```
import torchvision.models as models
import torchvision.transforms as transforms
from PIL import Image
from sklearn.metrics.pairwise import cosine_similarity
import pandas as pd
import numpy as np
import os
from google.colab import drive

# Google Drive 마운트 및 CSV 파일 로드
from google.colab import drive
drive.mount('/content/drive')
csv_path = '/content/drive/MyDrive/final/survey_style_preferences.csv'
df = pd.read_csv(csv_path)
```

→ Mounted at /content/drive

```
# ResNet-18 모델 로드 (중간 레이어 feature vector 추출용)
class FeatureExtractor(torch.nn.Module):
    def __init__(self, model):
        super(FeatureExtractor, self).__init__()
        self.features = torch.nn.Sequential(*list(model.children())[:-1]) # avgpool까지만 사용

def forward(self, x):
    return self.features(x).view(x.size(0), -1) # Flatten features

# 모델 설정
device = torch.device("cuda" if torch.cuda.is_available() else "cpu")
resnet18 = models.resnet18(pretrained=False) # 사전학습 없이
model = FeatureExtractor(resnet18).to(device)
model.eval()
```

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- /usr/local/lib/python3.10/dist-packages/torchvision/models/_utils.py:208: UserWarning: The parameter 'pretrained' is deprecated since 0.13 and ma
      warnings.warn(
     /usr/local/lib/python3.10/dist-packages/torchyision/models/ utils.py:223: UserWarning: Arguments other than a weight enum or `None` for 'weights
      warnings.warn(msg)
    FeatureExtractor(
       (features): Sequential(
         (0): Conv2d(3, 64, kernel_size=(7, 7), stride=(2, 2), padding=(3, 3), bias=False)
         (1): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
         (2): ReLU(inplace=True)
         (3): MaxPool2d(kernel_size=3, stride=2, padding=1, dilation=1, ceil_mode=False)
         (4): Sequential(
           (0): BasicBlock(
             (conv1): Conv2d(64, 64, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False)
             (bn1): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
             (relu): ReLU(inplace=True)
             (conv2): Conv2d(64, 64, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False)
             (bn2): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
           (1): BasicBlock(
             (conv1): Conv2d(64, 64, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False)
             (bn1): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
             (relu): ReLU(inplace=True)
             (conv2): Conv2d(64, 64, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False)
             (bn2): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
         (5): Sequential(
           (0): BasicBlock(
             (conv1): Conv2d(64, 128, kernel_size=(3, 3), stride=(2, 2), padding=(1, 1), bias=False)
             (bn1): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
             (relu): ReLU(inplace=True)
             (conv2): Conv2d(128, 128, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False)
             (bn2): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
             (downsample): Sequential(
               (0): Conv2d(64, 128, kernel_size=(1, 1), stride=(2, 2), bias=False)
               (1): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
           (1): BasicBlock(
             (conv1): Conv2d(128, 128, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False)
             (bn1): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
             (relu): ReLU(inplace=True)
             (conv2): Conv2d(128, 128, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False)
             (bn2): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
         (6): Sequential(
           (0): BasicBlock(
             (conv1): Conv2d(128, 256, kernel_size=(3, 3), stride=(2, 2), padding=(1, 1), bias=False)
```

(relu): ReLU(inplace=True)

(bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)

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(bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
             (downsample): Sequential(
               (0): Conv2d(128, 256, kernel_size=(1, 1), stride=(2, 2), bias=False)
               (1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
# 이미지 전처리 정의
transform = transforms.Compose([
    transforms.Resize((224, 224)),
    transforms.ToTensor().
    transforms.Normalize(mean=[0.485, 0.456, 0.406], std=[0.229, 0.224, 0.225]),
])
# Validation 이미지 폴더 내 모든 이미지 feature vector 추출
validation_folder = '/content/drive/MyDrive/final/validation_image'
validation_features = {}
for img_name in os.listdir(validation_folder):
    img_path = os.path.join(validation_folder, img_name)
    image = Image.open(img_path).convert('RGB')
    image = transform(image).unsqueeze(0).to(device)
    with torch.no_grad():
        feature = model(image).cpu().numpy()
   validation_features[img_name] = feature # 이미지 이름을 key로 저장
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(conv2): Conv2d(256, 256, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False)

/usr/local/lib/python3.10/dist-packages/PIL/Image.py in copy(self)
1272 """
1273 self.load()
-> 1274 return self._new(self.im.copy())
1275
1276 __copy__ = copy

KeyboardInterrupt:

```
# 유사도를 기반으로 Validation 이미지 선호 여부 예측
threshold = 0.8 # 임계값 설정 (예: 코사인 유사도 0.8 이상이면 선호로 예측)
for _, row in df.iterrows():
   user_id = row['응답자 ID']
   # 응답자의 선호/비선호 이미지 feature 벡터 준비
   liked vectors = []
   disliked_vectors = []
   if pd.notna(row['Validation 스타일 선호']):
       liked_images = row['Validation 스타일 선호'].split(" | ")
       liked_vectors = [validation_features[img] for img in liked_images if img in validation_features]
   if pd.notna(row['Validation 스타일 비선호']):
       disliked_images = row['Validation 스타일 비선호'].split(" | ")
       disliked_vectors = [validation_features[img] for img in disliked_images if img in validation_features]
   # 모든 Validation 이미지에 대해 응답자의 선호 여부 예측
   for val_img_name, val_feature in validation_features.items():
       val_feature = np.array(val_feature).reshape(1, -1)
       # 선호 이미지와의 유사도 계산
       liked_similarities = [cosine_similarity(val_feature, liked_vector)[0][0] for liked_vector in liked_vectors]
       \label{liked_similarities} disliked\_similarity (val\_feature, \ disliked\_vector) [0] [0] \ for \ disliked\_vector \ in \ disliked\_vectors]
       # 유사도 평균을 사용하여 예측 결정
       avg_liked_similarity = np.mean(liked_similarities) if liked_similarities else 0
       avg_disliked_similarity = np.mean(disliked_similarities) if disliked_similarities else 0
       # 임계값 비교를 통해 선호 여부 예측
       prediction = '선호' if avg_liked_similarity >= threshold and avg_liked_similarity > avg_disliked_similarity else '비선호'
       # 결과 저장
```

```
results.append({
            '응답자 ID': user_id,
           'Validation 이미지': val_img_name,
           '예측 선호 여부': prediction,
           '선호 유사도 평균': avg_liked_similarity,
           '비선호 유사도 평균': avg_disliked_similarity
results
# 결과를 DataFrame으로 변환하여 확인
results_df = pd.DataFrame(results)
print(results_df)
# 결과를 CSV 파일로 저장
output_path = '/content/drive/MyDrive/final/3-2 prediction_results.csv'
results_df.to_csv(output_path, index=False, encoding='utf-8-sig')
print(f"Prediction results saved to {output_path}")
\rightarrow \overline{\phantom{a}}
     NameError
                                              Traceback (most recent call last)
     <ipython-input-12-e6f853770614> in <cell line: 3>()
           1 # 결과를 CSV 파일로 저장
           2 output_path = '/content/drive/MyDrive/Colab Notebooks/지은/3-2 prediction_results.csv'
     ----> 3 results_df.to_csv(output_path, index=False, encoding='utf-8-sig')
           5 print(f"Prediction results saved to {output_path}")
     NameError: name 'results_df' is not defined
```

선호도 데이터를 바탕으로 예측 정확도 측정

```
import pandas as pd
preference_predict = pd.read_csv('/content/drive/MyDrive/final/3-2 prediction_results.csv')
preference_true = pd.read_csv('/content/drive/MyDrive/final/스타일선호도.csv')
num_accurate_predict = 0 # 정확하게 예측한 data 수
for _, row in preference_predict.iterrows():
   user_id = row['응답자 ID']
   val_img = row['Validation 이미지']
   print(user_id)
   if row['예측 선호 여부'] == '선호':
       if val_img in preference_true[preference_true['응답자ID'] == user_id]['Validation 스타일 선호'].values:
           num_accurate_predict += 1
   else:
       if val_img in preference_true[preference_true['응답자ID'] == user_id]['Validation 스타일 비선호'].values:
           num_accurate_predict += 1
accuracy = num_accurate_predict / len(preference_predict) * 100
print(f"Accuracy: {accuracy}%")
```

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