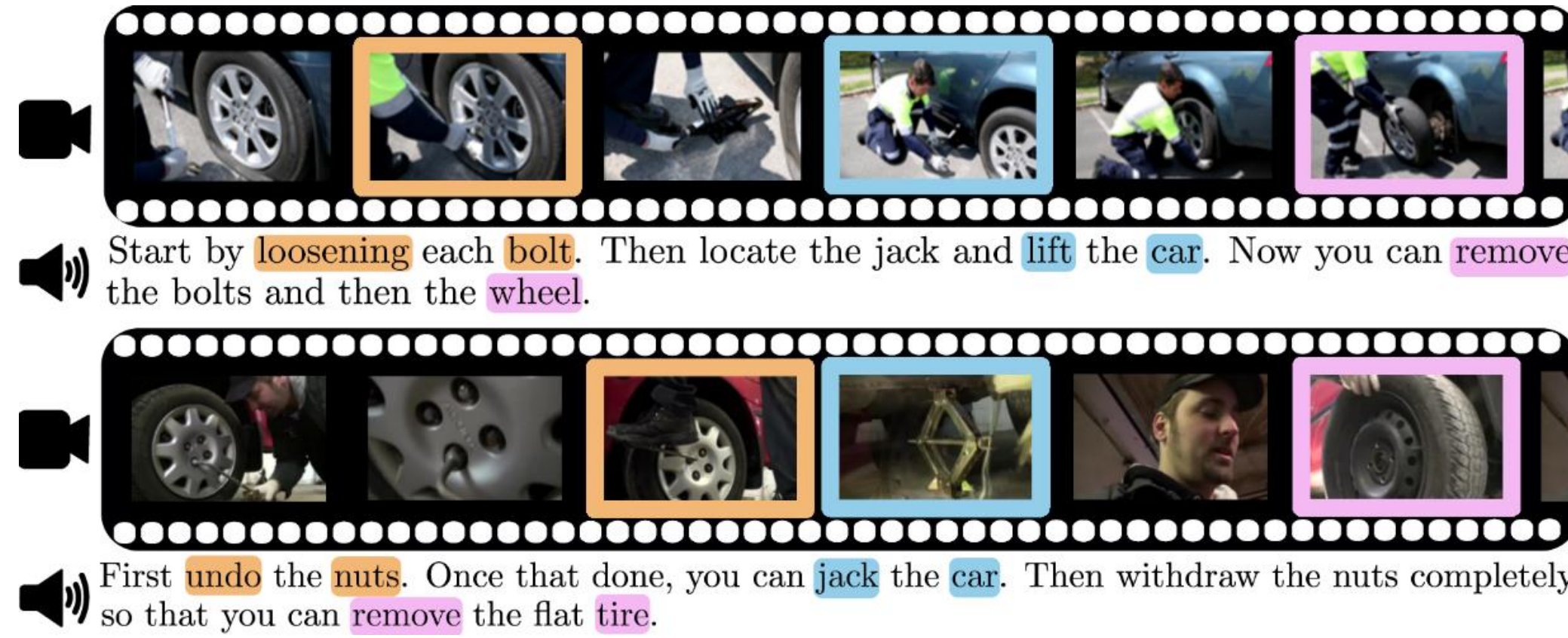


Goal and overview

The problem:

Automatically learn the main steps to complete a given task from narrated instruction videos.

Input: A set of narrated instruction videos.



Outputs:

- List of K (input) main steps
- Visual and linguistic representations of the steps
- Temporal localization of each step in the videos

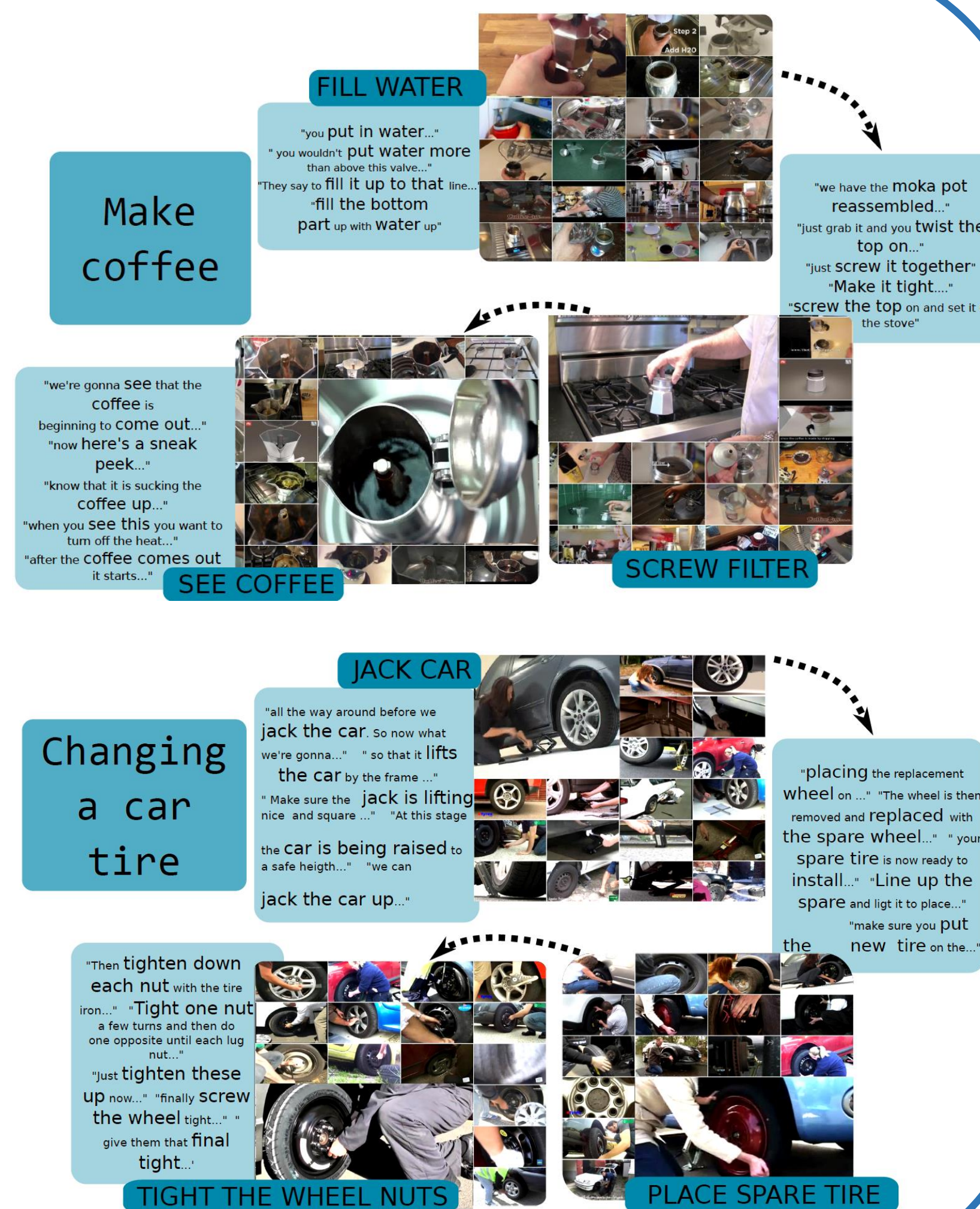
- Loosen nuts
- Jack the car
- Remove the flat tire

Contributions:

- Collected and annotated a new dataset of narrated instruction videos,
- Developed an unsupervised learning method that takes advantage of the complementary nature of the text and video,
- Experimentally demonstrated recovery of main steps and their locations in video.

A new dataset

- 5 tasks:
 - Changing a car tire
 - Repot a plant
 - Make coffee
 - Perform CPR
 - Jump a car battery
- 30 videos per task (total of 800,000 frames).
- Manual correction of the ASR transcripts.
- Manual annotation of 7-10 main steps and time localization in each video (only used for evaluation).



Approach

Assumptions

- Each task is performed by an ordered sequence of steps.
- People do what they say (roughly) when they say it

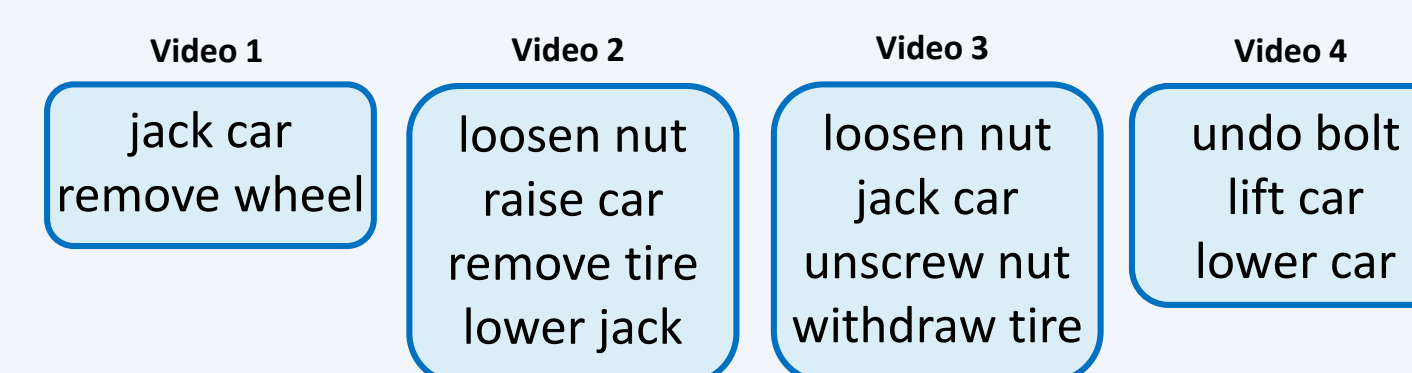
Two linked clustering steps

- Text clustering : multiple sequence alignment
- Discriminative video clustering under text constraints

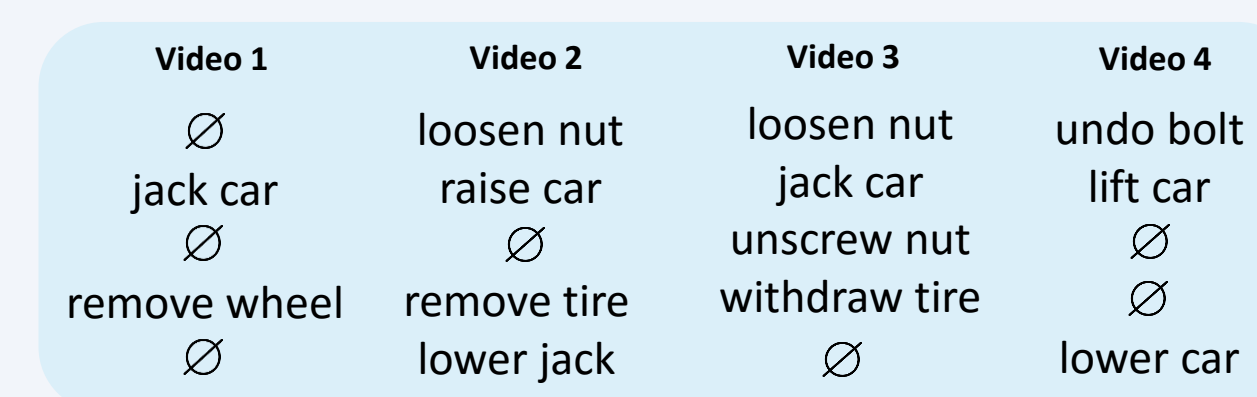
Multiple sequence alignment (MSA)

- Text signals are first processed into sequences of direct object relations: *Let's now jack the car.* → DOBJ = (jack, car)
- Similarity between dobj is obtained using Wordnet.

INDIVIDUAL INPUT SEQUENCES:



OUTPUT OF THE MSA:



New! Formulate MSA as a QP, and approximately solve it with Frank-Wolfe

FINAL OUTPUT:

Video 1	Video 2	Video 3	Video 4	Agreement	Discovered list of steps
∅	loosen nut	loosen nut	undo bolt	3	1) Loosen nut 2) Jack car 3) Remove wheel
jack car	raise car	jack car	lift car	4	
∅	∅	unscrew nut	∅	1	
remove wheel	remove tire	withdraw tire	∅	3	
∅	lower jack	∅	lower car	2	

1st STEP : TEXT
Identify main steps

Discriminative clustering under text constraints

CONSTRAINED OPTIMIZATION PROBLEM:

$$\underset{Z}{\text{minimize}} \quad h(Z) \quad \text{s.t.} \quad Z \in \mathcal{Z}, \quad AZ \geq R$$

where $h(Z)$ is a discriminative clustering cost [1]:

$$h(Z) = \min_{W \in \mathbb{R}^{K \times d}} \frac{1}{2T} \|Z - XW\|_F^2 + \frac{\lambda}{2} \|W\|_F^2$$

Representation of video (IDTF, CNN) [TxK] matrix

Discriminative loss on data

Regularizer

Linear action classifier [dxK] matrix

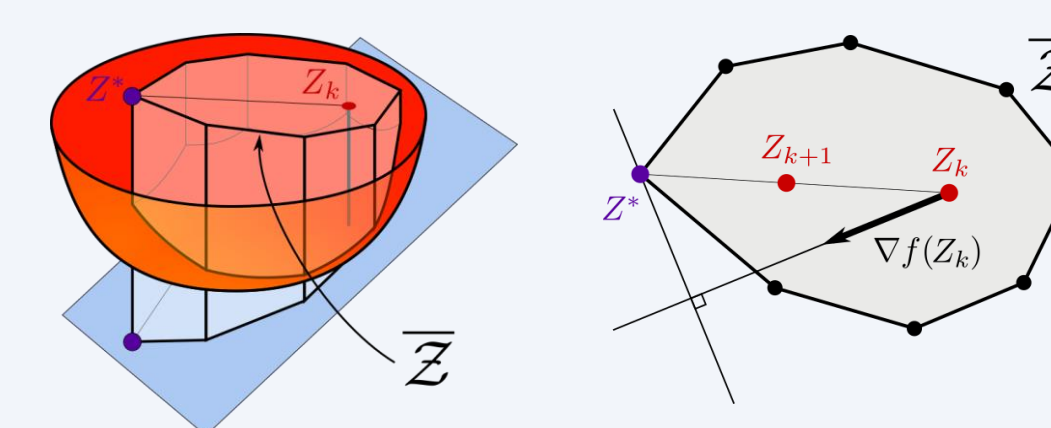
ordered script

Subtitle Alignment [SxT] matrix

weak textual constraints

OPTIMIZATION METHOD [2,3]:

- Optimize convex relaxation using Frank-Wolfe
- Use DP as the linear oracle
- Cost classifier based rounding



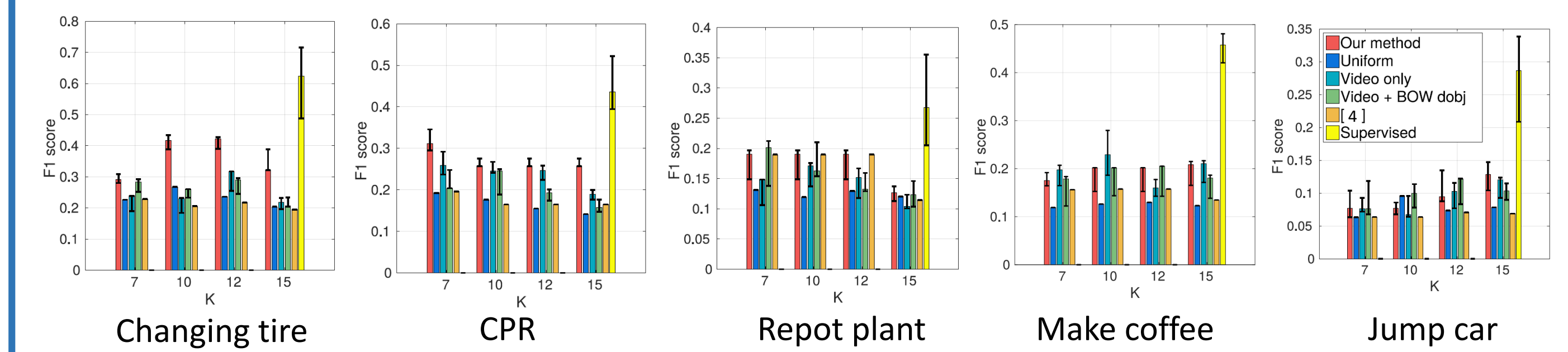
2nd STEP : VIDEO
Temporally localize each steps

Experiments

Script discovery results

Changing a tire		Performing CPR		Repot a plant		Make coffee		Jump car	
GT (11)	$K \leq 10$	GT (7)	$K \leq 10$	GT (7)	$K \leq 10$	GT (10)	$K \leq 10$	GT (12)	$K \leq 10$
get tools out start loose	get tire loosen nut jack car	open airway check pulse	open airway put hand lift chin	take plant put soil loosen roots	remove plant use soil loosen soil	add coffee fill water screw filter	put coffee fill chamber put filter	connect red A connect red B	connect cable charge battery connect end
jack car unscrew wheel remove wheel put wheel screw wheel lower car tight wheel	remove nut take tire lower jack tighten nut	give breath do compressions	do compression open airway do compression give breath	water plant	water plant	put stove see coffee pour coffee	take minutes make coffee see coffee make cup	remove cable A remove cable B	remove cable start car remove cable disconnect cable
Precision Recall	0.9 0.9	Precision Recall	0.4 0.57	Precision Recall	1 0.86	Precision Recall	0.67 0.6	Precision Recall	0.83 0.42

Localizing instruction steps in video



Qualitative results



References

- Bach and Harchaoui. DIFFRAC: A discriminative and flexible framework for clustering. In *NIPS*, 2007.
- Bojanowski et al. Weakly supervised action labeling in videos under ordering constraints. In *ECCV*, 2014.
- Bojanowski et al. Weakly-Supervised Alignment of Video With Text. In *ECCV*, 2015.
- Malmaud et al. What's cookin'? Interpreting cooking videos using text, speech and vision. In *NACL*, 2015.

Check out our project webpage for code/data!

