

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
try:
    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/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]

        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

```

```
# 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')")
```



```

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
            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

    Parameters:
    -----
    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)
        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

        # Apply mask
        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:")
```



