```
%cd /content/drive/MyDrive/ImageBind finetune/ImageBind-LoRA
/content/drive/MyDrive/ImageBind finetune/ImageBind-LoRA
!pip install -r requirements.txt --quiet
\rightarrow
       Preparing metadata (setup.py) ... done
                                                    890.1/890.1 MB 1.5 MB/s eta 0:00:00
                                                    24.3/24.3 MB 64.8 MB/s eta 0:00:00
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       Preparing metadata (setup.py) ... done
                                                   - 7.1/7.1 MB 118.5 MB/s eta 0:00:00
       Installing build dependencies ... done
       Getting requirements to build wheel ... done
       Preparing metadata (pyproject.toml) \dots done
                                                    11.6/11.6 MB 118.3 MB/s eta 0:00:00
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                                                    12.4/12.4 MB 106.3 MB/s eta 0:00:00
                                                   - 82.7/82.7 kB 12.4 MB/s eta 0:00:00
       Building wheel for pytorchvideo (setup.py) ... done
       Building wheel for fvcore (setup.py) ... done
       Building wheel for iopath (setup.py) ... done
       Building wheel for mayavi (pyproject.toml) ... done
     ERROR: pip's dependency resolver does not currently take into account all the pac
     torchtext 0.18.0 requires torch>=2.3.0, but you have torch 1.13.0 which is incomp
import os
current_directory = os.getcwd()
```

Resources X ···

You are not subscribed. Learn more Available: 49.34 compute units Usage rate: approximately 4.82 per hour You have 1 active session.

Manage sessions

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Not connected to runtime.

Change runtime type

```
import torch
import logging
import data
from models import imagebind_model
from models.imagebind_model import ModalityType, load_module
from models import lora as LoRA
logging.basicConfig(level=logging.INFO, force=True)
lora = False
linear_probing = False
device = "cuda:0" if torch.cuda.is_available() else "cpu"
load_head_post_proc_finetuned = True
assert not (linear_probing and lora), \
   "Linear probing is a subset of LoRA training procedure for ImageBind. " \
   "Cannot set both linear_probing=True and lora=True. "
if lora and not load_head_post_proc_finetuned:
   lora_factor = 12 / 0.07
else.
   lora_factor = 1
/usr/local/lib/python3.10/dist-packages/torchvision/transforms/_functional_video.
     /usr/local/lib/python3.10/dist-packages/torchvision/transforms/_transforms_video.
      warnings.warn(
image_paths = ["test_data/00088_2.jpg", "test_data/00095.jpg", "test_data/00158_4.jpg"
for i in range(len(image_paths)):
   image_paths[i] = current_directory + "/" + image_paths[i]
for i in range(len(audio_paths)):
   audio_paths[i] = current_directory + "/" + audio_paths[i]
model = imagebind_model.imagebind_huge(pretrained=True)
# if lora:
#
     model.modality_trunks.update(
         LoRA.apply_lora_modality_trunks(model.modality_trunks, rank=4,
#
                                        modality_names=[ModalityType.TEXT, Modality
     LoRA.load_lora_modality_trunks(model.modality_trunks,
                                   checkpoint_dir=".checkpoints/lora/", postfix="_l
#
     if load_head_post_proc_finetuned:
         load_module(model.modality_postprocessors, module_name="postprocessors",
                     checkpoint_dir=".checkpoints/lora/", postfix="_last")
         load module(model.modality heads, module name="heads",
                     checkpoint_dir=".checkpoints/lora/", postfix="_last")
#
  elif linear_probing:
     load_module(model.modality_heads, module_name="heads",
                 checkpoint_dir="./.checkpoints/lora/", postfix="_last")
model.eval()
model.to(device)
₹
```

```
(mith). Lith(
            (fc1): Linear(in_features=768, out_features=3072, bias=True)
            (act): GELU(approximate='none')
            (fc2): Linear(in_features=3072, out_features=768, bias=True)
            (drop): Dropout(p=0.0, inplace=False)
          (norm_2): LayerNorm((768,), eps=1e-06, elementwise_affine=True)
        (6): BlockWithMasking(
          (attn): MultiheadAttention(
            (out_proj): NonDynamicallyQuantizableLinear(in_features=768,
out_features=768, bias=True)
          (drop_path): DropPath(drop_prob=0.055)
          (norm_1): LayerNorm((768,), eps=1e-06, elementwise_affine=True)
          (mlp): Mlp(
            (fc1): Linear(in_features=768, out_features=3072, bias=True)
            (act): GELU(approximate='none')
            (fc2): Linear(in_features=3072, out_features=768, bias=True)
            (drop): Dropout(p=0.0, inplace=False)
          (norm_2): LayerNorm((768,), eps=1e-06, elementwise_affine=True)
        (7): BlockWithMasking(
          (attn): MultiheadAttention(
            (out_proj): NonDynamicallyQuantizableLinear(in_features=768,
out_features=768, bias=True)
          (drop_path): DropPath(drop_prob=0.064)
          (norm_1): LayerNorm((768,), eps=1e-06, elementwise_affine=True)
          (mlp): Mlp(
            (fc1): Linear(in_features=768, out_features=3072, bias=True)
            (act): GELU(approximate='none')
            (fc2): Linear(in_features=3072, out_features=768, bias=True)
            (drop): Dropout(p=0.0, inplace=False)
          (norm_2): LayerNorm((768,), eps=1e-06, elementwise_affine=True)
        (8): BlockWithMasking(
          (attn): MultiheadAttention(
            (out_proj): NonDynamicallyQuantizableLinear(in_features=768,
out_features=768, bias=True)
          (drop_path): DropPath(drop_prob=0.073)
          (norm_1): LayerNorm((768,), eps=1e-06, elementwise_affine=True)
          (mlp): Mlp(
```

```
# Load data
#audio_query_paths = [current_directory + "/test_data/00222_3.wav"] # Example: Choos
inputs_audio = {
   ModalityType.AUDIO: data.load_and_transform_audio_data(audio_paths, device),
inputs_image = {
   ModalityType.VISION: data.load_and_transform_vision_data(image_paths, device),
# Generate embeddings for audio query
with torch.no_grad():
   embeddings_audio = model(inputs_audio)
with torch.no_grad():
   embeddings_image = model(inputs_image)
vision_embeddings = embeddings_image[ModalityType.VISION]
audio_query_embedding = embeddings_audio[ModalityType.AUDIO]
count crt=0
for audio_index, audio_query_embedding in enumerate(embeddings_audio['audio']):
  similarity_scores = torch.softmax(vision_embeddings @ audio_query_embedding.T * (lo
  most_relevant_image_index = torch.argmax(similarity_scores)
  most_relevant_image_path = image_paths[most_relevant_image_index]
  if most_relevant_image_index==audio_index:
   count_crt+=1
  print("Audio query:",audio_paths[audio_index].split('/')[-1] )
  print("Most relevant image for audio query:", most_relevant_image_path.split('/')[-
Audio query: 00088_2.wav
     Most relevant image for audio query: 00222_3.jpg
     Audio query: 00095.wav
     Most relevant image for audio query: 00095.jpg
     Audio query: 00158_4.wav
     Most relevant image for audio query: 00169_3.jpg
     Audio query: 00169_3.wav
     Most relevant image for audio query: 00222_3.jpg
     Audio query: 00220.wav
     Most relevant image for audio query: 00220.jpg
     Audio query: 00222 3.wav
     Most relevant image for audio query: 00220.jpg
count_crt
→ 2
for index, audio_query_emb in enumerate(embeddings_audio['audio']):
  print(index, audio_query_emb)

    0 tensor([ 0.0478, -0.2131, 0.2315, ..., 0.5279, -0.7604, -0.0132],

            device='cuda:0')
     1 tensor([ 0.2381, -0.0772, -0.2055, ..., -0.4301, 0.3229, -0.4043],
            device='cuda:0')
     2 tensor([ 0.0859, -0.0992, 0.6365, ..., -0.1347, -0.8802, 0.0288],
            device='cuda:0')
     3 tensor([-0.3175, -0.6430, -0.4416, ..., 0.0071, 0.6742, 0.1453],
            device='cuda:0')
     4 tensor([ 0.5646, -0.1296, -0.1822, ..., -0.1235, 0.5655, 0.2960],
            device='cuda:0')
     5 tensor([-0.2691, -0.3681, -0.1953, ..., -0.4585, 0.3020, 0.2901],
            device='cuda:0')
vision_embeddings.shape
→ torch.Size([6, 1024])
vision_embeddings
```