(+ Code) (+ Text

data loading

```
from google.colab import drive
drive.mount('/content/drive')
import os
import shutil
import random
import numpy as np
import pandas as pd
import cv2
import matplotlib.pyplot as plt
from google.colab import drive
try:
    drive.mount('/content/drive')
   print("Google Drive mounted successfully!")
   print("Running locally or Drive already mounted")
lgg_dataset_path = '/content/drive/MyDrive/ddd'
output_dir = '/content/drive/MyDrive/brain_tumor_dataset'
os.makedirs(os.path.join(output_dir, 'images'), exist_ok=True)
os.makedirs(os.path.join(output_dir, 'masks'), exist_ok=True)
def prepare dataset(num samples=10, random selection=True):
    case_dirs = [d for d in os.listdir(lgg_dataset_path)
                if os.path.isdir(os.path.join(lgg_dataset_path, d))]
    print(f"Found {len(case dirs)} case directories")
    if random_selection:
        selected_cases = random.sample(case_dirs, min(len(case_dirs), num_samples))
    else:
        selected_cases = case_dirs[:min(len(case_dirs), num_samples)]
    print(f"Selected {len(selected_cases)} cases")
    processed_count = 0
    for case_id in selected_cases:
        case_dir = os.path.join(lgg_dataset_path, case_id)
       files = os.listdir(case_dir)
        image_files = [f for f in files if f.endswith('.tif') and '_mask' not in f]
        for img_file in image_files:
            mask_file = img_file.replace('.tif', '_mask.tif')
            if mask_file in files:
                img_path = os.path.join(case_dir, img_file)
                mask_path = os.path.join(case_dir, mask_file)
                img = cv2.imread(img_path)
                mask = cv2.imread(mask_path, cv2.IMREAD_GRAYSCALE)
                if img is not None and mask is not None:
                    img = cv2.cvtColor(img, cv2.COLOR_BGR2RGB)
                    flair = img[:, :, 1]
```

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out_img_name = f"brain_tumor_{processed_count:03d}.png"
                    out_mask_name = f"brain_tumor_{processed_count:03d}_mask.png"
                    out_img_path = os.path.join(output_dir, 'images', out_img_name)
                    out_mask_path = os.path.join(output_dir, 'masks', out_mask_name)
                    cv2.imwrite(out_img_path, flair)
                    cv2.imwrite(out_mask_path, mask)
                    processed_count += 1
                    print(f"Processed {processed_count}/{num_samples}: {out_img_name}")
                    if processed_count >= num_samples:
                        return
def visualize dataset(dataset path, num samples=5):
    images_dir = os.path.join(dataset_path, 'images')
    masks_dir = os.path.join(dataset_path, 'masks')
    image_files = sorted(os.listdir(images_dir))
    image_files = image_files[:min(len(image_files), num_samples)]
    plt.figure(figsize=(12, 4 * len(image_files)))
    for i, img_file in enumerate(image_files):
        if img_file.replace('.png', '_mask.png') in os.listdir(masks_dir):
           mask_file = img_file.replace('.png', '_mask.png')
           mask_file = next((m for m in os.listdir(masks_dir) if m.startswith(img_file.split('.')[0])), None)
        if mask_file:
           img_path = os.path.join(images_dir, img_file)
           mask_path = os.path.join(masks_dir, mask_file)
           img = cv2.imread(img_path, cv2.IMREAD_GRAYSCALE) # Read as grayscale
           mask = cv2.imread(mask_path, cv2.IMREAD_GRAYSCALE)
           plt.subplot(len(image_files), 2, i * 2 + 1)
           plt.imshow(img, cmap='gray')
           plt.title(f"MRI Image: {img_file}")
           plt.axis('off')
           plt.subplot(len(image_files), 2, i * 2 + 2)
           plt.imshow(mask, cmap='gray')
           plt.title(f"Tumor Mask: {mask_file}")
           plt.axis('off')
    plt.tight_layout()
    plt.show()
print("Preparing dataset...")
prepare_dataset(num_samples=10, random_selection=True)
print("Dataset preparation completed!")
print("Visualizing prepared dataset...")
visualize_dataset(output_dir, num_samples=5)
print("\nDataset is ready to use with the brain tumor segmentation code!")
print(f"Images directory: {os.path.join(output_dir, 'images')}")
print(f"Masks directory: {os.path.join(output_dir, 'masks')}")
print("\nUpdate these paths in the main code:")
print("base_dir = '", output_dir, "'")
print("images_dir = os.path.join(base_dir, 'images')")
print("masks_dir = os.path.join(base_dir, 'masks')")
```

Tumor Mask: brain tumor 000 mask.png

Tumor Mask: brain_tumor_001_mask.png

import os import cv2 import numpy as np

try:

except:

import matplotlib.pyplot as plt from google.colab import drive

drive.mount('/content/drive')

def __init__(self, base_dir=None):

self.base_dir = base_dir self.images = [] self.masks = []

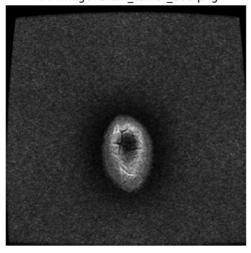
self.processed_images = [] self.segmented_masks = [] self.metrics = {}

print("Loading dataset...")

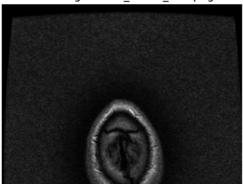
 ${\tt class\ BrainTumorSegmentation:}$

```
Trive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force_remount=True).
    Google Drive mounted successfully!
    Preparing dataset...
    Found 1 case directories
    Selected 1 cases
    Processed 1/10: brain_tumor_000.png
Processed 2/10: brain_tumor_001.png
    Processed 3/10: brain_tumor_002.png
    Processed 4/10: brain_tumor_003.png
    Processed 5/10: brain_tumor_004.png
    Processed 6/10: brain_tumor_005.png
    Processed 7/10: brain_tumor_006.png
    Processed 8/10: brain_tumor_007.png
    Processed 9/10: brain_tumor_008.png
    Processed 10/10: brain_tumor_009.png
    Dataset preparation completed!
    Visualizing prepared dataset...
```

MRI Image: brain_tumor_000.png



MRI Image: brain_tumor_001.png



from sklearn.model_selection import train_test_split

print("Google Drive mounted successfully!")

print("Running locally or Drive already mounted")

```
from \ sklearn. metrics \ import \ accuracy\_score, \ jaccard\_score, \ f1\_score, \ precision\_score, \ recall\_score
```

def load_dataset(self, images_dir, masks_dir, max_samples=None):

```
image_files = sorted(os.listdir(images_dir))
    loaded_count = 0
    for img_file in image_files:
        if not img_file.endswith(('.jpg', '.png', '.jpeg', '.tif')):
        if '_mask' not in img_file:
            mask_file = img_file.replace('.png', '_mask.png')
mask_file = mask_file.replace('.tif', '_mask.tif')
mask_file = mask_file.replace('.jpg', '_mask.jpg')
            image_path = os.path.join(images_dir, img_file)
            mask_path = os.path.join(masks_dir, mask_file)
            if not os.path.exists(mask path):
                print(f"Warning: No mask found for {img_file}")
            image = cv2.imread(image_path)
            mask = cv2.imread(mask_path, cv2.IMREAD_GRAYSCALE)
            if image is not None and mask is not None:
                 if len(image.shape) == 3 and image.shape[2] == 3:
                     gray_image = image[:, :, 1]
                 else:
                     gray_image = image.copy()
                 if len(gray_image.shape) == 3:
                     gray_image = cv2.cvtColor(gray_image, cv2.COLOR_BGR2GRAY)
                 _, binary_mask = cv2.threshold(mask, 127, 255, cv2.THRESH_BINARY)
                 self.images.append(gray_image)
                 self.masks.append(binary_mask)
                 loaded_count += 1
                 if max_samples is not None and loaded_count >= max_samples:
                     break
    print(f"Loaded {len(self.images)} images and {len(self.masks)} masks.")
def preprocess_images(self):
    print("Preprocessing images...")
    self.processed_images = []
    for image in self.images:
        if len(image.shape) > 2:
            gray_image = cv2.cvtColor(image, cv2.COLOR_RGB2GRAY)
        else:
            gray_image = image.copy()
        if gray_image.max() > 0:
            normalized = ((gray_image - gray_image.min()) /
                          (gray_image.max() - gray_image.min()) * 255).astype(np.uint8)
        else:
            normalized = gray image
        clahe = cv2.createCLAHE(clipLimit=2.0, tileGridSize=(8, 8))
        enhanced = clahe.apply(normalized)
        blurred = cv2.GaussianBlur(enhanced, (5, 5), 0)
        self.processed_images.append(blurred)
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```
print(f"Preprocessed {len(self.processed images)} images.")
def segment_tumors(self, method='watershed'):
   print(f"Segmenting tumors using {method} method...")
   self.segmented_masks = []
   for image in self.processed_images:
       if method == 'threshold':
            _, segmented = cv2.threshold(image, 0, 255, cv2.THRESH_BINARY + cv2.THRESH_OTSU)
       elif method == 'watershed':
            _, thresholded = cv2.threshold(image, 0, 255, cv2.THRESH_BINARY + cv2.THRESH_OTSU)
            kernel = np.ones((3, 3), np.uint8)
            opening = cv2.morphologyEx(thresholded, cv2.MORPH_OPEN, kernel, iterations=2)
            sure_bg = cv2.dilate(opening, kernel, iterations=3)
            dist_transform = cv2.distanceTransform(opening, cv2.DIST_L2, 5)
            _, sure_fg = cv2.threshold(dist_transform, 0.7*dist_transform.max(), 255, 0)
            sure_fg = np.uint8(sure_fg)
            unknown = cv2.subtract(sure_bg, sure_fg)
            _, markers = cv2.connectedComponents(sure_fg)
            markers = markers + 1
            markers[unknown == 255] = 0
            markers = cv2.watershed(cv2.cvtColor(image, cv2.COLOR_GRAY2BGR), markers)
            segmented = np.zeros_like(image)
            segmented[markers > 1] = 255
        elif method == 'kmeans':
            image_data = image.reshape((-1, 1))
            image_data = np.float32(image_data)
            criteria = (cv2.TERM_CRITERIA_EPS + cv2.TERM_CRITERIA_MAX_ITER, 100, 0.2)
            k = 3
            _, labels, centers = cv2.kmeans(image_data, k, None, criteria, 10, cv2.KMEANS_RANDOM_CENTERS)
            centers = np.uint8(centers)
            brightest_cluster = np.argmax(centers)
            segmented = np.zeros_like(image)
            segmented[labels.reshape(image.shape) == brightest_cluster] = 255
       else:
           raise ValueError(f"Unknown segmentation method: {method}")
        segmented = self.post_process_mask(segmented)
        self.segmented_masks.append(segmented)
   print(f"Segmented {len(self.segmented masks)} images.")
def post_process_mask(self, mask):
   if mask.dtype != np.uint8:
       mask = mask.astype(np.uint8)
   contours, _ = cv2.findContours(mask, cv2.RETR_EXTERNAL, cv2.CHAIN_APPROX_SIMPLE)
   filled_mask = np.zeros_like(mask)
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cv2.drawContours(filled_mask, contours, -1, 255, -1)
    nb_components, output, stats, _ = cv2.connectedComponentsWithStats(filled_mask, connectivity=8)
    sizes = stats[1:, -1]
   min_size = 100
   processed_mask = np.zeros_like(filled_mask)
    for i in range(1, nb_components):
        if sizes[i - 1] >= min_size:
            processed_mask[output == i] = 255
    return processed_mask
def evaluate_segmentation(self, ground_truth_masks=None):
    if ground_truth_masks is None:
        ground_truth_masks = self.masks
    if len(ground truth masks) != len(self.segmented masks):
        raise ValueError("Mismatch between number of ground truth masks and segmented masks")
    dice_scores = []
    jaccard_scores = []
    precision_scores = []
    recall_scores = []
    for gt_mask, pred_mask in zip(ground_truth_masks, self.segmented_masks):
        gt_binary = np.where(gt_mask > 0, 1, 0).flatten()
        pred_binary = np.where(pred_mask > 0, 1, 0).flatten()
        dice = f1_score(gt_binary, pred_binary, zero_division=1)
        dice scores.append(dice)
        iou = jaccard_score(gt_binary, pred_binary, zero_division=1)
        jaccard_scores.append(iou)
        precision = precision_score(gt_binary, pred_binary, zero_division=1)
        recall = recall_score(gt_binary, pred_binary, zero_division=1)
        precision scores.append(precision)
        recall_scores.append(recall)
    self.metrics = {
        'dice coefficient': np.mean(dice scores),
        'jaccard_index': np.mean(jaccard_scores),
        'precision': np.mean(precision_scores),
        'recall': np.mean(recall_scores)
    }
    print("Segmentation Evaluation Metrics:")
   print(f" Dice Coefficient (F1-Score): {self.metrics['dice_coefficient']:.4f}")
   print(f" Jaccard Index (IoU): {self.metrics['jaccard_index']:.4f}")
print(f" Precision: {self.metrics['precision']:.4f}")
   print(f" Recall: {self.metrics['recall']:.4f}")
    return self.metrics
def visualize_results(self, num_samples=5):
    num samples : int
        Number of samples to visualize
        plt.subplot(num_samples, 3, i * 3 + 1)
        plt.imshow(self.images[i], cmap='gray')
        plt.title(f"Original Image {i+1}")
        plt.axis('off')
        plt.subplot(num_samples, 3, i * 3 + 2)
        plt.imshow(self.masks[i], cmap='gray')
        plt.title(f"Ground Truth Mask {i+1}")
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```
plt.axis('off')
            plt.subplot(num_samples, 3, i * 3 + 3)
            plt.imshow(self.segmented_masks[i], cmap='gray')
            plt.title(f"Segmented Mask \{i+1\}")
            plt.axis('off')
       plt.tight_layout()
       plt.show()
    def overlay_results(self, num_samples=5):
        num_samples = min(num_samples, len(self.images))
       plt.figure(figsize=(12, 4 * num_samples))
        for i in range(num_samples):
            display_img = cv2.cvtColor(self.images[i], cv2.COLOR_GRAY2RGB)
            plt.subplot(num_samples, 2, i * 2 + 1)
            overlay = display img.copy()
            green_mask = np.zeros_like(overlay)
            green_mask[:, :, 1] = self.masks[i]
            alpha = 0.5
            cv2.addWeighted(green_mask, alpha, overlay, 1 - alpha, 0, overlay)
            plt.imshow(overlay)
            plt.title(f"Original + Ground Truth {i+1}")
            plt.axis('off')
            plt.subplot(num samples, 2, i * 2 + 2)
            overlay = display_img.copy()
            red_mask = np.zeros_like(overlay)
            red_mask[:, :, 0] = self.segmented_masks[i] # Red channel
            cv2.addWeighted(red_mask, alpha, overlay, 1 - alpha, 0, overlay)
            plt.imshow(overlay)
            plt.title(f"Original + Segmentation {i+1}")
            plt.axis('off')
       plt.tight_layout()
        plt.show()
    def run_full_pipeline(self, images_dir, masks_dir, max_samples=None, segmentation_method='watershed'):
                self.load_dataset(images_dir, masks_dir, max_samples)
        self.preprocess_images()
        self.segment_tumors(method=segmentation_method)
        self.evaluate_segmentation()
        self.visualize_results()
        self.overlay_results()
        return self.metrics
if __name__ == "__main__":
   base_dir = '/content/drive/MyDrive/brain_tumor_dataset'
    images_dir = os.path.join(base_dir, 'images')
```

```
masks_dir = os.path.join(base_dir, 'masks')
tumor_segmentation = BrainTumorSegmentation(base_dir)
methods = ['threshold', 'watershed', 'kmeans']
results = {}
for method in methods:
   print(f"\n{'-'*50}")
   print(f"Running segmentation with {method.upper()} method")
   print(f"{'-'*50}")
   metrics = tumor_segmentation.run_full_pipeline(
       images_dir=images_dir,
       masks_dir=masks_dir,
       max_samples=20,
       segmentation_method=method
    results[method] = metrics
print("\nComparison of Segmentation Methods:")
print(f"{'Method':<12} {'Dice':<8} {'IoU':<8} {'Precision':<10} {'Recall':<8}")</pre>
print("-" * 50)
for method, metrics in results.items():
    print(f"{method:<12} {metrics['dice_coefficient']:.4f} {metrics['jaccard_index']:.4f} "</pre>
          f"{metrics['precision']:.4f}
                                        {metrics['recall']:.4f}")
```