

DLCV HW5

2018/05/15

Reminder

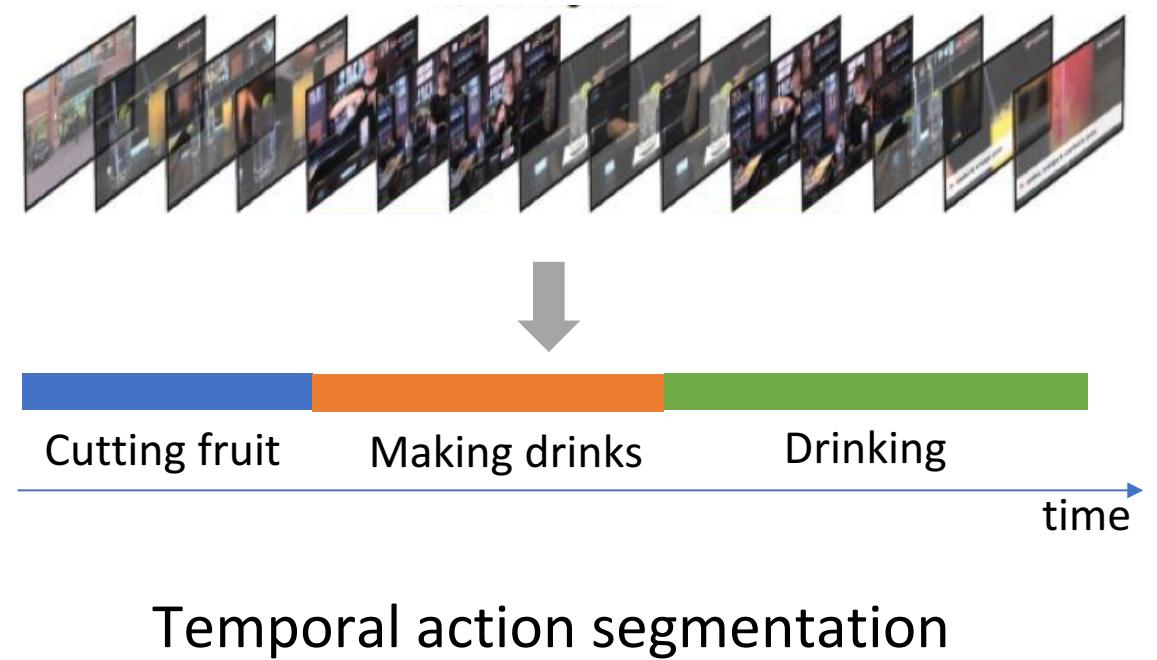
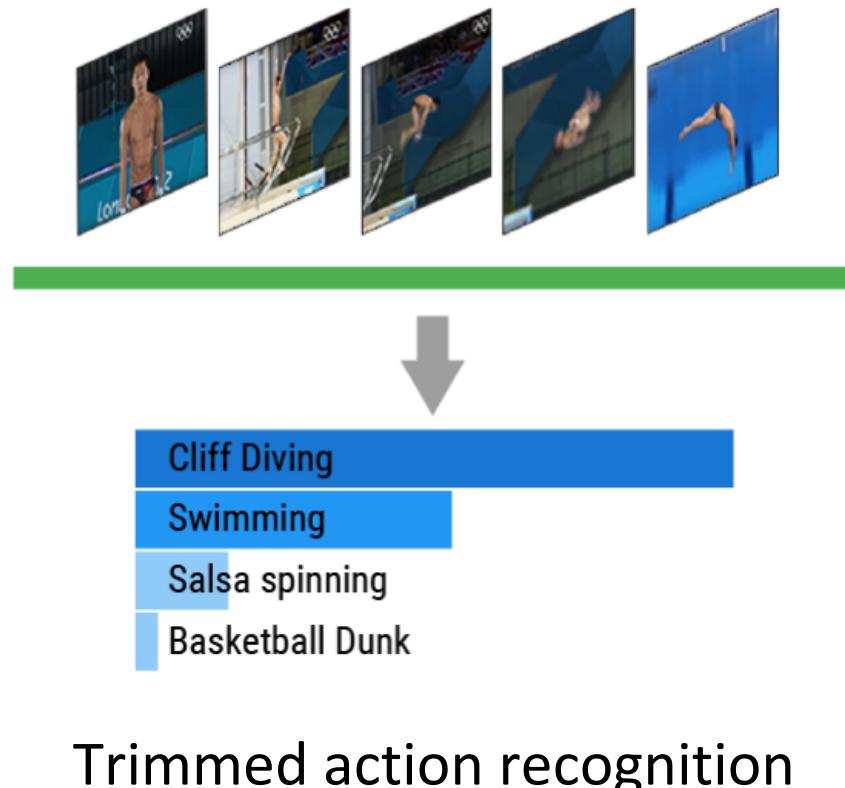
- Last Homework! 
- 3.5 weeks long
- 5/25 停修截止

Goal

- Ability to extract state-of-the-art deep CNN features
- Implement recurrent neural networks (RNN) for action recognition
- Extend RNN models for solving sequence-to-sequence problems

Task Description

- In this assignment, you will learn to perform both **trimmed action recognition** and **temporal action segmentation** in full-length videos.

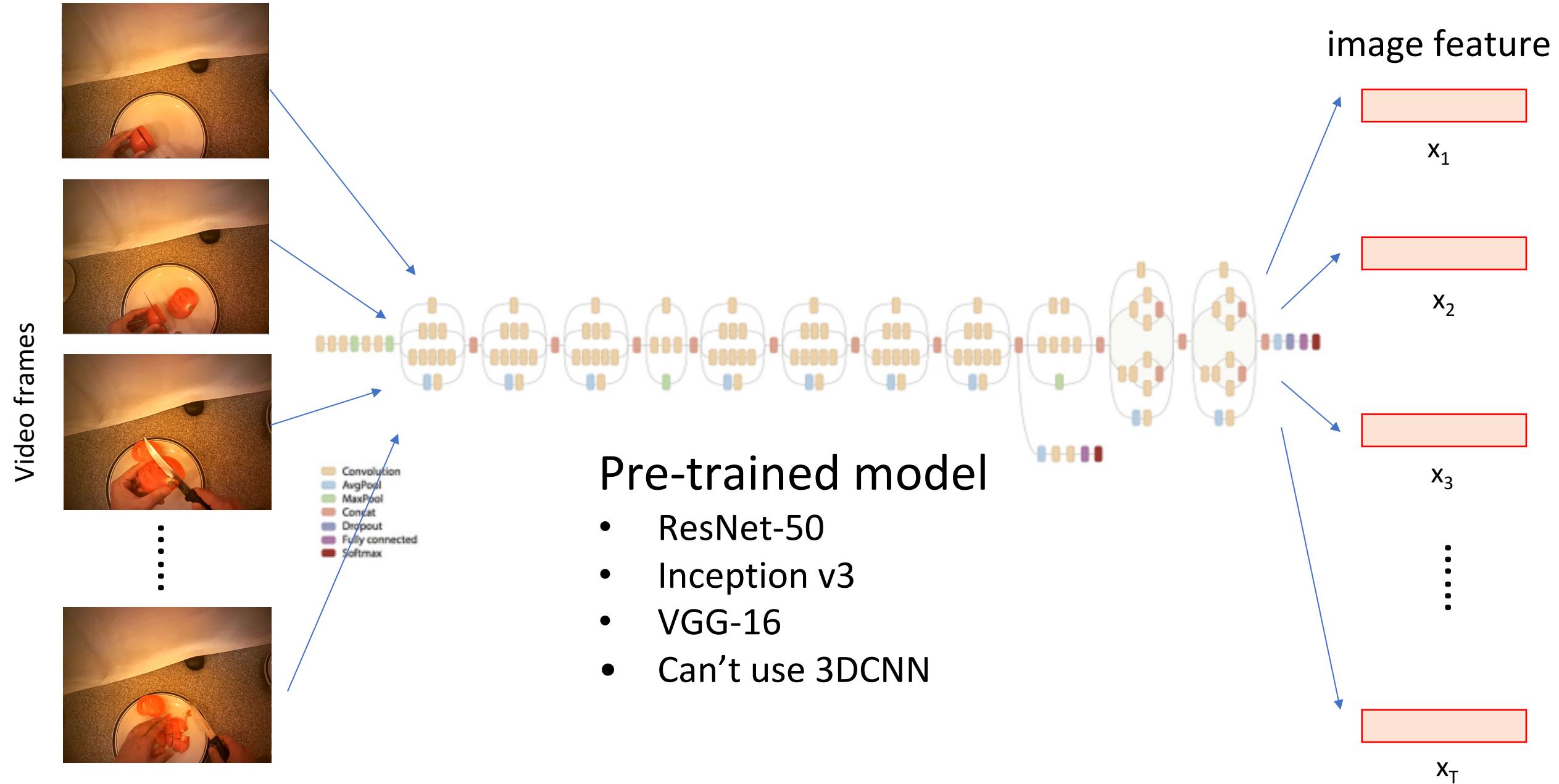


Task Description

- Task 1 : Data preprocessing
 - Extract state-of-the-art CNN features for action recognition
- Task 2 : Trimmed action recognition
 - Training your RNN model with sequences of CNN features and labels
- Task 3 : Temporal action segmentation
 - Extend your RNN model for sequence-to-sequence prediction

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Video frames



- Convolution
- AvgPool
- MaxPool
- Concat
- Dropout
- Fully connected
- Softmax

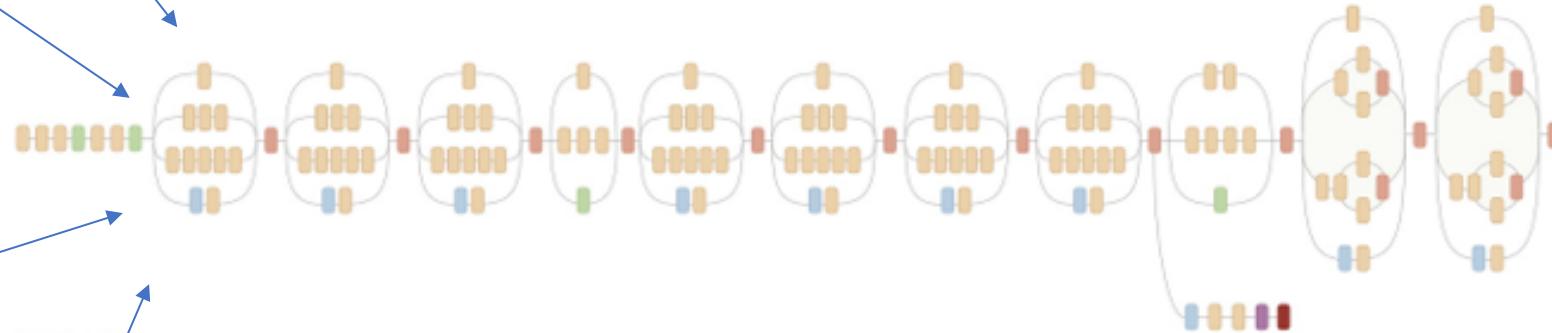
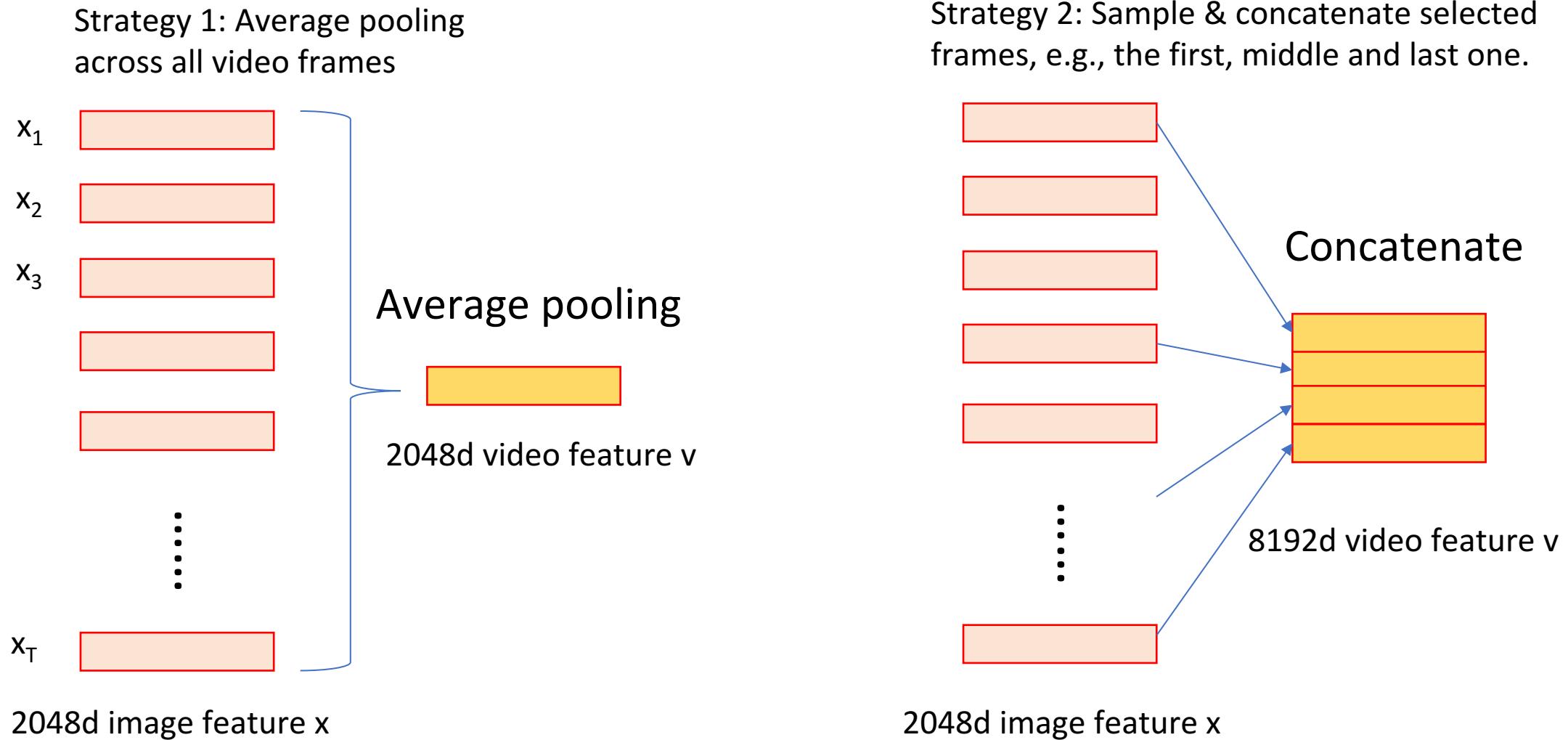


image feature

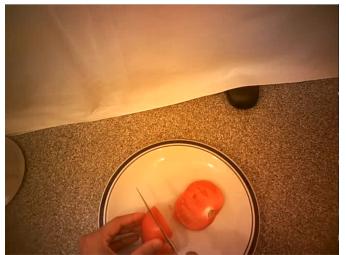
Note that you should down-sample your
video (e.g., to 2fps) to reduce computations

CNN-based video features



CNN-based video features

- You are welcome to design your own feature selection strategies.
(Please provide details in your report.)
- Some common preprocessing techniques
 - Average pooling
 - Concatenate
 - Fusion (Sum up image features by some weights)
 - Dimension reduction (PCA, etc.)



Pre-trained model

Convolution
Avgpool
Maxpool
Concat
Dropout
Fully connected
Softmax

CNN-based
video features

FC

predicted
labels

softmax

action labels

cross-entropy

You can keep this part fixed or
fine-tuned with a very small learning rate.

Training these parts

(☞ It might take lots of computation times and
resources for fine-tuning the pre-trained model.)

Task Description

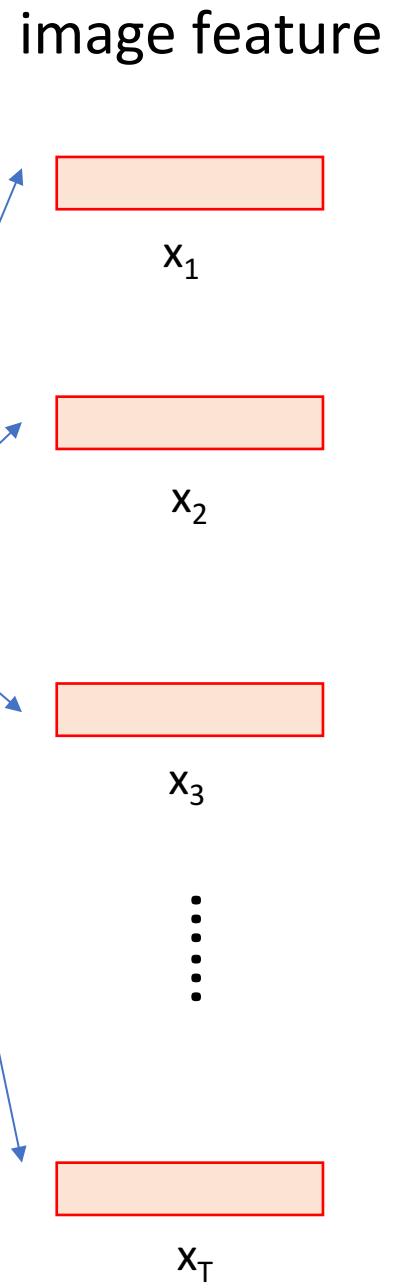
- Task 1 : Data preprocessing
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Video frames

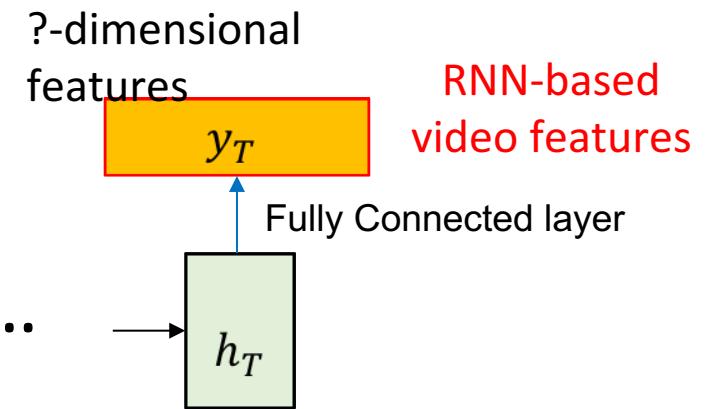
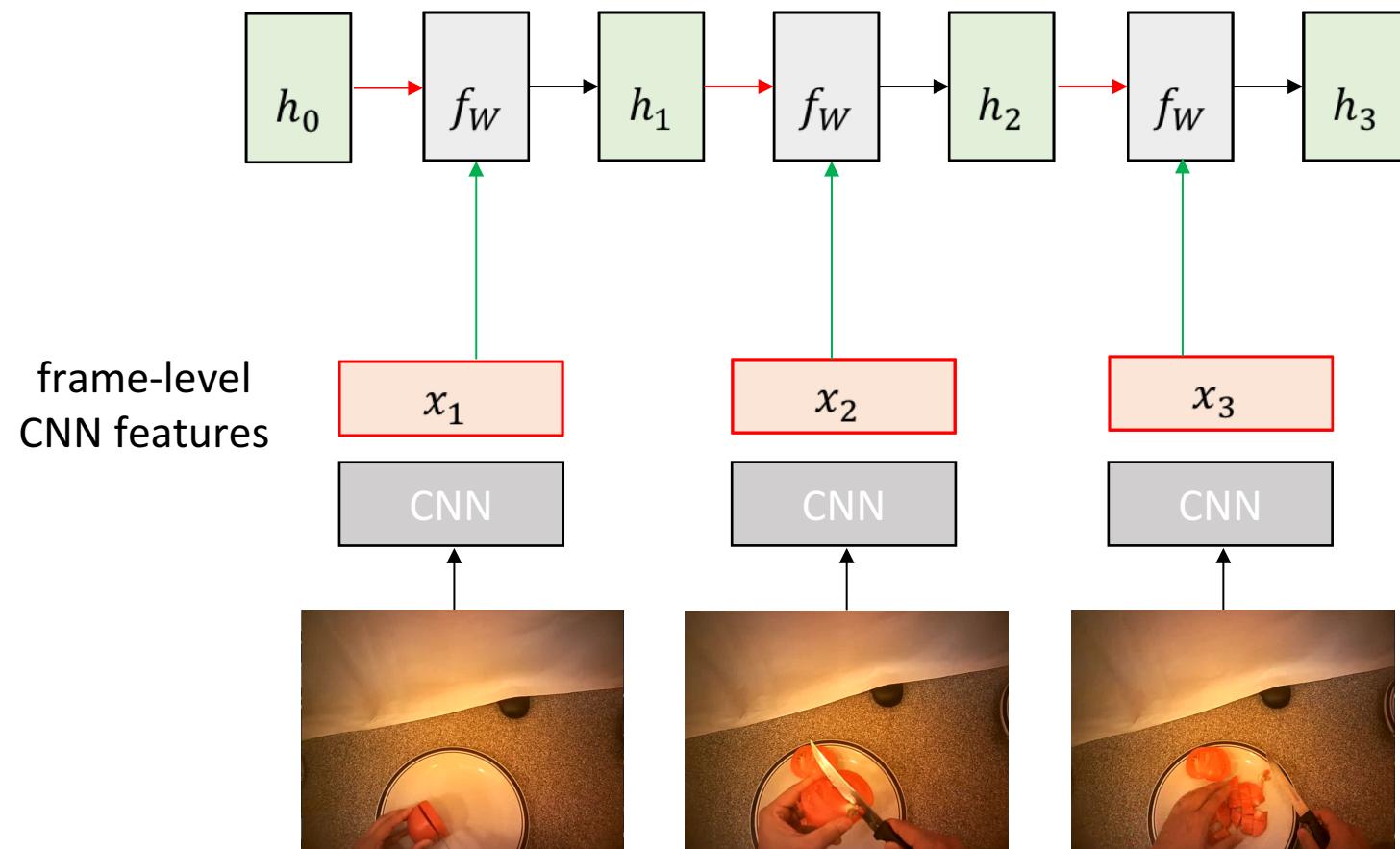


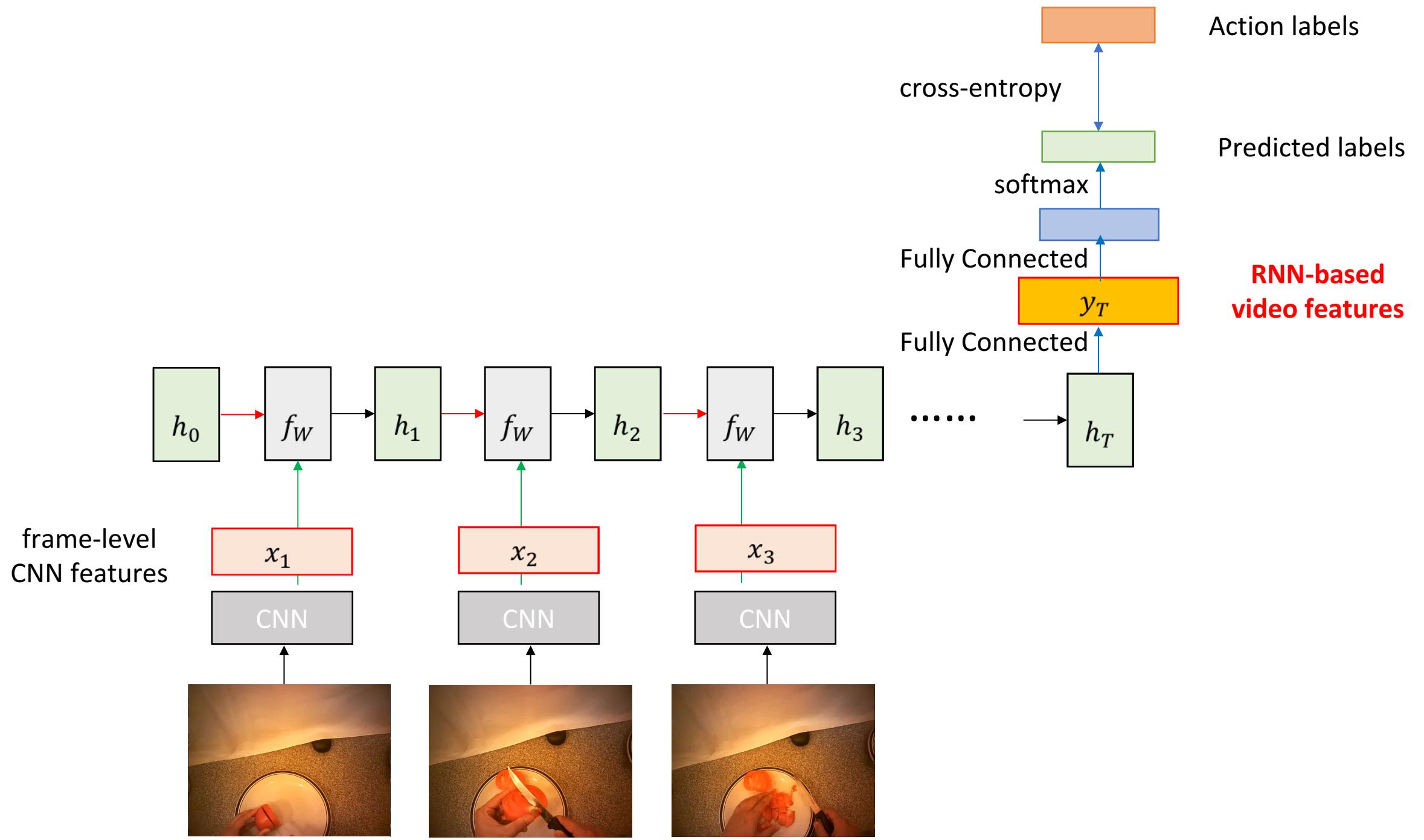
- Convolution
- AvgPool
- MaxPool
- Concat
- Dropout
- Fully connected
- Softmax

Pre-trained model

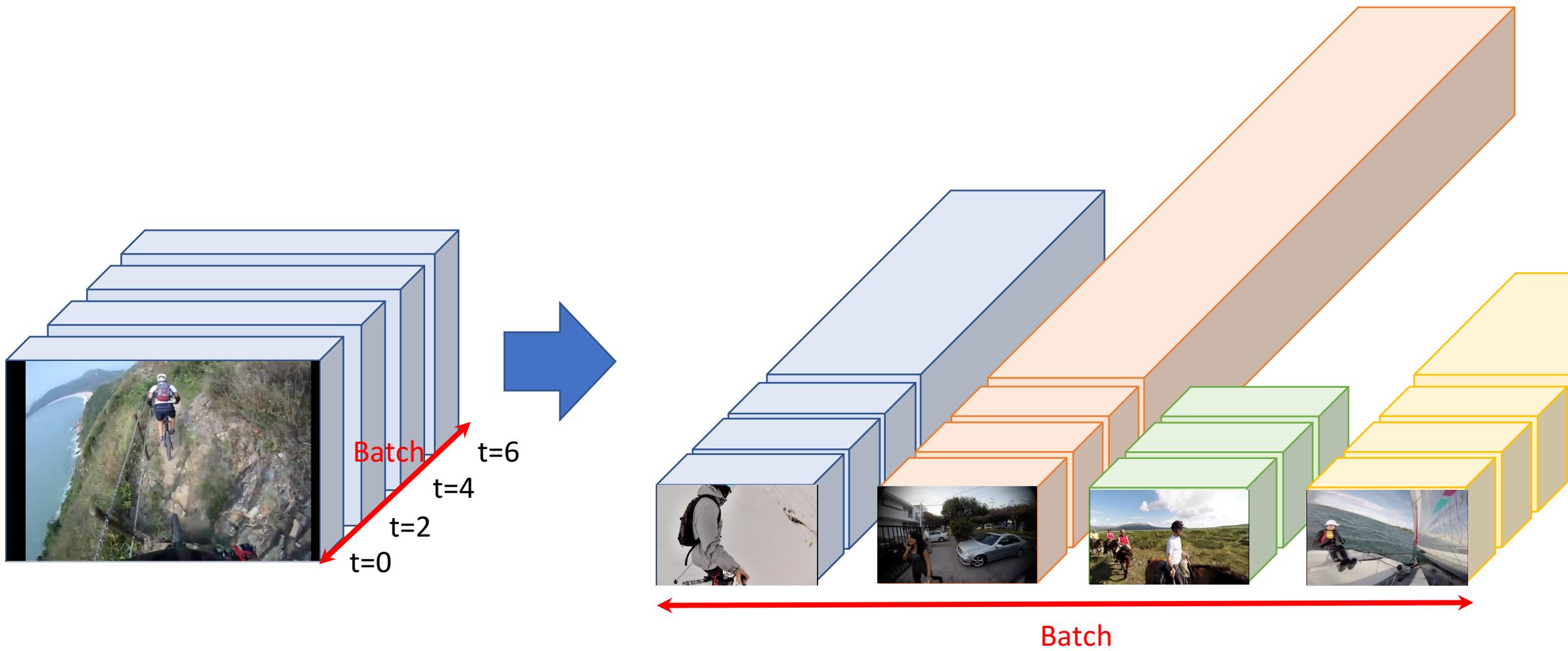


Feel free to design your LSTM structure here, e.g., bidirectional, multi-layers, etc.





Input with varying length in a batch



Solution

- Batch size = 1
- Zero padding and take valid output only

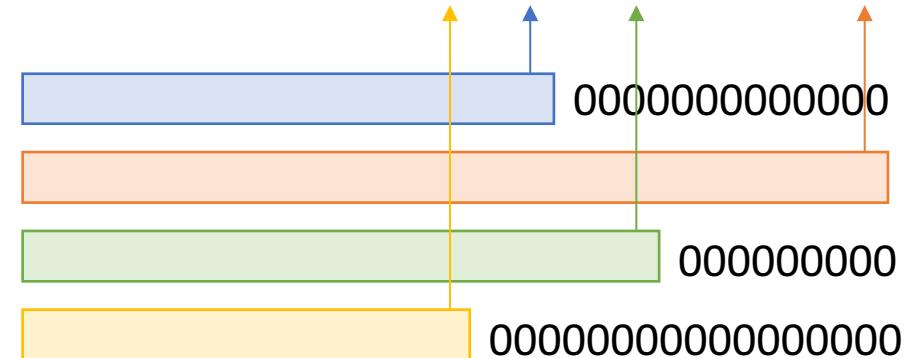
```
tf.nn.dynamic_rnn(  
    cell,  
    inputs,  
    sequence_length=None,  
    initial_state=None,  
    dtype=None,  
    parallel_iterations=None,  
    swap_memory=False,  
    time_major=False,  
    scope=None  
)
```

`torch.nn.utils.rnn.pack_padded_sequence(input, lengths, batch_first=False)` [source]

Packs a Tensor containing padded sequences of variable length.

Input can be of size $T \times B \times *$ where T is the length of the longest sequence (equal to `lengths[0]`), B is the batch size, and $*$ is any number of dimensions (including 0). If `batch_first` is True $B \times T \times *$ inputs are expected.

The sequences should be sorted by length in a decreasing order, i.e. `input[:, 0]` should be the longest sequence, and `input[:, B-1]` the shortest one.

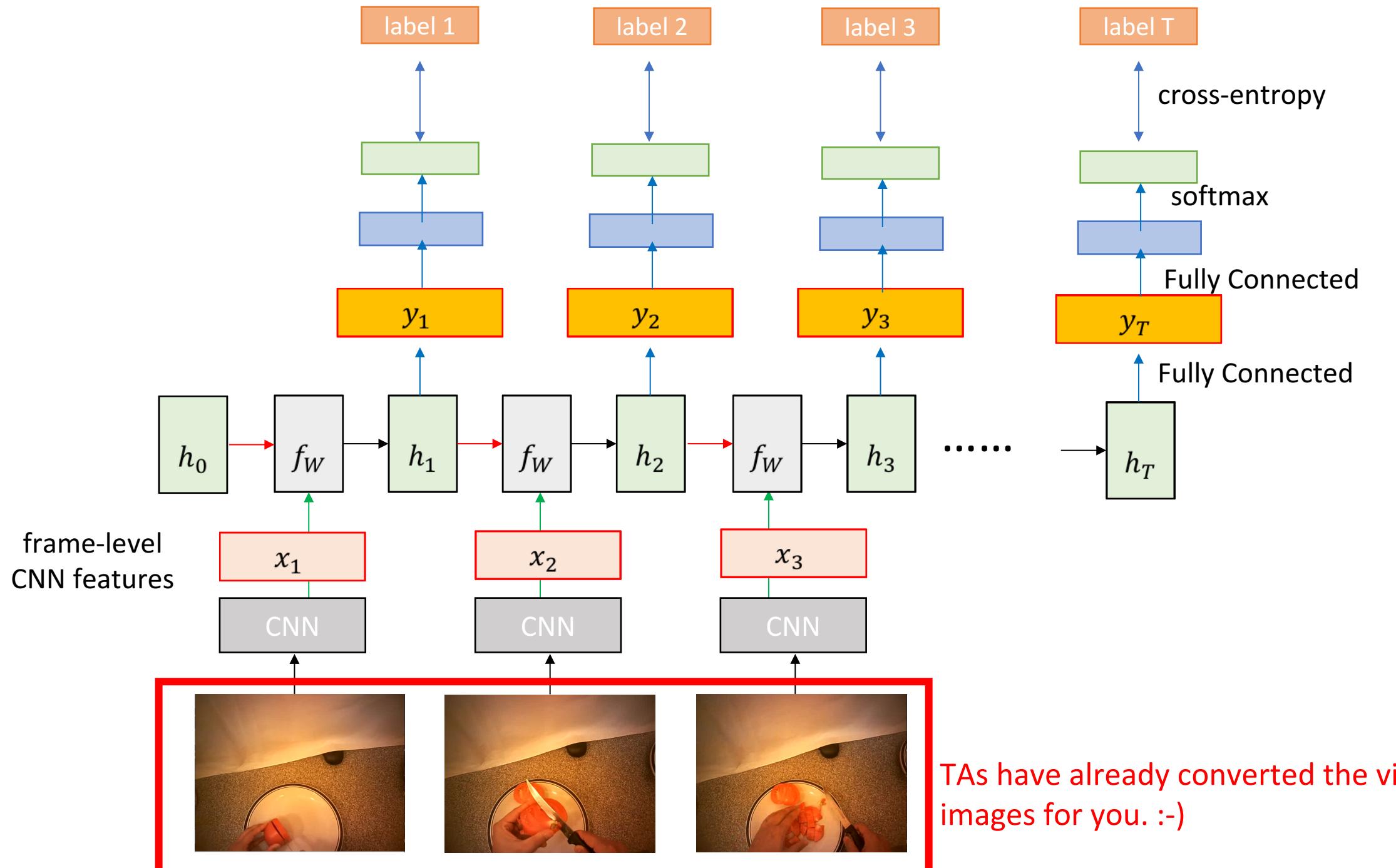


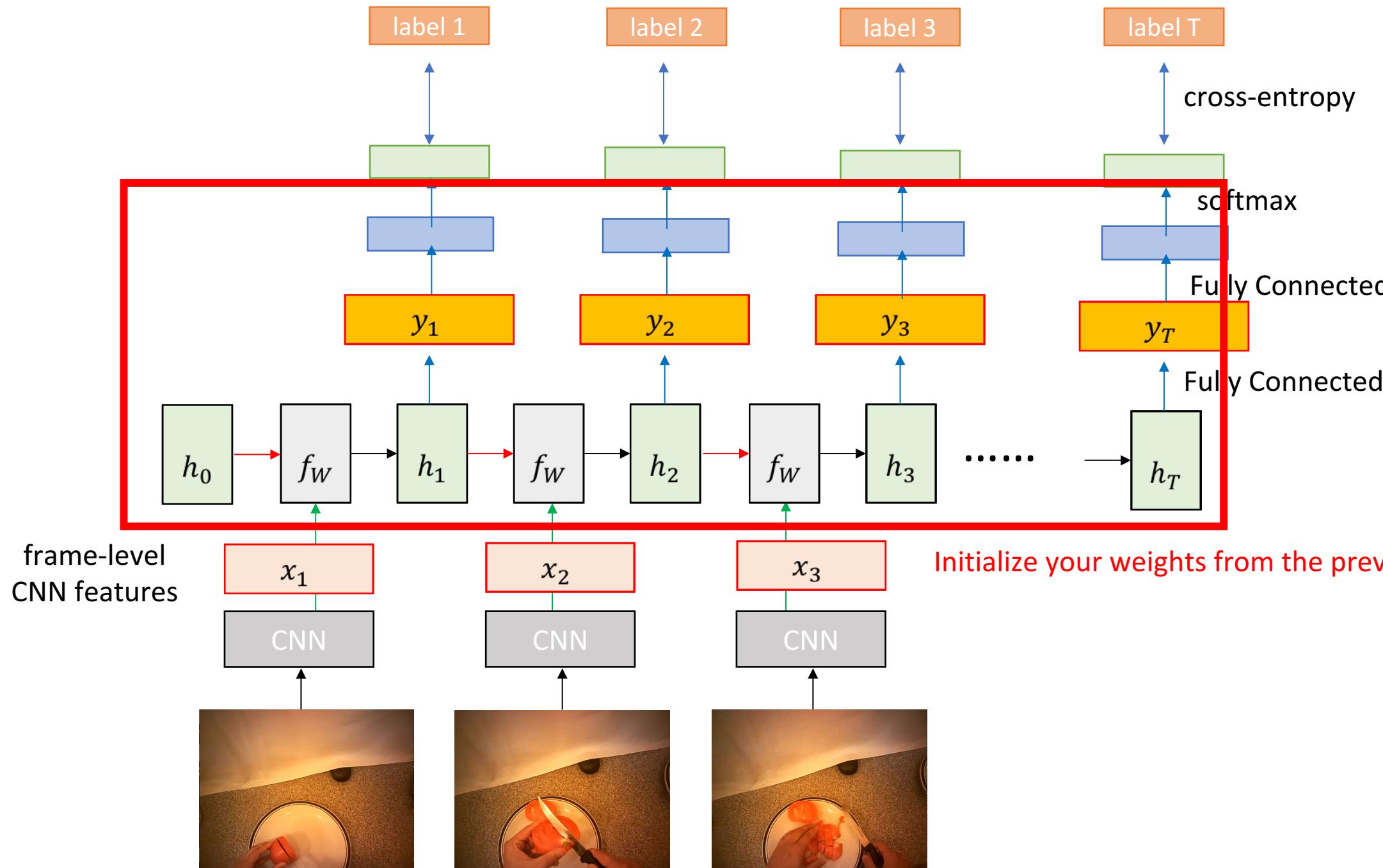
Reference of variable sequence length

- TensorFlow
 - <http://www.wildml.com/2016/08/rnns-in-tensorflow-a-practical-guide-and-undocumented-features/>
- Pytorch
 - <https://zhuanlan.zhihu.com/p/34418001>
 - <https://discuss.pytorch.org/t/understanding-pack-padded-sequence-and-pad-packed-sequence/4099>
- Keras
 - <https://github.com/keras-team/keras/issues/40>

Task Description

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How to handle very long sequence

- Max time steps should be about 250-500
- Use Truncated Backpropagation Through Time (TBTT)
 - https://www.tensorflow.org/tutorials/recurrent#truncated_backpropagation
 - <https://discuss.pytorch.org/t/implementing-truncated-backpropagation-through-time/15500/4>
- Cut / Down-sample full-length videos to suitable training size
- Reference
 - <https://machinelearningmastery.com/handle-long-sequences-long-short-term-memory-recurrent-neural-networks/>
 - <https://arxiv.org/abs/1803.04831>

Dataset



- Total 37 full-length videos
(each 5-20 mins in 24 fps)
- Split into 4151 trimmed videos
(each 5-20 secs in 24 fps)
- 21 action classes
- # of videos (Full / Trimmed)
Training : 23 / 2472
Validation : 7 / 792
Test : 7 / 887

Dataset

- Trimmed videos (For Task 1 & For Task 2)
 - train,valid – 240x320 trimmed videos are named as:
`<Video_category>/<Video_name><some_index>.mp4`
 - `gt_train.csv/gt_valid.csv`
`<Video_index>, <Video_name>, <Video_category>, <Start_times>, <End_times>, <Action_labels>, <Nouns>`

action labels

Other 0
Inspect/Read 1
Open 2
Take 3
Cut 4
Put 5
Close 6
Move Around 7
Divide/Pull Apart 8
Pour 9
Transfer 10

Dataset

- Full-length videos (For Task 3)
 - train,valid – 240x320 video frames in folder are named as:
<Video_category>/<Frame_index>.jpg
 - groundtruth - *<Video_category>.txt*
sequence of action labels correspond to their frame index.

action labels

Other	0
Inspect/Read	1
Open	2
Take	3
Cut	4
Put	5
Close	6
Move Around	7
Divide/Pull Apart	8
Pour	9
Transfer	10

Provided data

- Download the dataset and helper functions on ceiba
- Helper function to read videos as ndarray
 - sudo pip install sk-video
 - sudo apt-get install ffmpeg
- Helper function to read csv file as dictionary

Grading

- **Problem 1 : Data preprocessing (20%)**
 - Report your CNN performance and training strategies
- **Problem 2 : Trimmed action recognition (55%)**
 - Pass the validation baseline and the test baseline
 - Draw t-SNE graph for CNN-based (in Problem 1) & RNN-based video features (in Problem 2)
- **Problem 3 : Seq-to-Seq prediction in full-length videos (25%)**
 - Report your performance and make visualization figures
- **Bonus: Attention mechanisms (up to 20%)**

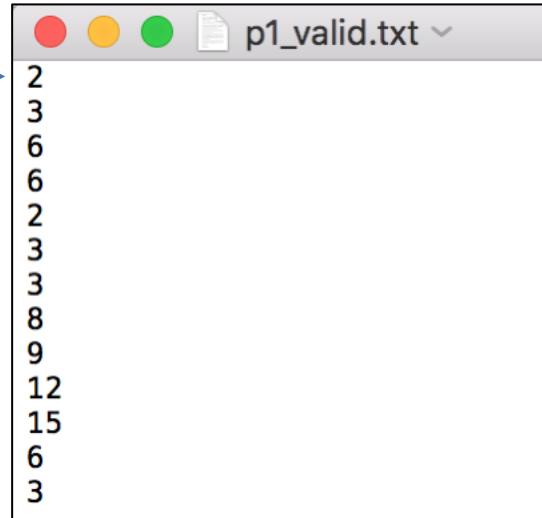
Grading

- Problem 1 : Data preprocessing (20%)
 - Describe your strategies of extracting CNN-based video features, training the model and other implementation details. (5%)
 - Report your video recognition performance using **CNN-based video features (10%)** and plot the learning curve of your model (5%).

Note that your code need to generate a txt file [\[p1_valid.txt\]](#) which contains 792 lines of numbers. Each number indicates the action label of the corresponding validation video.

Example of [p1_valid.txt]

Action label of #1 video →



```
2  
3  
6  
6  
2  
3  
3  
8  
9  
12  
15  
6  
3
```

↑
You need to have 792 lines

Grading

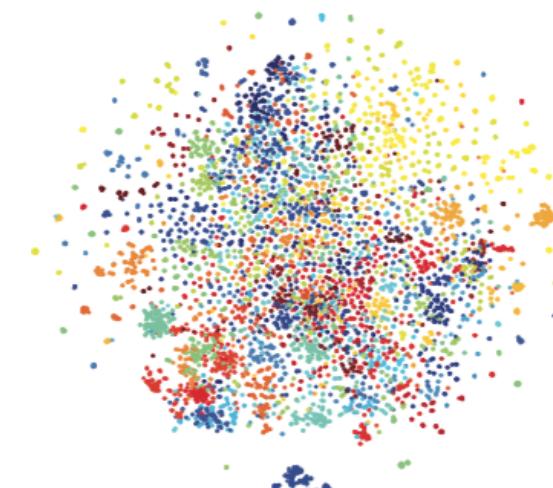
- Problem 2 : Trimmed action recognition (55%)
 - Describe your RNN models and implementation details for action recognition. (5%)
 - Your model should pass the baseline (valid: XXXXX / test: XXXXX) validation set (15%) / test set (20%, only TAs have the test set).

Your code need to generate [\[p2_result.txt\]](#) output file. Note that [\[p2_result.txt\]](#) would consist of either 792 lines for validation videos or 887 lines for test videos.

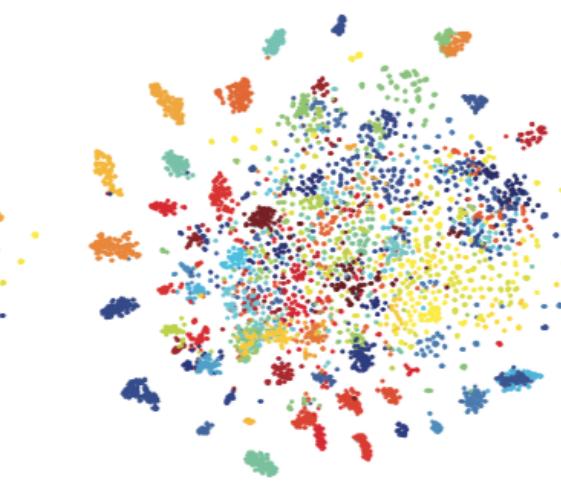
Grading

- Problem 2 : Trimmed action recognition (55%)
 - Visualize **CNN-based video features** and **RNN-based video features** to 2D space (with tSNE) in your report. You need to generate two separate graphs and color them with respect to different action labels. Do you see any improvement for action recognition? Please explain your observation (**15%**).

Example visualization:



CNN-based features



RNN-based features

本圖不代表實際data分佈情形，僅供同學參考

Grading

- Problem 3 : Temporal action segmentation (25%)
 - Describe any extension of your RNN models, training tricks, and post-processing techniques you used for temporal action segmentation. (5%)
 - Report validation accuracy and plot the learning curve (10%) in your report.
- For each video, you need to generate [[`<Video_category>.txt`](#)] which contains a sequence of action labels corresponding to each frame. Note that you need to generate 7 files for validation set in total.

Example of [<Video_category>.txt]

Action label of #1 frame →



```
0
0
0
0
0
9
9
9
9
9
12
12
12
3
3
```

Total # of lines must match the ground-truth file

Grading

- Problem 3 : Temporal action segmentation (25%)
 - Choose one video from the 7 validation videos to visualize the best prediction result in comparison with the ground-truth scores in your report. Please make your figure clear and explain your visualization results. You need to plot at least 300 continuous frames (2.5 mins). (**10%**)



Example of visualization results

Bonus (up to 20%)

- Extra points will be given, if you implement and integrate **attention mechanisms** to improve both trimmed action recognition and temporal action segmentation. Please note that you **CANNOT** use any pre-defined attention functions or models for this part, i.e., you need to implement attention mechanism in each time step.
- You will be graded by the details of your implementation. You need to show reasonable experiment results plus detailed discussions in your report. 0 point will be given if negligible improvement. Thus, do not try to work on bonus problem without making efforts.

Homework Policy - Submission

- **DLCV2018SPRING/hw5** on your GitHub repository should include the following files:
 - hw5_YourStudentID.pdf ([report template](#))
 - hw5_p1.sh (for **Problem 1**)
 - hw5_p2.sh (for **Problem 2**)
 - hw5_p3.sh (for **Problem 3**)
 - your python files (e.g., Training code & Testing code)
 - your model files (can be loaded by your python file)

Homework Policy - Submission

- If your model is larger than GitHub's maximum capacity (100MB), you can upload your model to another cloud service (e.g., Dropbox). However, your script file should be able to download the model automatically.
- Dropbox tutorial: [link](#)

Homework Policy - Execution

- TA will run your code as shown below
 - `bash hw5_p1.sh $1 $2`
 - \$1: directory of **trimmed** validation videos folder
 - \$2: directory of output labels folder
 - `bash hw5_p2.sh $1 $2`
 - \$1: directory of **trimmed** validation/**test** videos folder
 - \$2: directory of output labels folder
 - `bash hw5_p3.sh $1 $2`
 - \$1: directory of **full-length** validation videos folder
 - \$2: directory of output labels folder

Homework Policy - Packages

- Python : 3.6
- Tensorflow : 1.6
- Keras : 2.1.5
- Pytorch : 0.3.1
- h5py : 2.7.1
- Numpy : 1.14.2
- Pandas : 0.22.0
- Matplotlib, Scikit-image, Pillow, Scipy, **any video IO Lib**, Python standard Lib.
- E-mail or ask TA first if you want to import other packages.

Deadline



2018/06/08 (Fri) 23:59:59 (GMT+8)

Rules

- Delay quota: Deducted 30% each day excluding using 3 free late day quota this semester
- Academic Ethics: Discussion between classmates is encouraged, however, please do NOT copy (or let someone copy your) homework. TA will check the similarity of every submission.
- Rules Violation: Violation of any format/execution specification will result in zero points. Please follow homework spec carefully and ask without any hesitation.
- External Dataset: Using external dataset is forbidden for this homework

Academic Integrity

- Can discuss HW with peers, but DO NOT copy and/or share code
 - 任一次作業抄襲/被抄襲者，按校規論且本課程學期成績為F!
 - This is university policy and not negotiable.
- Do not directly use code from Internet unless you have permissions.
 - If not sure, ask!
 - If so, do specify in your HW/project.
- Do not use your published work as your final project.
 - However, you are encouraged to turn your high-quality projects into publications.

Very kind reminder from TA

- TA hours is not 24/7:
同學有任何問題TA都很樂意討論與回答，
不過請同學遵守**TA hours**的時間規定。
- TA不是通靈少女， 請不要貼error或learning curve叫TA Debug...
 - Google你的error message可以解決90%的疑慮
 - 請先看過[Facebook 討論版](#)置頂Q&A， 同學有問過的問題都會更新
 - 在[Facebook 討論版](#)詳細描述你的問題， TA跟熱心的同學都會幫忙回答
 - 寄信詳細描述你的問題到[TA信箱](#)， 並耐心等候TA的回覆
- 請不要私訊給任何一位TA！只會得到已讀訊息...