Deep Learning

Exercise 7: Transfer Learning

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Friday, April 23, 2020

Outline

- Pre-trained Networks
- Pace Recognition

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- Pre-trained Networks
 - Obtaining Pre-Trained Networks
 - Turn Networks to Deep Feature Extractors

Obtaining Pre-Trained Networks

Pre-Defined Networks

- Networks implemented in torchvision.models
 - → See http://pytorch.org/vision/stable/models.html
- Create pre-defined network, e.g.: network = torchvision.models.resnet18()
- Download pre-trained (on ImageNet) network: network = torchvision.models.resnet18(pretrained=True)

Input Normalization

- ullet RGB color images in resolution $3 \times 244 \times 244$ with pixels [0,1]
- Normalized by subtracting mean (0.485, 0.456, 0.406) and dividing by variance (0.229, 0.224, 0.225)

Turn Networks to Deep Feature Extractors

Problem

- Pre-trained networks return logits
- We only get object of ResNet class
- Need to overwrite forward

Redefining to the last

• Rewrite function:

```
def my_forward(self, x):
```

• Bind my_forward to object:

ResNet (Implementation)

```
def _forward_impl(self, x):
   x = self.conv1(x)
    x = self.bn1(x)
   x = self.relu(x)
    x = self.maxpool(x)
    x = self.layer1(x)
    x = self.laver2(x)
   x = self.layer3(x)
   x = self.layer4(x)
   x = self.avgpool(x)
    x = torch.flatten(x, 1)
   ## x = self.fc(x)
   return x
def forward(self, x):
   return self._forward_impl(x)
```

Outline

- Pace Recognition
 - Dataset
 - Face Identification

Dataset

Face Recognition Dataset

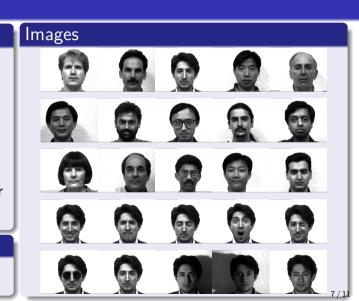
- Images of 15 people
- Grayscale images as .gif
- File: subjectXX.situation
- 11 different variations:

happy, sad, sleepy, surprised, wink glasses, noglasses, left, right, center

• Load: PIL.Image.open(...)

Download Face Data

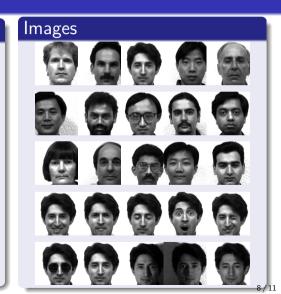
http://vision.ucsd.edu/content/ vale-face-database



Dataset

Image Preprocessing

- Scale image to 300 pixel height: torchyision.transforms.Resize(300)
- Take 224 × 224 center crop: torchvision.transforms.CenterCrop(224)
- Onvert to Tensor: torchvision.transforms.ToTensor()
- Convert to 3-plane image: tensor.repeat(3,1,1)
- Normalize (see above)
- Add batchsize dimension: tensor.unsqueeze(0)



Face Identification

Gallery

- Extract features for normal images
- Associate features with subject
 - \Rightarrow 15 features in gallery

Exemplary Results

normal	happy	sad	sleepy
15/15	14/15	15/15	15/15
surprised	wink	glasses	noglasses
13/15	15/15	15/15	13/15
leftlight	rightlight	centerlight	total
12/15	12/15	11/14	150/164

Probing

- Go through all variations
- Go through all subjects
- Extract feature for image
- Compare to all gallery features
 - → Compute gallery subject with lowest distance to probe
- Compare best gallery subject with probe subject
- Count correct identifications

Face Identification

Task 1: Pre-Trained Network

- Download pre-trained network via torchvision
 - \rightarrow Try out versions (18, 32, ...)
- Turn network into deep feature extractor

Task 2: Feature Extraction

- Write a function that:
 - Loads image via PIL
 - Crops the face
 - Transforms image to tensor
 - Extracts the deep feature
 - Returns flattened version as a numpy array

Face Identification

Task 3: Build Gallery

Store features for normal images

Task 4: Probing

- Take all images of all subjects
 - → Split by variation
- Extract features for image
- Compare to all gallery features
 - \rightarrow scipy.spatial.distance.euclidean
- Count correctly classified faces
 - → Split by variation
 - \rightarrow Compute total

Task 5: Variations (Optional)

- Different network topologies
- Other distance functions
- Different preprocessing
 - → Face detection?
- Raw pixel values of cropped faces as features
- Different variation in gallery
 - \rightarrow happy, glasses, rightlight