



Exercises

Advanced Machine Learning

Teaching Assistant:

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Overview

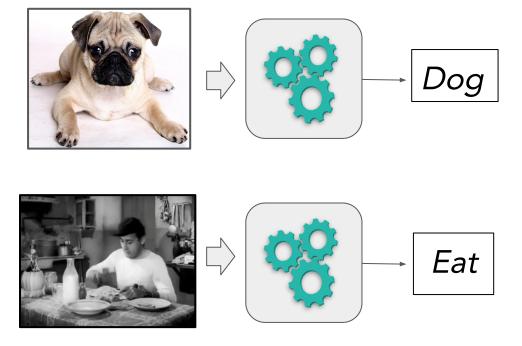
1. Train a Convolutional LSTM for First Person Action Recognition (FPAR):

Network: ResNet 34 + RNN

Videos: GTEA61

2. Exercise **Steps**:

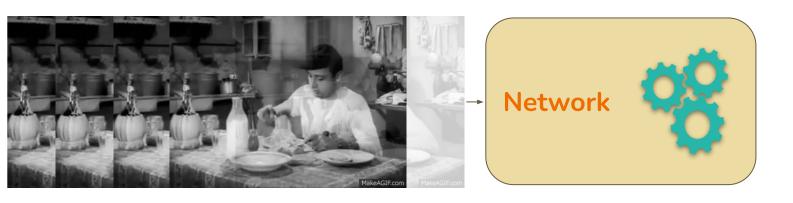
- Learning without Temporal Information (Avgpool)
- Learning with Temporal Information (LSTM)
- Learning with Spatio-Temporal Information (ConvLSTM)

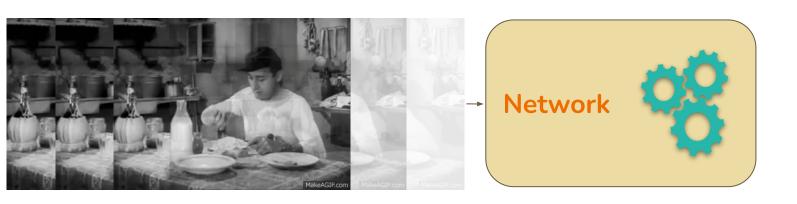


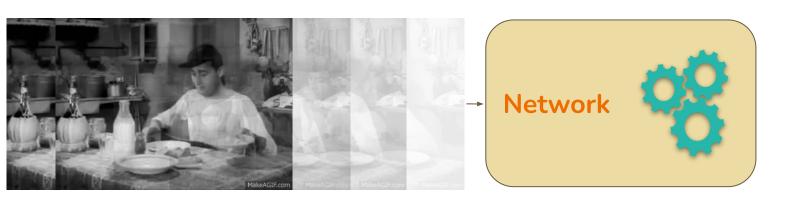


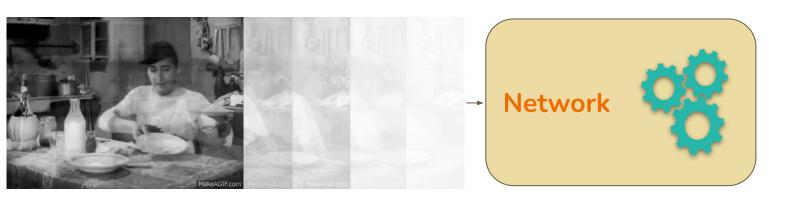




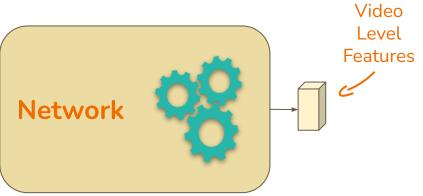














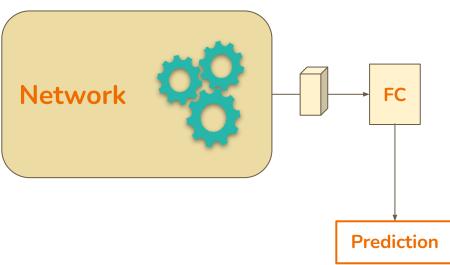


Image vs Frames

Image vs Frames

Don't worry! Already Implemented!!!

Training

Random Crop



Random Crop + Random Sampling

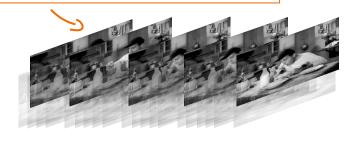


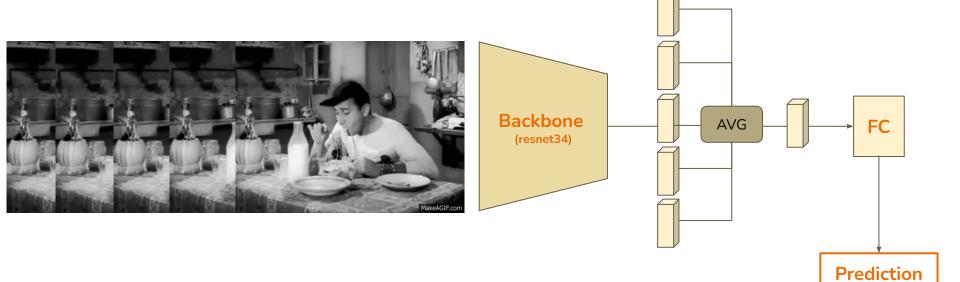
Test

Center Crop

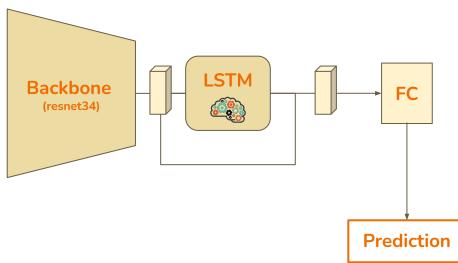


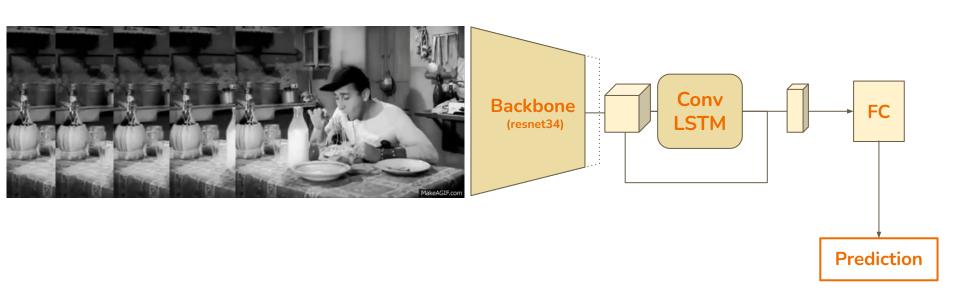
Center Crop + Uniform Sampling











Dataset: GTEA61

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- 61 actions
- 4 users:
 - o S1, S2, S3, S4

Training sets = S1 S3 S4 (labeled)

Validation set = Test set = S2.







Dataset:

https://drive.google.com/drive/folders/1_NAcoR0UGH1eLsiWMOx_Py8yeAocknA2?usp=sharing

Code templates

The template of the main code is available here:



Step 0: Before you start

Before you start

1. Study code and data:

- a. Read carefully the template code (including the comments) to understand how everything is done.
- b. Explore the data provided.

Run the code:

- a. try to run the code, "Learning without Temporal Information" is already implemented
- b. you have to stay connected

Step 1: LSTM

Step 1: LSTM

- 1. Set the variable homework step = 1 in MAIN PARAMS
- 2. Implement the **LSTM** in the class MyLSTM, section Model
 - a. you should implement the model in the init function and the forward
- 3. Train only the **LSTM** and the **Classifier** (maintaining freezed the backbone)
 - a. follow the same procedure used in Build Model Loss Opt and in Training to
 train only the classifier in the Step 0

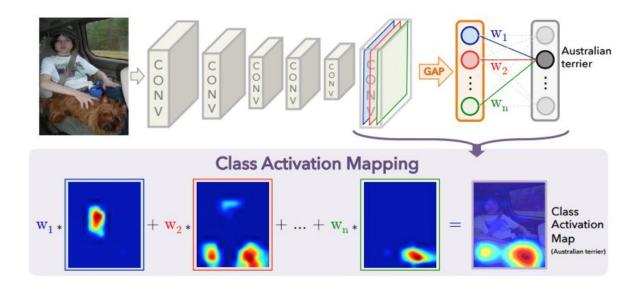
Step 2: ConvLSTM

Step 2: ConvLSTM

- 1. Set the variable homework step = 2 in MAIN PARAMS
- 2. Implement the ConvLSTM in the class MyConvLSTM, section Mode1
 - a. it is very similar to the standard LSTM with the difference that it use the convolution operation instead of nn.linear and it works on spatial_frames_feat instead of frames_feat
- 3. Feed into the ConvLSTM the features before the avgpool of the resnet34
 - a. named spatial frame feat
- 4. Load the pretrained-weights that you find in "/content/best model state dict rgb split2.pth"
- 5. Train <u>only</u> the **ConvLSTM** and the **Classifier**, maintaining freezed the backbone

Step 4: Class Activation Map (CAM)

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Bolei Zhou, et al. "Learning Deep Features for Discriminative Localization"

Step 4: Class Activation Map (CAM)

```
logit, feature_conv, _ = self.resNet(inputVariable[t])
bz, nc, h, w = feature_conv.size()

feature_conv1 = feature_conv.view(bz, nc, h*w)

probs, idxs = logit.sort(1, True)

class_idx = idxs[:, 0]

cam = torch.bmm(self.weight_softmax[class_idx].unsqueeze(1), feature_conv1)
```

```
cam img = F.softmax(cam, 1).data
cam img = cam img.cpu().numpy()
cam img = cam img.reshape(h, w)
cam img = cam img - np.min(cam img)
cam img = cam img / np.max(cam img)
cam img = np.uint8(255 * cam img)
output cam = cv2.resize(cam img, size upsample)
img = cv2.cvtColor(np.uint8(img), cv2.COLOR RGB2BGR)
heatmap = cv2.applyColorMap(output cam, cv2.COLORMAP JET)
result = heatmap * 0.3 + img * 0.5
```

NOW IT'S YOUR TURN, TRY!



Special thanks to Mirco Planamente for sharing the material