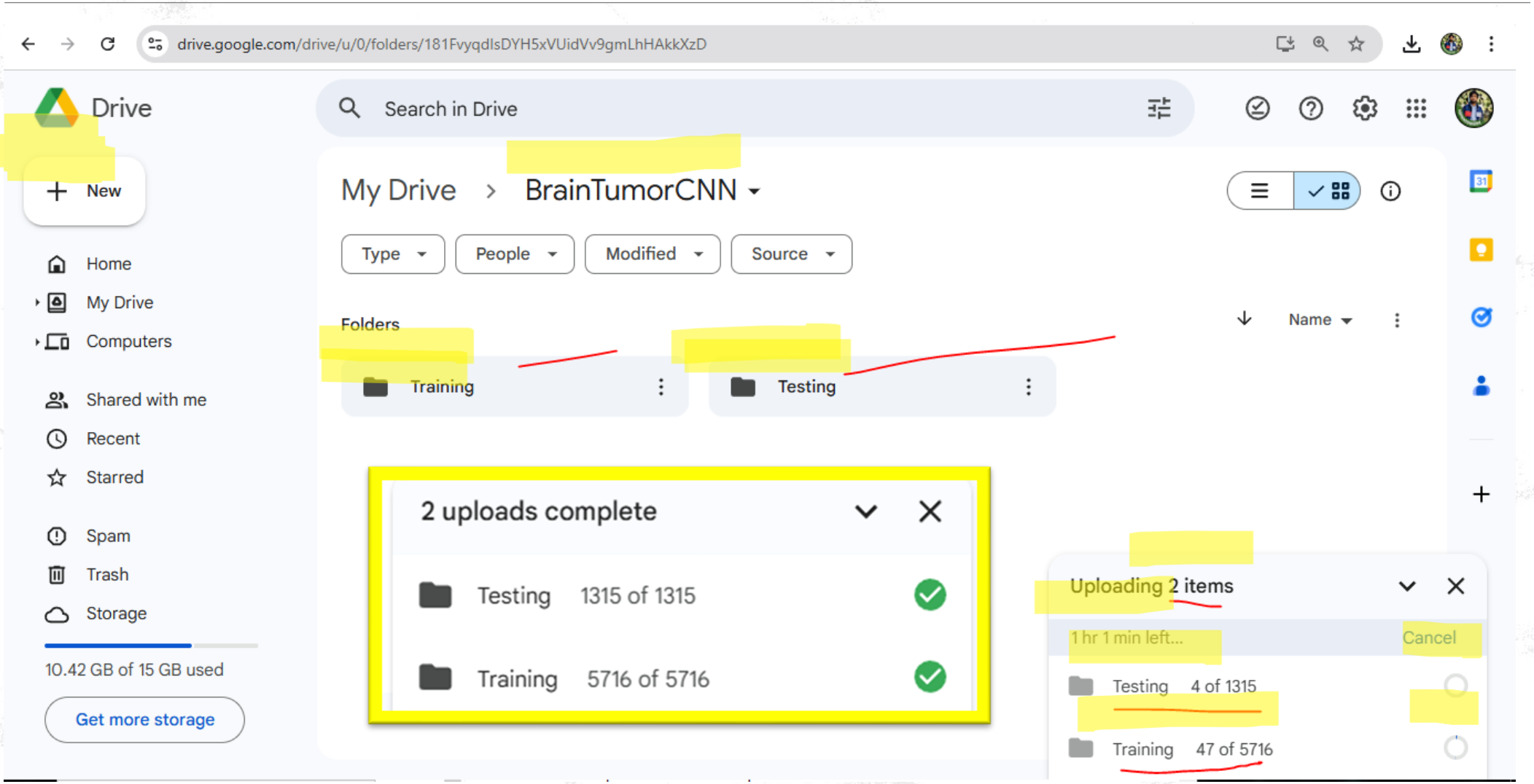
Create a new folder in Google Drive.

E.g: BrainTumorCNN

Upload Dataset to Drive

- Open that newly **created folder .**
- Then click on **New** again this this click on **folder upload.**
- Upload both **Testing** and **Training** folders we just extracted in our local pc to google drive inside that folder. To do So Drag and drop both Testing and Training folders we just extracted in our local pc to google drive inside that Drive folder. E.g: BrainTumorCNN is the name of that Drive folder **in my case.**
- You must Upload both Testing and Training folders **One by One.**
- Now **wait** for them to upload.

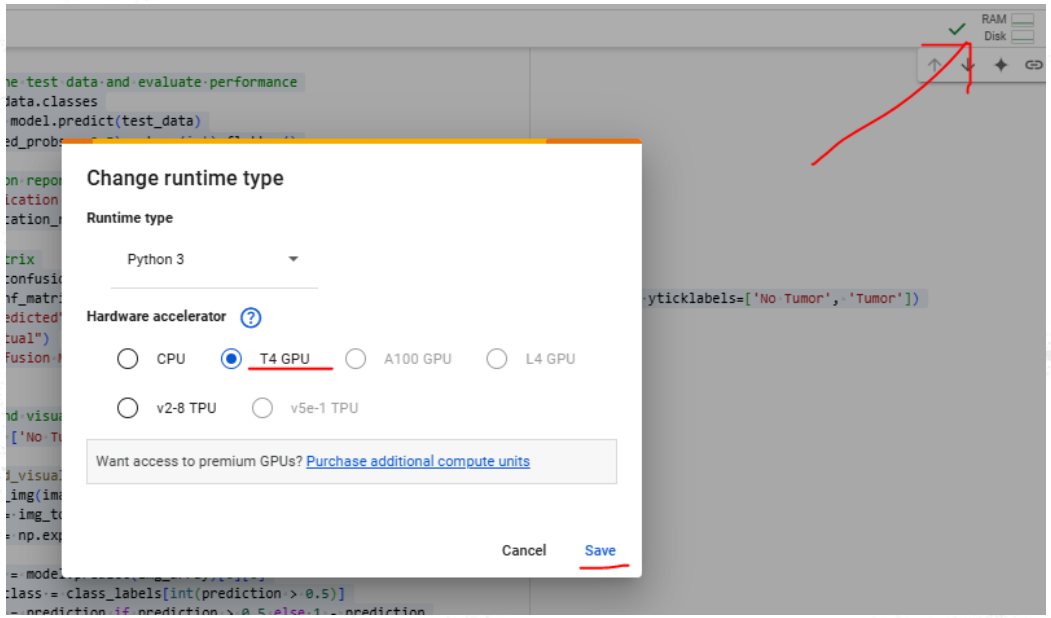
Upload Dataset to Drive



Setup in Google Colab

- **Steps**:

1. Open Google Colab.
2. Create a new notebook.
3. Set runtime type to T4 GPU
4. Mount Google Drive:

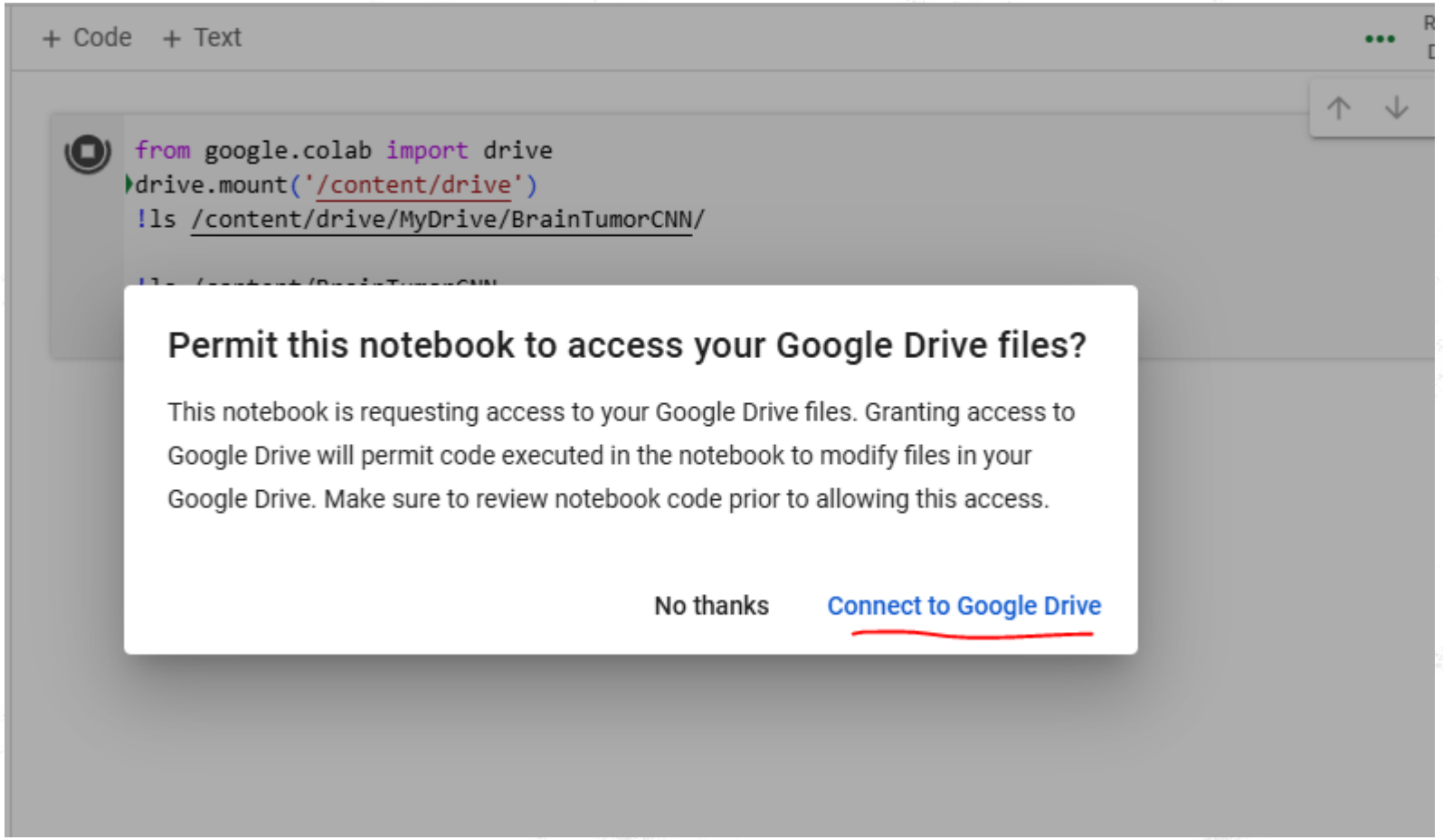



```
# Mount Google Drive
from google.colab import drive
drive.mount('/content/drive')
```

- Navigate to the folder where the dataset is located:

```
!ls /content/drive/MyDrive/BrainTumorCNN/
```

Permit notebook access to Your Google Drive



Ensuring Drive Connection

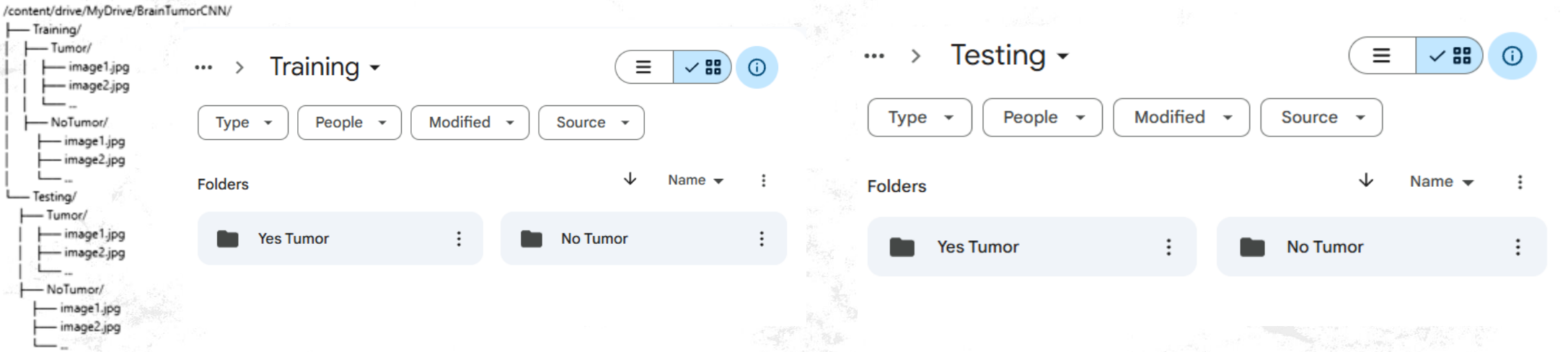
```
from google.colab import drive
drive.mount('/content/drive')

!ls /content/drive/MyDrive/BrainTumorCNN/
```

```
Drive already mounted at /content/drive; to
Testing  Training
```

Dataset Preparation

```
# Mount Google Drive
from google.colab import drive
drive.mount('/content/drive')

# Import necessary libraries
import os
import numpy as np
import matplotlib.pyplot as plt
from tensorflow.keras.preprocessing.image import ImageDataGenerator
from tensorflow.keras.utils import load_img, img_to_array
from tensorflow.keras import layers, models
from sklearn.metrics import classification_report, confusion_matrix
import seaborn as sns
```

Dataset paths




```
# Dataset paths

train_dir = '/content/drive/MyDrive/BrainTumorCNN/Training'

test_dir = '/content/drive/MyDrive/BrainTumorCNN/Testing'
# Verify the folder structure
print("Training folders:", os.listdir(train_dir))
print("Testing folders:", os.listdir(test_dir))
```

Data generators for training and testing

```
# Data generators for training and testing
train_datagen = ImageDataGenerator(rescale=1.0/255.0)
test_datagen = ImageDataGenerator(rescale=1.0/255.0)

# Load data from the organized folders
train_data = train_datagen.flow_from_directory(
    train_dir,
    target_size=(128, 128),
    batch_size=32,
    class_mode='binary’
)

test_data = test_datagen.flow_from_directory(
    test_dir,
    target_size=(128, 128),
    batch_size=32,
    class_mode='binary',
    shuffle=False  # For evaluation consistency
)
```

Define the CNN model

```
# Define the CNN model
model = models.Sequential([
    layers.Conv2D(32, (3, 3), activation='relu', input_shape=(128, 128, 3)),
    layers.MaxPooling2D((2, 2)),
    layers.Conv2D(64, (3, 3), activation='relu'),
    layers.MaxPooling2D((2, 2)),
    layers.Conv2D(128, (3, 3), activation='relu'),
    layers.MaxPooling2D((2, 2)),
    layers.Flatten(),
    layers.Dense(128, activation='relu'),
    layers.Dense(1, activation='sigmoid')  # Binary classification
])
```

Compile / Train/ Save the model

```
# Compile the model
model.compile(optimizer='adam',
              loss='binary_crossentropy',
              metrics=['accuracy'])

# Train the model
history = model.fit(train_data, validation_data=test_data, epochs=10)

# Evaluate the model
test_loss, test_acc = model.evaluate(test_data, verbose=2)
print(f"Test Accuracy: {test_acc:.2f}")

# Save the model
model_save_path = '/content/drive/MyDrive/braintumor_binary.h5'
model.save(model_save_path)
print(f"Model saved successfully at {model_save_path}!")
```

Load the model (if needed later)

1. **Save the model:**

```
model.save('/content/drive/MyDrive/braintumor.h5')
print("Model saved successfully!")
```

2. **Load the model later:**

```
from tensorflow.keras.models import load_model
model =
load_model('/content/drive/MyDrive/braintumor.h5')
```

Visualize Training Results

```
# Plot accuracy and loss graphs
plt.plot(history.history['accuracy'], label='Training Accuracy')
plt.plot(history.history['val_accuracy'], label='Validation Accuracy')
plt.legend()
plt.title("Accuracy over Epochs")
plt.show()

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

# Predict on the test data and evaluate performance
y_true = test_data.classes
y_pred_probs = model.predict(test_data)
y_pred = (y_pred_probs > 0.5).astype(int).flatten()
```

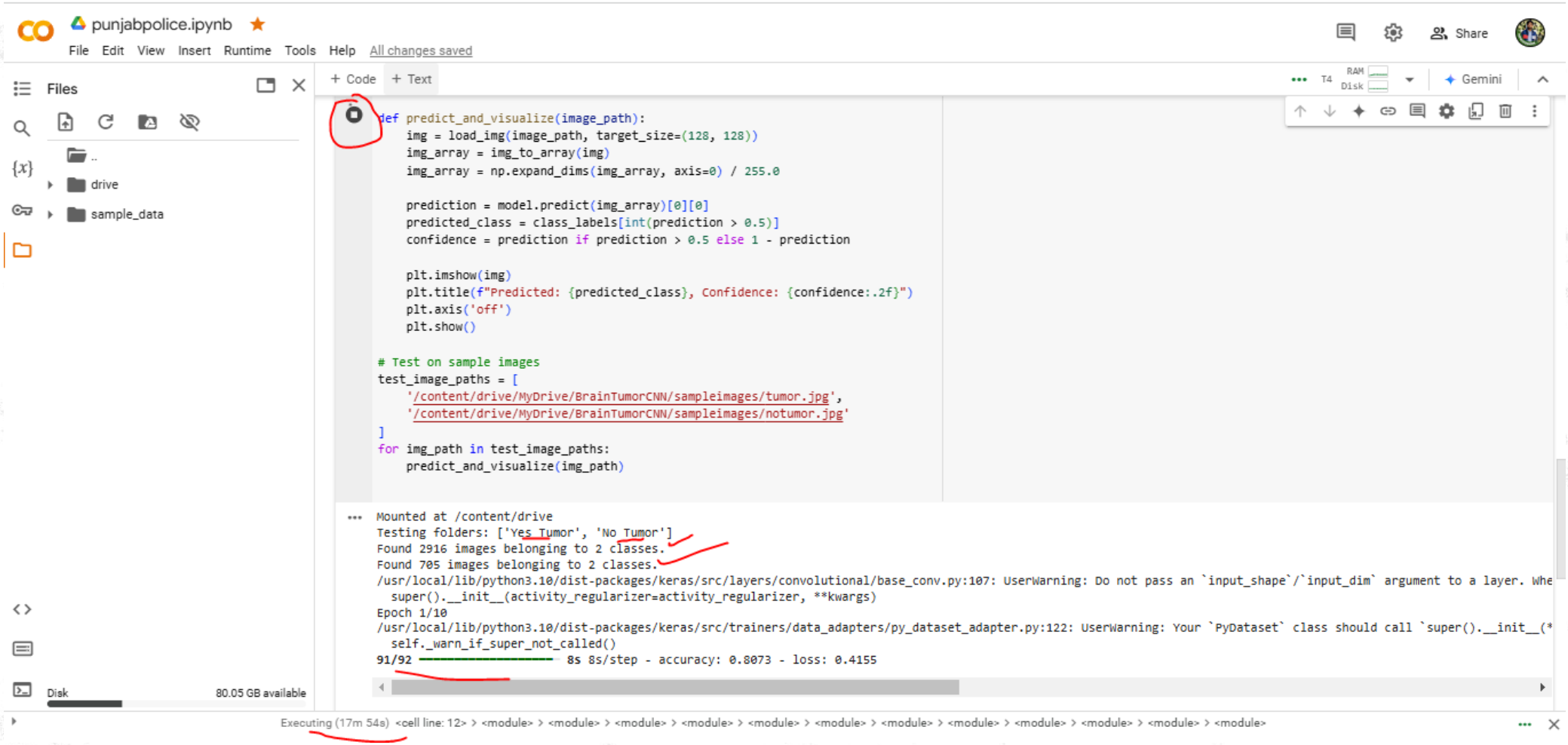
```
# Classification report
print("Classification Report:")
print(classification_report(y_true, y_pred, target_names=['No Tumor',
'Tumor’]))

# Confusion matrix
conf_matrix = confusion_matrix(y_true, y_pred)
sns.heatmap(conf_matrix, annot=True, fmt='d', cmap='Blues',
xticklabels=['No Tumor', 'Tumor'], yticklabels=['No Tumor', 'Tumor'])
plt.xlabel("Predicted")
plt.ylabel("Actual")
plt.title("Confusion Matrix")
plt.show()
```

Visualize Training Results

```
# Prediction and visualization function

class_labels = ['No Tumor', 'Tumor']
def predict_and_visualize(image_path):

    img = load_img(image_path, target_size=(128, 128))

    img_array = img_to_array(img)

    img_array = np.expand_dims(img_array, axis=0) / 255.0


    prediction = model.predict(img_array)[0][0]

    predicted_class = class_labels[int(prediction > 0.5)]

    confidence = prediction if prediction > 0.5 else 1 - prediction


    plt.imshow(img)

    plt.title(f"Predicted: {predicted_class}, Confidence: {confidence:.2f}")

    plt.axis('off')

    plt.show()
```

```
# Test on sample images

test_image_paths =
['/content/drive/MyDrive/BrainTumorCNN/sampleimages/tumor.jpg’,

'/content/drive/MyDrive/BrainTumorCNN/sampleimages/notumor.jpg'

]

for img_path in test_image_paths:

    predict_and_visualize(img_path)
```

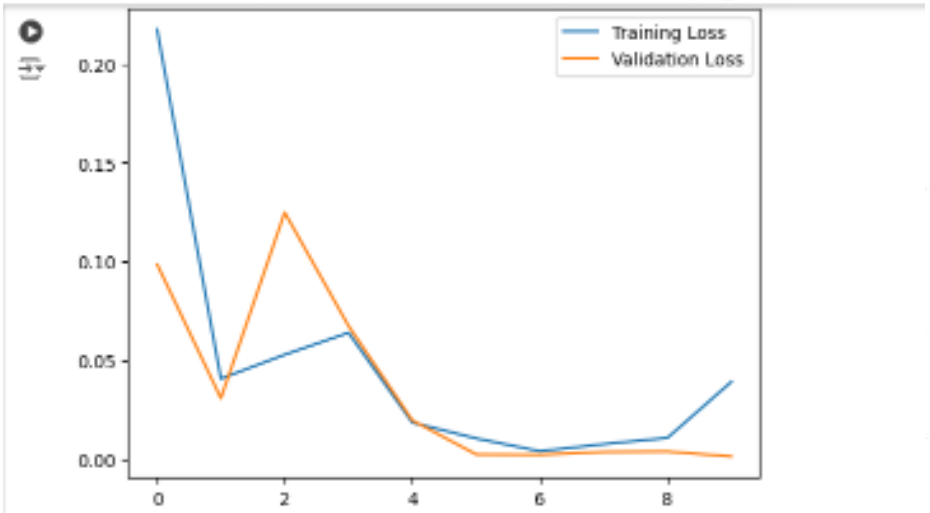
Run the Code and wait.....



Will take forever wait until model is created .....☺

Check results

https://colab.research.google.com/drive/1c7S07QIDgW4K73jo5AcxIaBMfcbvU2GL#scrollTo=Lq9o3b1gWPBE




```
23/23 ━━━━━━━━━━━━━━━━━━━━ 4s 132ms/step
Classification Report:
              precision    recall  f1-score   support

    No Tumor       1.00      1.00      1.00       405
       Tumor       1.00      1.00      1.00       300

    accuracy                           1.00       705
   macro avg       1.00      1.00      1.00       705
weighted avg       1.00      1.00      1.00       705
```

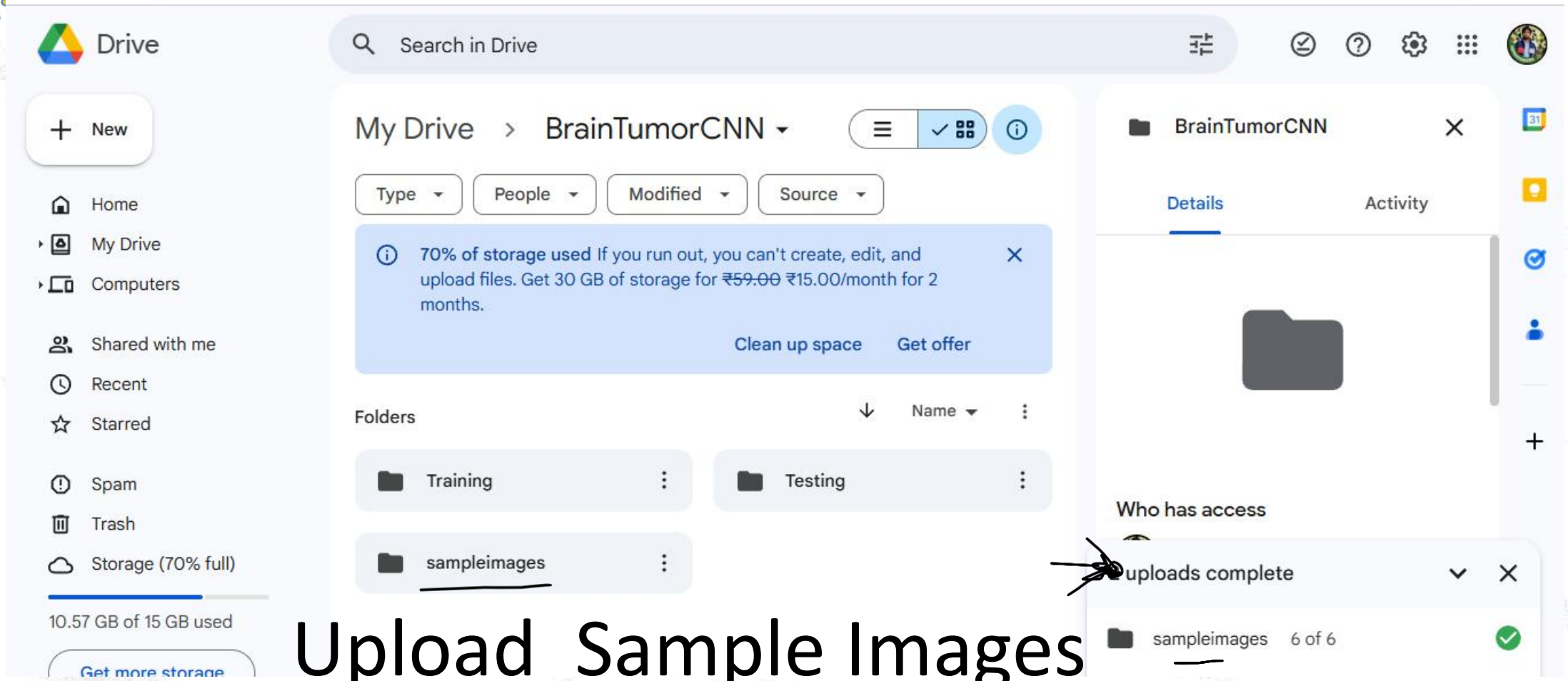
Results

- Achieved training and testing accuracy.
- Saved the trained model.

Next Steps:

- Visualized training progress with graphs.

Visualized training progress with graphs.

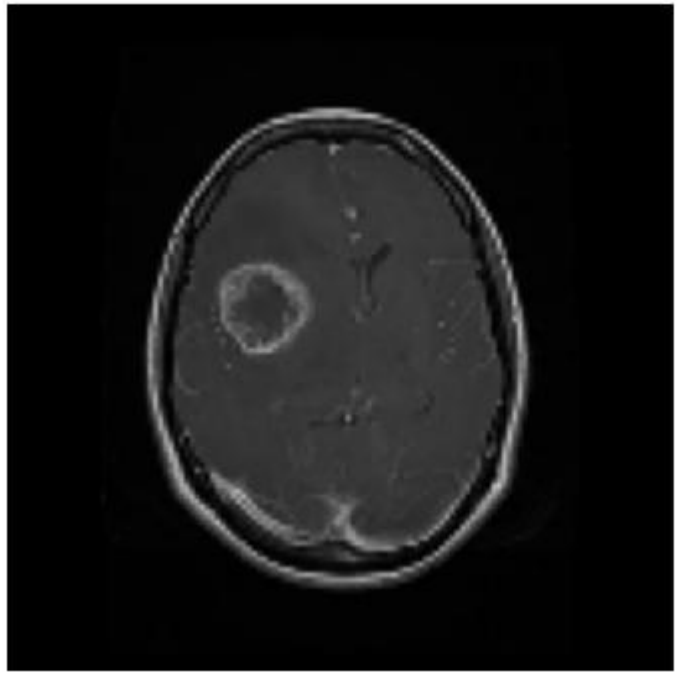


Upload Sample Images

See Results

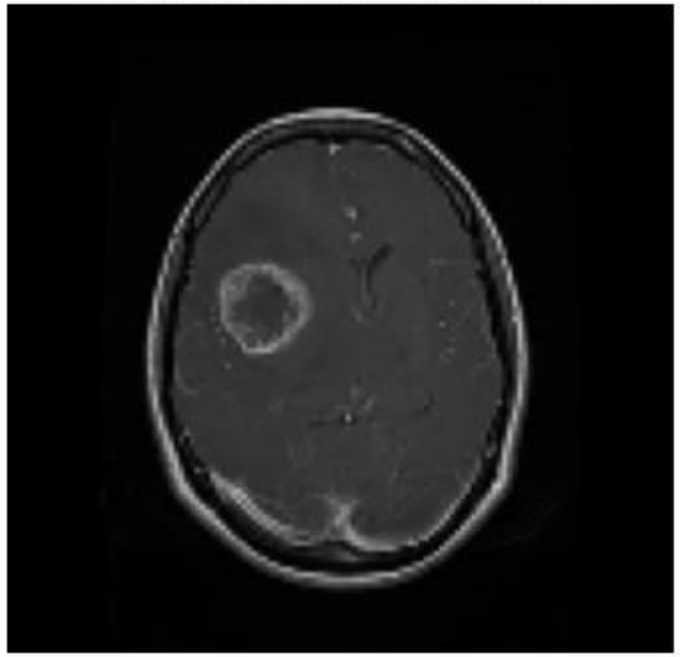
1/1 ——— 0s 16ms/step

Predicted: Tumor, Confidence: 0.98



1/1 ——— 0s 23ms/step

Predicted: Tumor, Confidence: 0.98



References

- Dataset: **Brain Tumor Dataset - Yes/No Class**

https://www.kaggle.com/datasets/princelv84/brain-tumor-dataset-yesno-class

- TensorFlow documentation
  https://www.tensorflow.org/api_docs/python/tf
- Google Colab
  https://colab.research.google.com/drive/1c7S07QIDgW4K73jo5AcxIaBMfcbvU2GL?usp=sharing