

Microsoft Research
Faculty
Summit
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Recognizing Human Activities At Scale

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An image is worth one thousand words



Image:

“a tiger attacking a person on a grass field”



“a man packing a suitcase in a store”



Source: YouTube

Video:

“the tiger is being **playful**”

“the man is **unpacking** the suitcase”

“the person is attempting to **pick** the lock”

Video Generation and Consumption is Huge

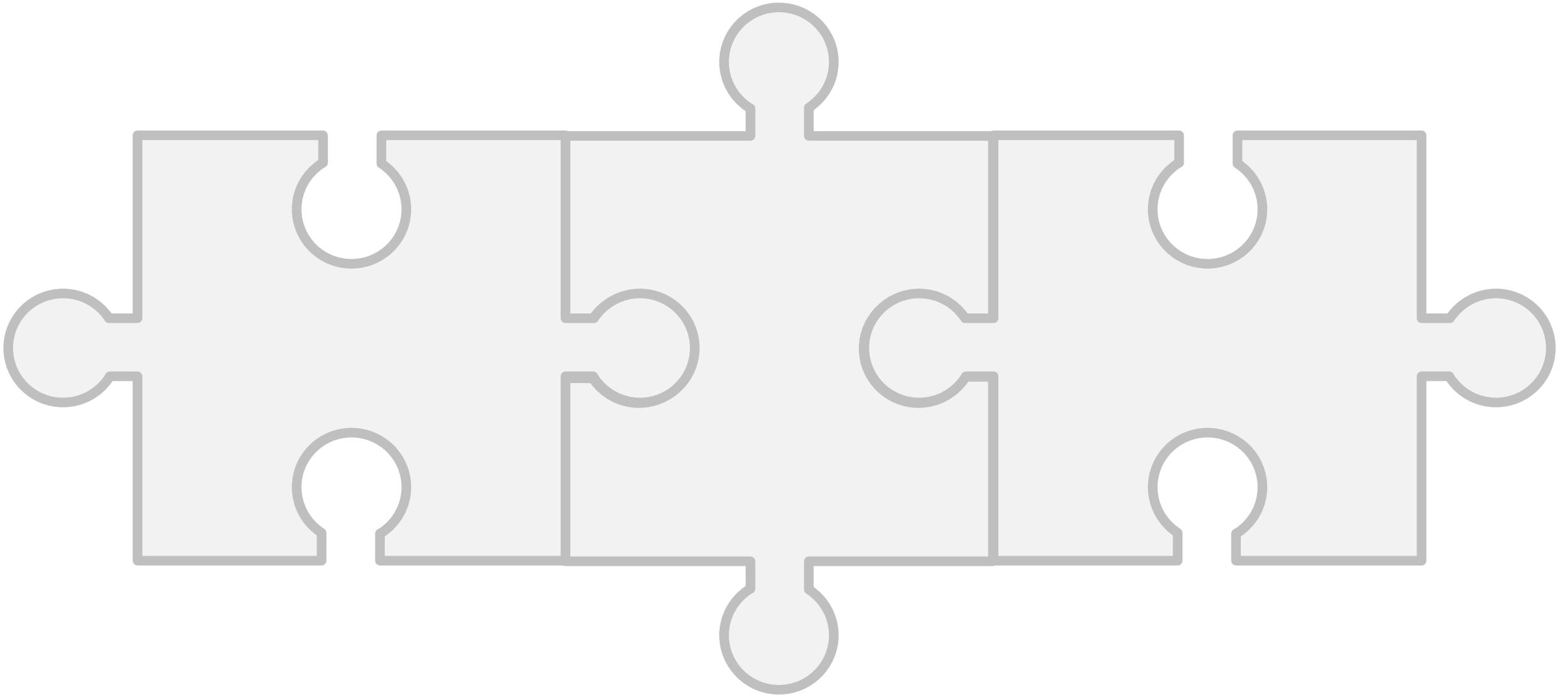
Netflix: 100 million hours watched per day

YouTube: 400 hours uploaded per minute

Cisco: ~1 million minutes of video per second by 2020
~200 peta-pixels/second



Recognizing Human Actions



Action Recognition in Videos



KTH dataset [Schuldt, 2004]



HOHA dataset [Laptev, 2008]



UCF101 dataset [Soomro, 2012]

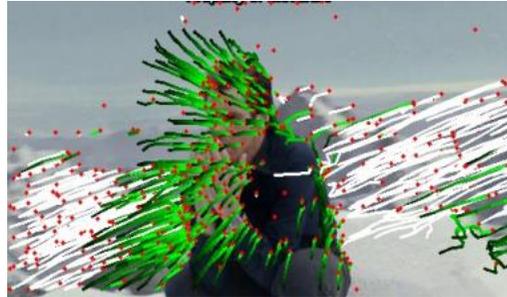
Short, pre-trimmed videos, only containing one action

Action Recognition in Videos

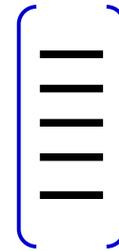
Traditional action classification pipeline



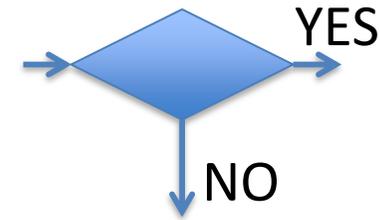
Input Video
(pre-trimmed)



Feature extraction
(handcrafted/learned)

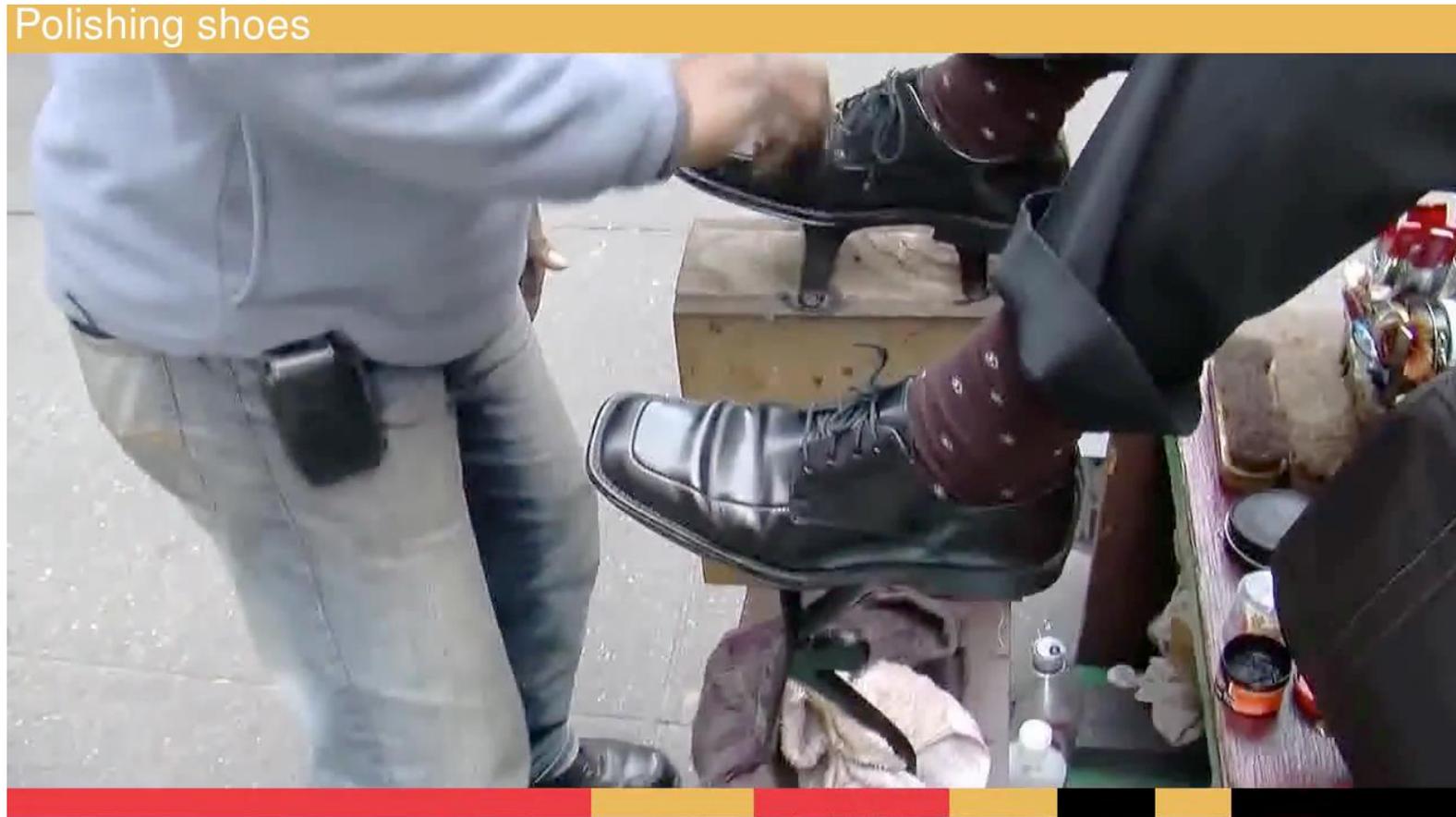


Feature
encoding



Classifier

Temporal Detection of Actions



Temporal Detection of Actions

Long input video



┌─── “Polishing shoes” ───┐
Classifier

- Apply complex classifier at each temporal location frame
- Exhaustive search
- Repeat for all actions we want to detect
- Questionable scalability

Fast Activity Proposals for Action Detection

Long input video



Generic Activity Proposal

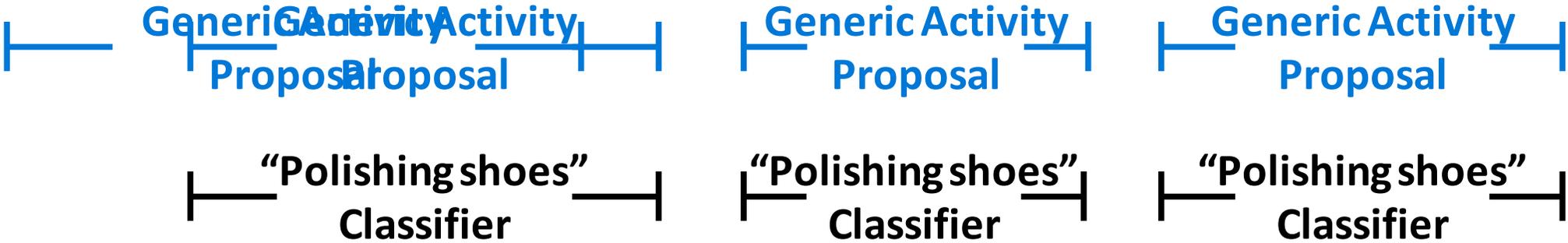
- Runs very quickly (>130 fps)
- Find all temporal intervals that contain “any activity”
- Retrieve action intervals with *high recall*

[Caba Heilbron, Niebles & Ghanem. CVPR 2016]

[Escorcia, Caba Heilbron, Niebles & Ghamen, ECCV 2016]

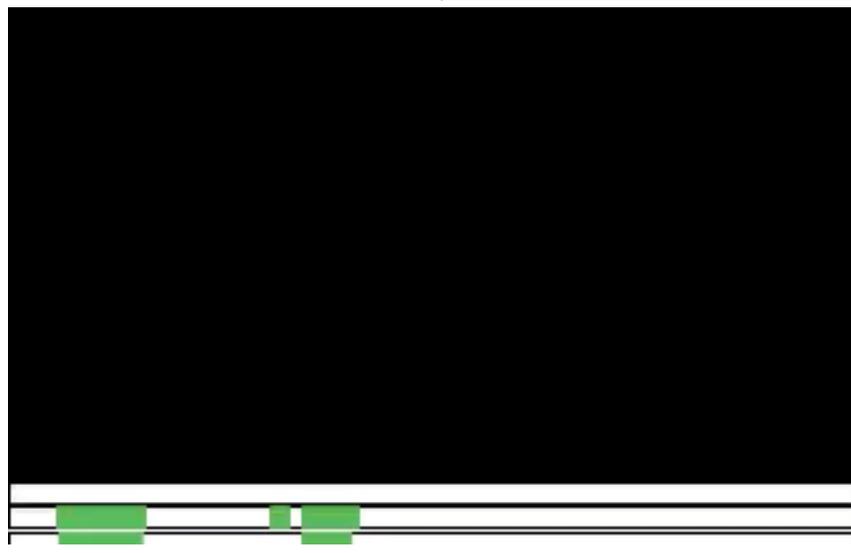
Fast Activity Proposals for Action Detection

Long input video



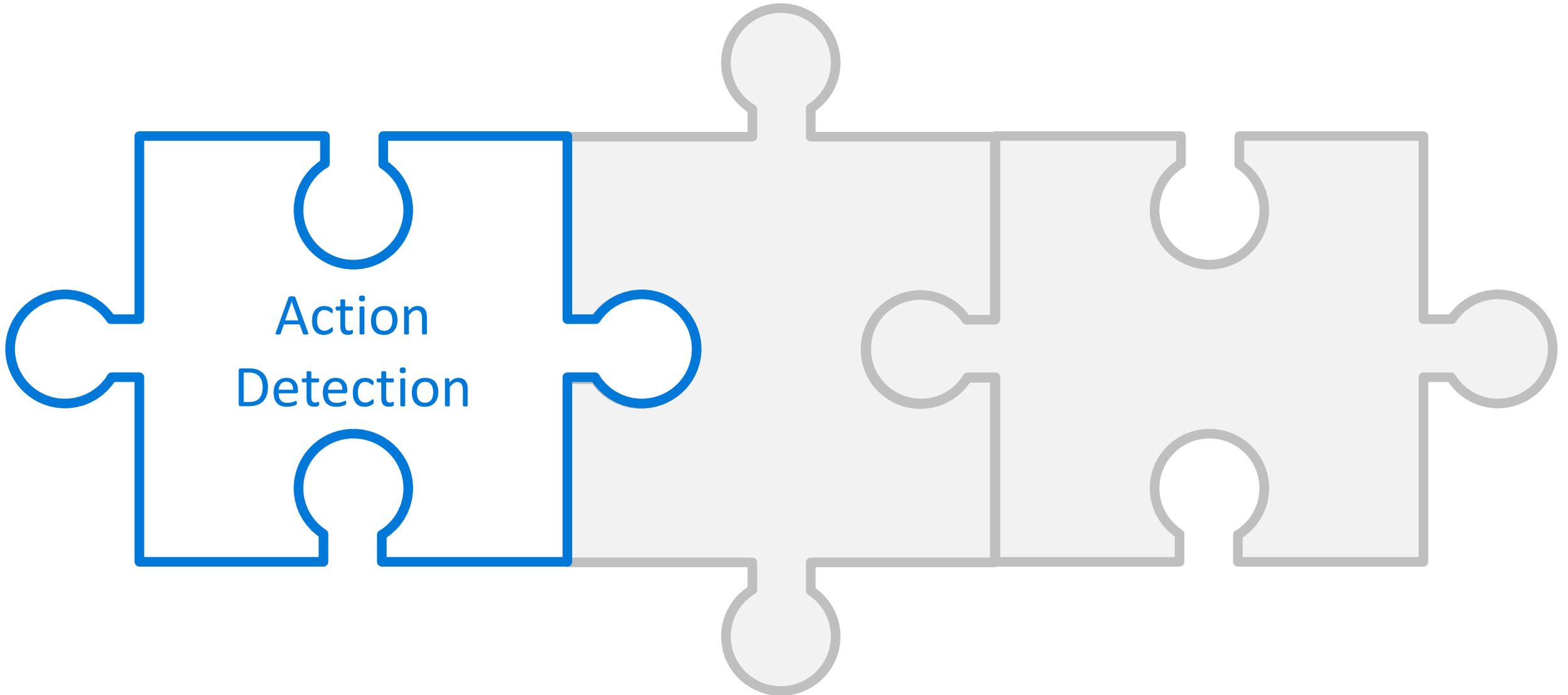
[Caba Heilbron, Niebles & Ghanem. CVPR 2016]
[Escorcia, Caba Heilbron, Niebles & Ghamen, ECCV 2016]

Fast Activity Proposals for Action Detection

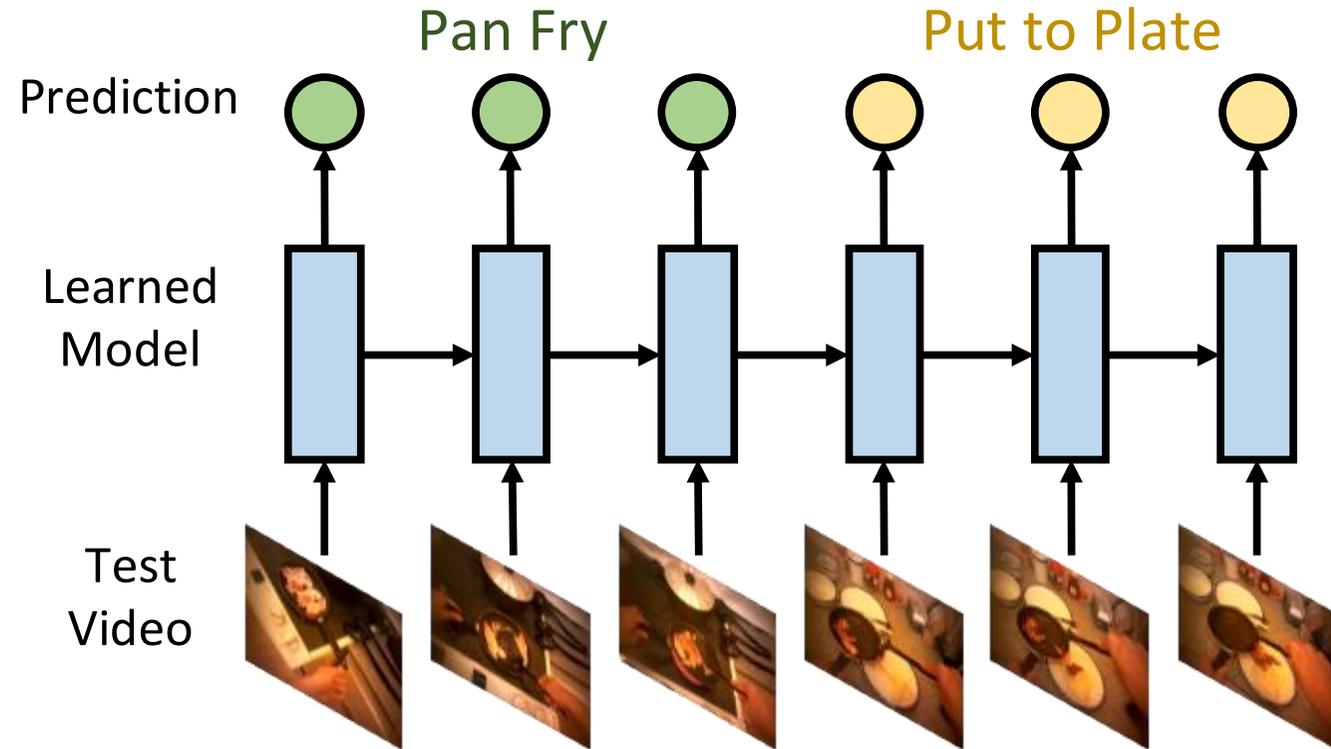


[Caba Heilbron, Niebles & Ghanem. CVPR 2016]
[Escorcía, Caba Heilbron, Niebles & Ghamen, ECCV 2016]

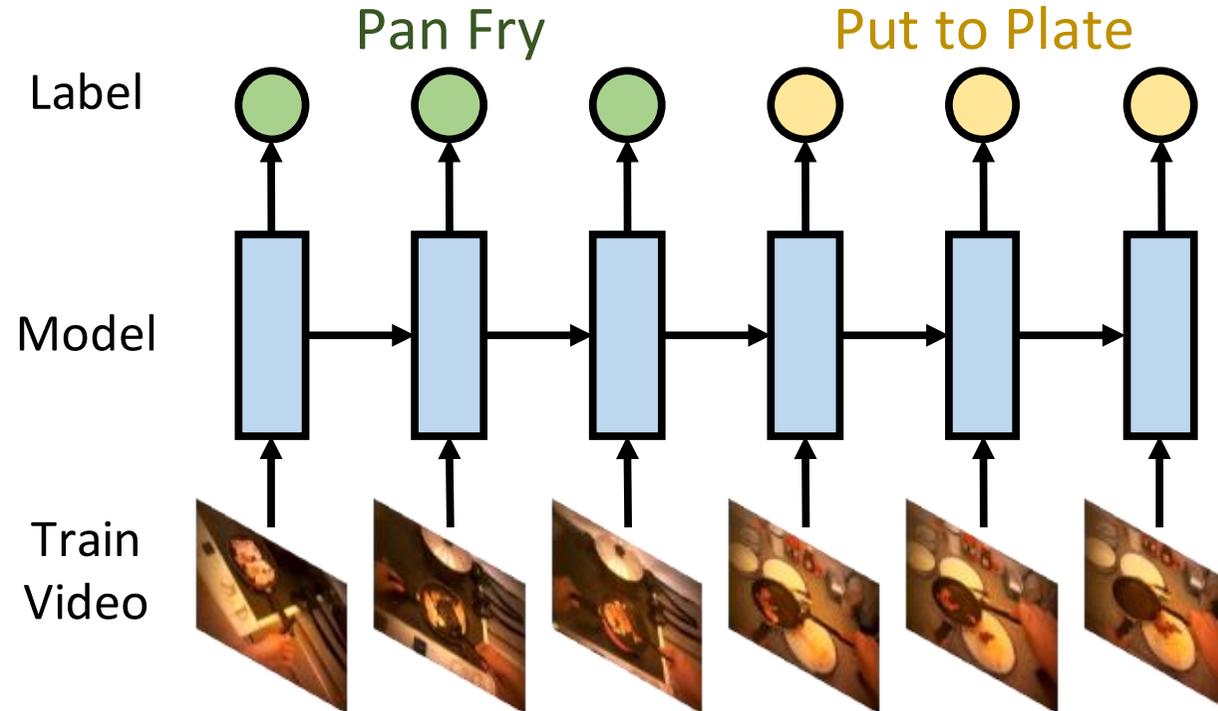
Recognizing Human Actions



Temporal Action Labeling

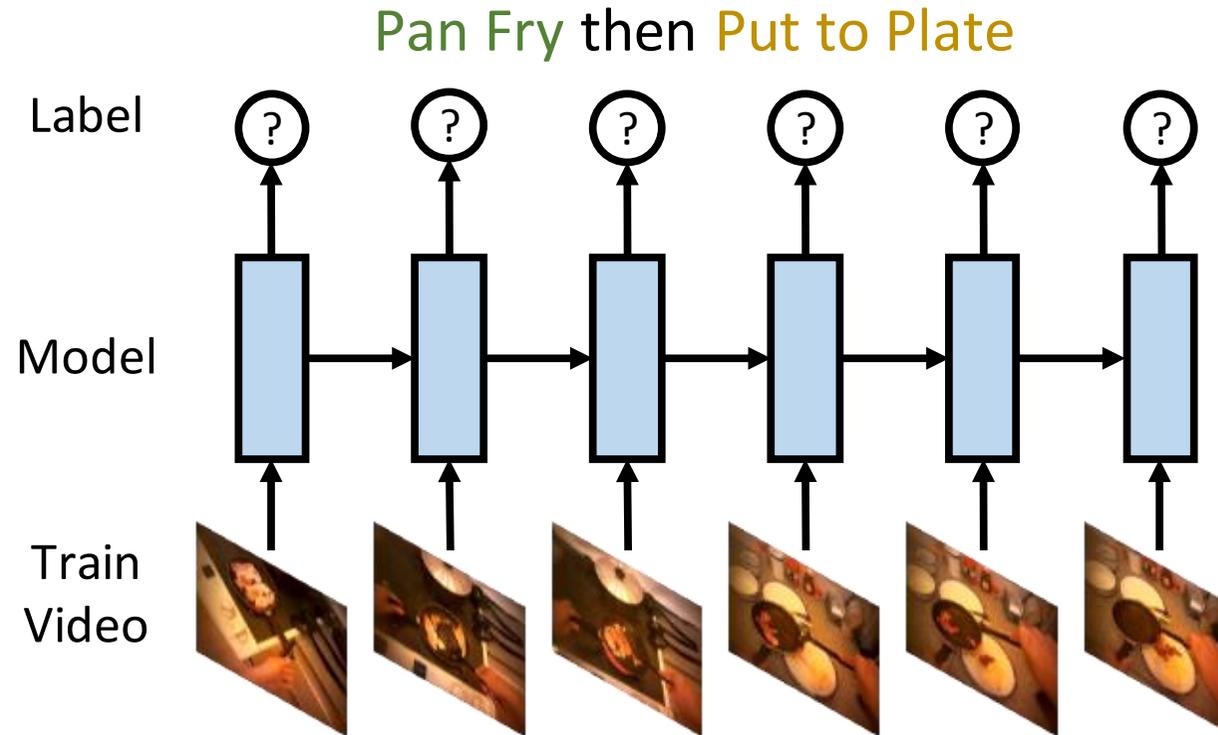


Fully Supervised Learning



- Many training videos with per frame action labels
- Costly to annotate!

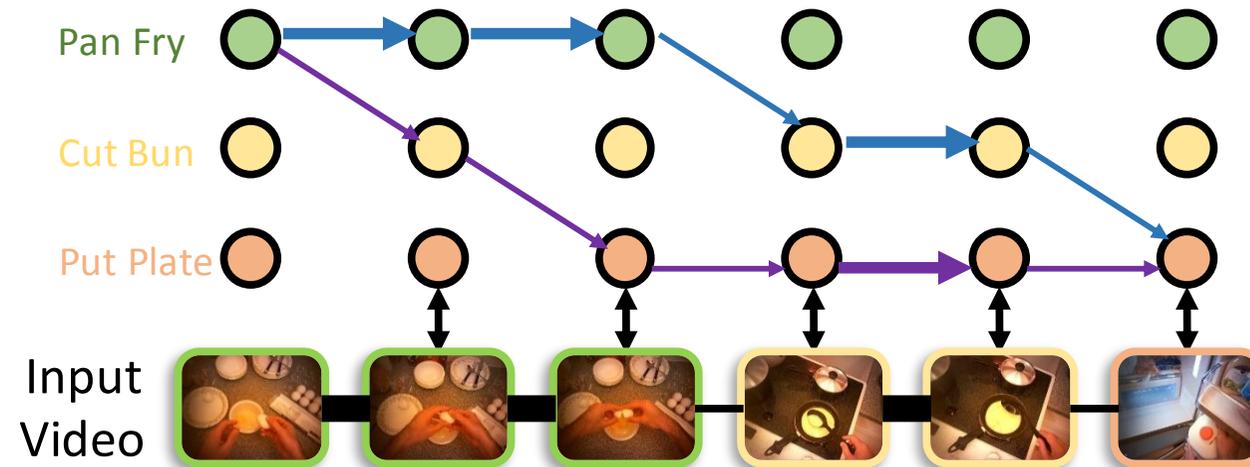
Weakly-Supervised Learning



- Only use action ordering
- Disambiguate by aggregating across with many training videos

[Huang, Fei-Fei & Niebles, ECCV 2016]

Extended Connectionist Temporal Classification



- Extends the CTC framework
- Explores space of frame-to-labels assignments efficiently
- Incorporates pairwise frame similarities

[Huang, Fei-Fei & Niebles, ECCV 2016]

Evolution of Training Frame-to-Label correspondence



Our approach starts without label correspondences for the training videos and iteratively improves the localization of the actions.

Weakly Supervised Activity Segmentation Results



[Huang, Fei-Fei & Niebles, ECCV 2016]

Weakly Supervised Action Detection Results

Drive Car



[Huang, Fei-Fei & Niebles, ECCV 2016]

Weakly Supervised Action Detection Results

Drive Car



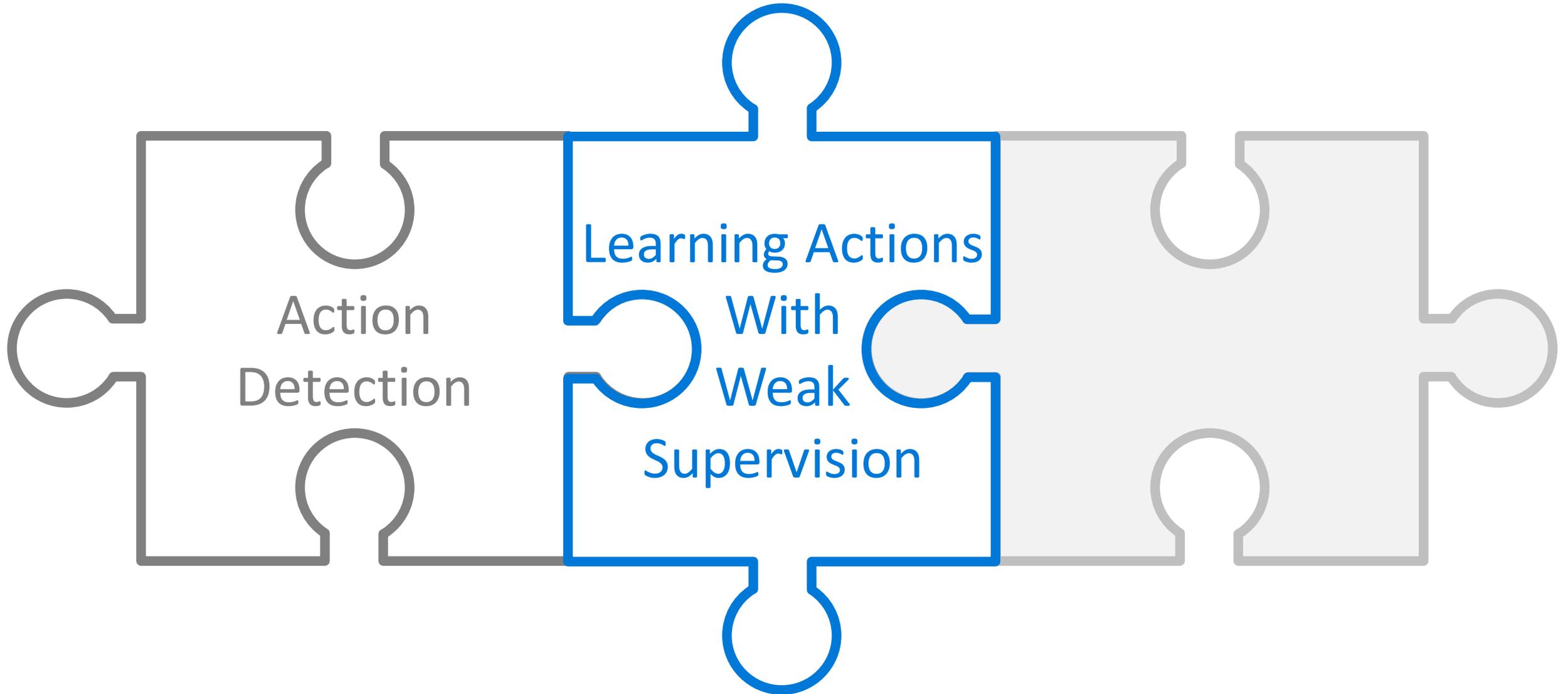
[Huang, Fei-Fei & Niebles, ECCV 2016]

Hierarchical Modeling of Composable Activities

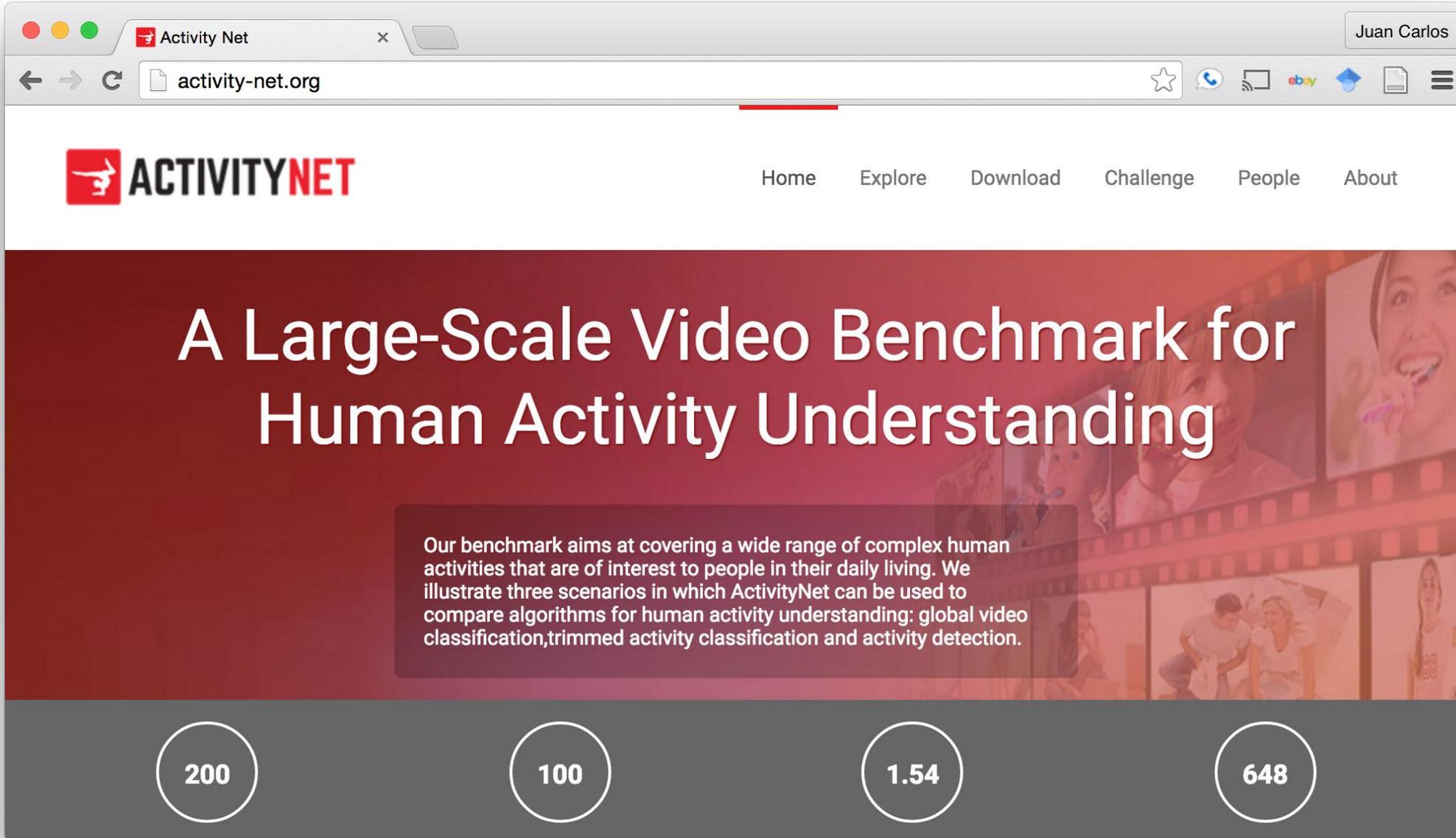


[Lillo, Soto & Niebles,
CVPR 2014]
[Lillo, Niebles & Soto,
CVPR 2016]

Recognizing Human Actions



ActivityNet – www.activity-net.org



The screenshot shows a web browser window with the URL activity-net.org. The page features the ActivityNet logo (a red square with a white hand icon) and a navigation menu with links for Home, Explore, Download, Challenge, People, and About. The main content area has a dark red background with a filmstrip overlay showing various human activities. The headline reads "A Large-Scale Video Benchmark for Human Activity Understanding". Below this, a text box explains the benchmark's goal: "Our benchmark aims at covering a wide range of complex human activities that are of interest to people in their daily living. We illustrate three scenarios in which ActivityNet can be used to compare algorithms for human activity understanding: global video classification, trimmed activity classification and activity detection." At the bottom, four white circles contain the numbers 200, 100, 1.54, and 648.

Activity Net

activity-net.org

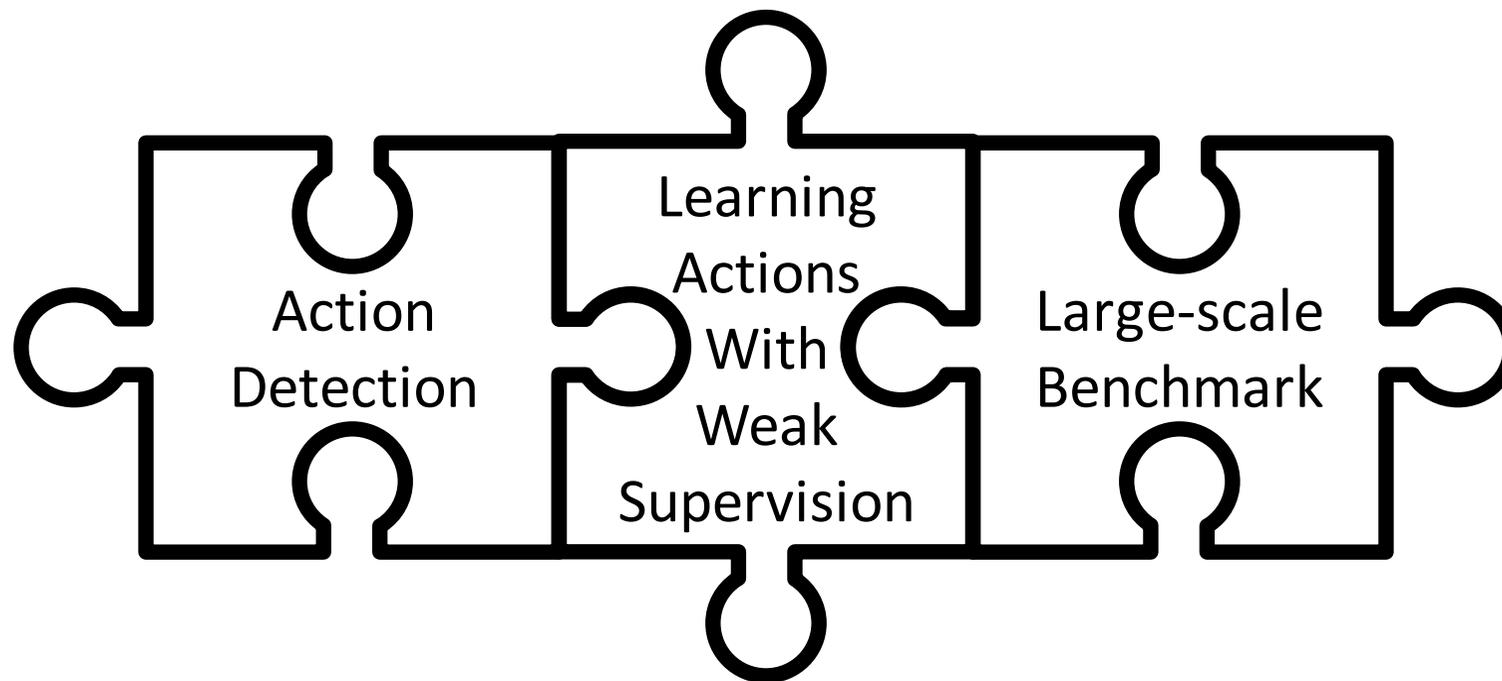
Home Explore Download Challenge People About

A Large-Scale Video Benchmark for Human Activity Understanding

Our benchmark aims at covering a wide range of complex human activities that are of interest to people in their daily living. We illustrate three scenarios in which ActivityNet can be used to compare algorithms for human activity understanding: global video classification, trimmed activity classification and activity detection.

200 100 1.54 648

[Caba Heilbron,
Escorcia, Ghanem &
Niebles, CVPR 2015]



Thank you!

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