data loading

```
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
# Mount Google Drive if in Colab
    drive.mount('/content/drive')
    print("Google Drive mounted successfully!")
except:
    print("Running locally or Drive already mounted")
# Set paths
# Adjust these paths based on where you unzipped the LGG dataset
lgg_dataset_path = '/content/drive/MyDrive/brain tumor/ddd' # Path to the original LGG dataset
output_dir = '/content/drive/MyDrive/brain_tumor_dataset' # Where to save the organized dataset
# Create output directories
os.makedirs(os.path.join(output_dir, 'images'), exist_ok=True)
os.makedirs(os.path.join(output_dir, 'masks'), exist_ok=True)
# Function to extract and prepare dataset
def prepare_dataset(num_samples=10, random_selection=True):
    Prepare a subset of the LGG dataset for segmentation
    Parameters:
    -----
    num samples : int
        Number of image-mask pairs to extract
    random_selection : bool
       Whether to select random samples or the first ones
    # Get all case directories
    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:
        # Randomly select cases
        selected_cases = random.sample(case_dirs, min(len(case_dirs), num_samples))
    else:
        # Take the first N cases
        selected_cases = case_dirs[:min(len(case_dirs), num_samples)]
    print(f"Selected {len(selected_cases)} cases")
    # Counter for processed images
    processed_count = 0
    # Process each selected case
    for case id in selected cases:
        case_dir = os.path.join(lgg_dataset_path, case_id)
        # Get all files in this case directory
        files = os.listdir(case dir)
        # Get image files (without mask suffix)
        image files = [f for f in files if f.endswith('.tif') and ' mask' not in f]
```

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for img_file in image_files:
            # Get corresponding mask file
            mask_file = img_file.replace('.tif', '_mask.tif')
            if mask_file in files:
                # Full paths
                img_path = os.path.join(case_dir, img_file)
                mask_path = os.path.join(case_dir, mask_file)
                # Read images
                img = cv2.imread(img_path)
                mask = cv2.imread(mask_path, cv2.IMREAD_GRAYSCALE)
                if img is not None and mask is not None:
                    # Convert BGR to RGB
                    img = cv2.cvtColor(img, cv2.COLOR BGR2RGB)
                    # Take only the FLAIR channel (channel 1) from the original image
                    # In LGG dataset, FLAIR is the 2nd channel (index 1)
                    flair = img[:, :, 1]
                    # Output filenames
                    out_img_name = f"brain_tumor_{processed_count:03d}.png"
                    out_mask_name = f"brain_tumor_{processed_count:03d}_mask.png"
                    # Save paths
                    out_img_path = os.path.join(output_dir, 'images', out_img_name)
                    out_mask_path = os.path.join(output_dir, 'masks', out_mask_name)
                    # Save images
                    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
# Function to visualize the prepared dataset
def visualize_dataset(dataset_path, num_samples=5):
    Visualize the prepared dataset
    Parameters:
    _____
    dataset_path : str
        Path to the prepared dataset
    num_samples : int
       Number of samples to visualize
    images_dir = os.path.join(dataset_path, 'images')
    masks_dir = os.path.join(dataset_path, 'masks')
    # Get all image files
    image_files = sorted(os.listdir(images_dir))
    # Limit to the requested number of samples
    image_files = image_files[:min(len(image_files), num_samples)]
    # Create figure
    plt.figure(figsize=(12, 4 * len(image_files)))
    for i, img_file in enumerate(image_files):
        # Get corresponding mask file - handle different naming conventions
        if img_file.replace('.png', '_mask.png') in os.listdir(masks_dir):
            mask_file = img_file.replace('.png', '_mask.png')
        else:
            # Try alternative mask naming if needed
```

```
mask_file = next((m for m in os.listdir(masks_dir) if m.startswith(img_file.split('.')[0])), None)
        if mask_file:
            # Read images
            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)
            # Display image
            plt.subplot(len(image_files), 2, i * 2 + 1)
            plt.imshow(img, cmap='gray')
            plt.title(f"MRI Image: {img_file}")
            plt.axis('off')
            # Display mask
            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()
# Execute dataset preparation
print("Preparing dataset...")
prepare_dataset(num_samples=10, random_selection=True)
print("Dataset preparation completed!")
# Visualize the prepared dataset
print("Visualizing prepared dataset...")
visualize_dataset(output_dir, num_samples=5)
# Print instructions for using the dataset with the segmentation code
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')")
Running locally or Drive already mounted
     Preparing dataset...
     FileNotFoundError
                                               Traceback (most recent call last)
     <ipython-input-5-5c06eaa0785b> in <cell line: 0>()
         160 # Execute dataset preparation
         161 print("Preparing dataset...")
     --> 162 prepare_dataset(num_samples=10, random_selection=True)
         163 print("Dataset preparation completed!")
         164
     <ipython-input-5-5c06eaa0785b> in prepare_dataset(num_samples, random_selection)
          37
          38
                 # Get all case directories
                 case_dirs = [d for d in os.listdir(lgg_dataset_path)
     ---> 39
          40
                             if os.path.isdir(os.path.join(lgg_dataset_path, d))]
     FileNotFoundError: [Errno 2] No such file or directory: '/content/drive/MyDrive/brain tumor/ddd'
 Next steps:
             Explain error
import os
import cv2
import numpy as np
```

```
import matplotlib.pyplot as plt
from google.colab import drive
from sklearn.model_selection import train_test_split
from sklearn.metrics import accuracy_score, jaccard_score, f1_score, precision_score, recall_score
# Mount Google Drive (for Google Colab)
try:
    drive.mount('/content/drive')
    print("Google Drive mounted successfully!")
except:
   print("Running locally or Drive already mounted")
class BrainTumorSegmentation:
    def __init__(self, base_dir=None):
        Initialize the Brain Tumor Segmentation class
        Parameters:
        ____
        base dir : str
            Base directory containing the dataset
        self.base dir = base dir
        self.images = []
        self.masks = []
        self.processed_images = []
        self.segmented_masks = []
        self.metrics = {}
    def load_dataset(self, images_dir, masks_dir, max_samples=None):
        Load dataset images and their corresponding masks
        Parameters:
        images_dir : str
            Directory containing MRI scan images
        masks_dir : str
            Directory containing corresponding mask images
        max_samples : int, optional
            Maximum number of samples to load
        print("Loading dataset...")
        # List all files in the directories
        image files = sorted(os.listdir(images dir))
        # For LGG dataset, find corresponding mask files
        loaded_count = 0
        for img_file in image_files:
            if not img_file.endswith(('.jpg', '.png', '.jpeg', '.tif')):
                continue
            # Construct mask filename based on LGG naming pattern
            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)
                # Check if mask file exists
                if not os.path.exists(mask_path):
                    print(f"Warning: No mask found for {img_file}")
                    continue
                # Read image and mask
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image = cv2.imread(image path)
            mask = cv2.imread(mask_path, cv2.IMREAD_GRAYSCALE)
            if image is not None and mask is not None:
                # Handle grayscale vs color images
                if len(image.shape) == 3 and image.shape[2] == 3:
                    # For LGG dataset: take FLAIR channel (usually the most informative)
                    # FLAIR is typically in the middle channel (index 1)
                    gray_image = image[:, :, 1] # Extract FLAIR channel
                else:
                    gray_image = image.copy()
                # Convert to single channel and normalize
                if len(gray_image.shape) == 3:
                    gray_image = cv2.cvtColor(gray_image, cv2.COLOR_BGR2GRAY)
                # Normalize mask to binary (0 and 255)
                _, binary_mask = cv2.threshold(mask, 127, 255, cv2.THRESH_BINARY)
                # Add to the dataset
                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):
    """Preprocess the loaded MRI images"""
    print("Preprocessing images...")
    self.processed_images = []
    for image in self.images:
        # Ensure image is grayscale
        if len(image.shape) > 2:
            gray_image = cv2.cvtColor(image, cv2.COLOR_RGB2GRAY)
        else:
           gray_image = image.copy()
        # Normalize pixel values to range [0, 255]
        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
        # Apply CLAHE (Contrast Limited Adaptive Histogram Equalization)
        clahe = cv2.createCLAHE(clipLimit=2.0, tileGridSize=(8, 8))
        enhanced = clahe.apply(normalized)
        # Apply Gaussian blur to reduce noise
        blurred = cv2.GaussianBlur(enhanced, (5, 5), 0)
        # Store preprocessed image
        self.processed_images.append(blurred)
    print(f"Preprocessed {len(self.processed_images)} images.")
def segment_tumors(self, method='watershed'):
    Segment tumor regions from preprocessed images
   Parameters:
    method: str
        Segmentation method to use ('threshold', 'watershed', 'kmeans')
```

```
print(f"Segmenting tumors using {method} method...")
self.segmented_masks = []
for image in self.processed_images:
   if method == 'threshold':
       # Apply Otsu's thresholding
       _, segmented = cv2.threshold(image, 0, 255, cv2.THRESH_BINARY + cv2.THRESH_OTSU)
   elif method == 'watershed':
       # Watershed algorithm
       # Find sure background
       _, thresholded = cv2.threshold(image, 0, 255, cv2.THRESH_BINARY + cv2.THRESH_OTSU)
       # Noise removal with morphological operations
       kernel = np.ones((3, 3), np.uint8)
       opening = cv2.morphologyEx(thresholded, cv2.MORPH OPEN, kernel, iterations=2)
       # Sure background area
       sure_bg = cv2.dilate(opening, kernel, iterations=3)
       # Finding sure foreground area
       dist transform = cv2.distanceTransform(opening, cv2.DIST L2, 5)
       _, sure_fg = cv2.threshold(dist_transform, 0.7*dist_transform.max(), 255, 0)
       # Finding unknown region
       sure_fg = np.uint8(sure_fg)
       unknown = cv2.subtract(sure_bg, sure_fg)
       # Marker labelling
       _, markers = cv2.connectedComponents(sure_fg)
       # Add one to all labels so that background is not 0, but 1
       markers = markers + 1
       # Mark the unknown region with 0
       markers[unknown == 255] = 0
       # Apply watershed
       markers = cv2.watershed(cv2.cvtColor(image, cv2.COLOR_GRAY2BGR), markers)
        segmented = np.zeros_like(image)
       segmented[markers > 1] = 255
   elif method == 'kmeans':
       # Apply K-means clustering
        image data = image.reshape((-1, 1))
       image data = np.float32(image data)
       # Define criteria and apply kmeans
       criteria = (cv2.TERM CRITERIA EPS + cv2.TERM CRITERIA MAX_ITER, 100, 0.2)
       k = 3 # Number of clusters
       _, labels, centers = cv2.kmeans(image_data, k, None, criteria, 10, cv2.KMEANS_RANDOM_CENTERS)
       # Find the cluster that corresponds to the tumor (usually the brightest)
       centers = np.uint8(centers)
       brightest_cluster = np.argmax(centers)
       # Create mask based on the brightest cluster
       segmented = np.zeros_like(image)
       segmented[labels.reshape(image.shape) == brightest_cluster] = 255
   else:
       raise ValueError(f"Unknown segmentation method: {method}")
   # Post-processing: fill holes and remove small objects
   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):
    Apply post-processing to improve the segmentation mask
   Parameters:
   mask : numpy.ndarray
       Binary segmentation mask
    Returns:
    numpy.ndarray
       Post-processed binary mask
    # Convert to binary
    if mask.dtype != np.uint8:
       mask = mask.astype(np.uint8)
    # Fill holes
    contours, = cv2.findContours(mask, cv2.RETR_EXTERNAL, cv2.CHAIN_APPROX_SIMPLE)
    filled mask = np.zeros like(mask)
    cv2.drawContours(filled mask, contours, -1, 255, -1)
    # Remove small objects (noise)
    nb_components, output, stats, _ = cv2.connectedComponentsWithStats(filled_mask, connectivity=8)
    sizes = stats[1:, -1]
    min_size = 100 # Minimum size of objects to keep
    # Keep only components with size greater than min_size
    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):
    Evaluate segmentation performance against ground truth masks
    ____
    ground truth masks : list
        List of ground truth mask images
    Returns:
    dict
       Dictionary containing evaluation metrics
    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")
    # Initialize metrics
    dice_scores = []
    jaccard_scores = []
    precision_scores = []
    recall_scores = []
    for gt_mask, pred_mask in zip(ground_truth_masks, self.segmented_masks):
        # Binarize masks
        gt_binary = np.where(gt_mask > 0, 1, 0).flatten()
        pred_binary = np.where(pred_mask > 0, 1, 0).flatten()
        # Calculate Dice coefficient (F1 score)
        dice = f1 score(gt_binary, pred_binary, zero_division=1)
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dice scores.append(dice)
        # Calculate Jaccard index (IoU)
        iou = jaccard_score(gt_binary, pred_binary, zero_division=1)
        jaccard_scores.append(iou)
        # Calculate precision and recall
        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)
   # Calculate average metrics
    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):
   Visualize original images, ground truth masks, and segmented masks
   Parameters:
    num_samples : int
       Number of samples to visualize
    num_samples = min(num_samples, len(self.images))
   plt.figure(figsize=(15, 4 * num_samples))
    for i in range(num samples):
        # Original image
        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')
        # Ground truth mask
        plt.subplot(num_samples, 3, i * 3 + 2)
        plt.imshow(self.masks[i], cmap='gray')
        plt.title(f"Ground Truth Mask {i+1}")
        plt.axis('off')
        # Segmented mask
        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):
    Overlay segmentation results on original images for visualization
   Parameters:
```

```
num_samples : int
      Number of samples to visualize
   num_samples = min(num_samples, len(self.images))
   plt.figure(figsize=(12, 4 * num_samples))
   for i in range(num_samples):
        # Convert image to RGB for overlay
        display_img = cv2.cvtColor(self.images[i], cv2.COLOR_GRAY2RGB)
        # Original image with ground truth overlay
        plt.subplot(num samples, 2, i * 2 + 1)
        # Create overlay with ground truth mask (green)
        overlay = display_img.copy()
        # Create green mask
        green_mask = np.zeros_like(overlay)
        green mask[:, :, 1] = self.masks[i] # Green channel
        # Apply mask
        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')
        # Original image with segmentation overlay
       plt.subplot(num_samples, 2, i * 2 + 2)
        # Create overlay with segmented mask (red)
        overlay = display_img.copy()
        # Create red mask
        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'):
    Run the full segmentation pipeline
   Parameters:
    -----
    images_dir : str
       Directory containing MRI scan images
   masks_dir : str
       Directory containing corresponding mask images
   max_samples : int, optional
       Maximum number of samples to load
    segmentation_method : str
       Method to use for segmentation ('threshold', 'watershed', 'kmeans')
    # Load dataset
    self.load_dataset(images_dir, masks_dir, max_samples)
   # Preprocess images
```

```
self.preprocess images()
        # Segment tumors
        self.segment_tumors(method=segmentation_method)
        # Evaluate segmentation
        self.evaluate_segmentation()
        # Visualize results
        self.visualize_results()
        self.overlay_results()
        return self.metrics
# Execute the segmentation pipeline
if __name__ == "__main__":
    # Update these paths to your LGG dataset location
    base_dir = '/content/drive/MyDrive/brain_tumor_dataset'
    images_dir = os.path.join(base_dir, 'images')
    masks_dir = os.path.join(base_dir, 'masks')
    # Create segmentation object
    tumor segmentation = BrainTumorSegmentation(base dir)
    # Run full pipeline with all available segmentation methods
    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=10, # Use 10 images as requested
            segmentation_method=method
        results[method] = metrics
    # Compare results
    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}")
```