MedBLIP Image Captioning for Radiology XRay Images

depi final project

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MedBLIP - Image Captioning for Radiology X-Ray Images

Overview:

This repository implements BLIP (Bootstrapping Language-Image Pre-training) to generate captions from Radiology X-ray images.

What is BLIP?

A powerful vision-language pre-training model designed for efficient handling of image-captioning tasks.

Focus of the Project:

The project specifically targets medical imaging, applying BLIP to the Radiology Objects in Context (ROCO) dataset.

Importance:

Enhances the accessibility of medical imaging by generating descriptive captions, aiding in better understanding and interpretation of X-ray images.

checking for missing values:

Explanation:

- This function checks for missing or invalid image files in the DataFrame.
- It iterates through the image paths, verifying if each file exists and can be opened.
- If invalid files are found, those rows are removed from the DataFrame to ensure data integrity.

3. Checking for Missing or Invalid Files 口 Copy code def check and remove missing or invalid files(df, column name): missing_or_invalid_files = [] for idx, image_path in df[column_name].items(): if not os.path.isfile(image_path): # Check if file exists missing or invalid files.append(idx) img = Image.open(image_path) img.verify() # Verify that it is, indeed, an image except Exception as e: print(f"Error with file {image path}: {e}") missing_or_invalid_files.append(idx) if missing_or_invalid_files: print(f"Removing {len(missing or invalid files)} rows with missing or in df = df.drop(missing_or_invalid_files).reset_index(drop=True) print("All files are present and valid!") return df

Dataset Class creation:

Explanation:

- The ImageCaptioningDataset class inherits from Dataset, allowing for a customized dataset structure.
- The __init__ method initializes the dataset, processor, and image size.
- The __len__ method returns the total number of samples in the dataset.
- The __getitem__ method processes the image and caption, preparing them for input to the model.

4. Dataset Class Creation

```
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class ImageCaptioningDataset(Dataset): # Custom Dataset class for image caption
   def init (self, dataset, processor, image size=(224, 224)):
       self.dataset = dataset
       self.processor = processor
       self.image size = image size
       self.resize_transform = Resize(image_size)
   def __len__(self):
       return len(self.dataset) # Returns the number of samples in the dataset
   def __getitem__(self, idx):
       item = self.dataset.iloc[idx] # Get the sample at index idx
       img = Image.open(item['images']) # Open the image file
       encoding = self.processor(images=img, text=item["caption"], padding="max
       # Process the image and text
       encoding = {k: v.squeeze() for k, v in encoding.items()} # Adjust encod
       return encoding # Return processed data
```

Checkpoint Management **Functions:**

Explanation:

- The load_checkpoint function checks if a checkpoint file exists.
- If it does, the model and optimizer states are restored, allowing training to resume from where it left off.
- If not, it initializes the training from the beginning, setting the starting epoch to zero.
- The save_checkpoint function creates a directory for saving checkpoints if it doesn't exist.
- It saves the current epoch, model state, optimizer state, and average loss to a file.
- This function ensures that progress can be restored later without losing any training data.

11. Checkpoint Management Functions

Load Checkpoint Function

```
Copy code
 def load_checkpoint(model, optimizer, checkpoint_path):
     if os.path.exists(checkpoint_path):
         checkpoint = torch.load(checkpoint_path) # Load the checkpoint file
         model.load_state_dict(checkpoint['model_state_dict']) # Load the model
         optimizer.load_state_dict(checkpoint['optimizer_state_dict']) # Load th
         start_epoch = checkpoint['epoch'] # Get the starting epoch from the che
         print(f"Loaded checkpoint from {checkpoint_path} at epoch {start_epoch}'
         start_epoch = 0 # If no checkpoint, start from epoch 0
         print(f"No checkpoint found at {checkpoint path}. Starting from scratch.
     return start_epoch
Save Checkpoint Function
```

```
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def save_checkpoint(model, optimizer, epoch, avg_loss, checkpoint_path='checkpoi
   if not os.path.exists('checkpoints'):
       os.makedirs('checkpoints') # Create checkpoints directory if it doesn't
   torch.save({
       'epoch': epoch,
       'model_state_dict': model.state_dict(),  # Save the model state
        'optimizer state dict': optimizer.state dict(), # Save the optimizer st
        'avg_loss': avg_loss, # Save the average loss
   }, checkpoint path) # Save checkpoint file
   print(f"Checkpoint saved at epoch {epoch} with average loss: {avg_loss:.4f}"
```

Loss Calculation and Backpropagation:

Explanation:

- The loss calculated from the model's output is used for backpropagation.
- The optimizer clears previous gradients, computes the new gradients based on the current loss, and updates the model parameters accordingly.

```
9. Loss Calculation and Backpropagation

python

loss = outputs.loss # Get loss

# Backpropagation

optimizer.zero_grad() # Clear previous gradients
loss.backward() # Backpropagate the loss

optimizer.step() # Update model parameters
```

BLEU Score Calculation Function:

Explanation:

- The calculate_bleu_score function computes the BLEU score, a metric for evaluating the quality of generated text.
- It splits both reference captions and generated captions into words before calculating the score.
- This function can be used to assess the quality of captions generated by the model.

python def calculate_bleu_score(references, hypothesis): references = [ref.split() for ref in references] # Split reference sentence hypothesis = hypothesis.split() # Split hypothesis sentence into words score = sentence_bleu(references, hypothesis) # Calculate BLEU score return score # Return the BLEU score