

Learning from Narrated Videos

Jean-Baptiste Alayrac jbalayrac.com

3rd Workshop on YouTube-8M Large-Scale Video Understanding 28/10/2019

Success of Supervised Learning





Pose estimation

[Towards Accurate Multi-person Pose Estimation in the Wild, Papandreou, Zhu, Kanazawa, Toshev, Tompson, Bregler and Murphy, CVPR17]

Image Segmentation

[Mask R-CNN, He, Gkioxari, Dollár, and Girshisck, ICCV17]

Issues of Supervised Learning





Labels are expensive

Agreement: definition? granularity?

Issues of Supervised Learning



Labels are expensive



Even more problematic for videos!

Weakly supervised learning Use weaker and readily available source of supervision



Training info: image level label

[Barnard et al'03], [Joulin et al'10], [Deselaers et al'12], [Song et al'14], [Wang et al'14], [Cinbis et al'15], [Oquab et al'15], [Kantorov et al'16], [Bilen and Vedaldi'16]...

Weakly supervised learning Use weaker and readily available source of supervision



Training info: video narration (ASR)

[Alayrac et al'16/17], [Malmaud et al, 15], [Sener et al'15], [Huang et al'17], [Zhou et al'17], [Kuehne et al'17],... ...

What are instructional videos?



Depict complex, **goal-oriented** human activities (*e.g. how to change a car tire*)

Multimodal: video and language

San be obtained at **scale** (e.g. on YouTube), without manual annotation

Glossary

Tasks: a complex human activity involving interacting with objects and/or performing multiple small actions.

Example: "make pancakes", "change a car tire", ...

Steps: an atomic action composing a task.

Example: "crack egg", "remove tire", ...

Overview of the talk

1) Leveraging the structure of narrated videos

Making Meringue Pour egg Add sugar Whisk mixture



Making Pancakes Pour mixture





<u>Cross-task weakly supervised learning from</u> <u>instructional videos</u>, Dimitri Zhukov, Jean-Baptiste Alayrac, Ramazan Gokberk Cinbis, David Fouhey, Ivan Laptev, Josef Sivic, *CVPR2019*

2) Leveraging the scale of narrated videos



HowTo100M: Learning a Text-Video Embedding by Watching Hundred Million Narrated Video

Clips, Antoine Miech, Dimitri Zhukov, Jean-Baptiste Alayrac, Makarand Tapaswi, Ivan Laptev, Josef Sivic, *ICCV2019*

Cross-Task Weakly Supervised Learning from Instructional Videos, CVPR19











D. Zhukov*

D. Fouhey

G. Cinbis

I. Laptev

J. Sivic

How much can we leverage the structure in narrated videos and what can we get from that?

What do we mean by structure here?

Task: Make Meringue



Structure within task

What do we mean by structure here?

Task: Make Meringue



Structure within task

Task: Making Pancakes



Task: Making Lemonade





Structure across task

What do we mean by structure here?



Structure within task

Structure across task

Weakly supervised learning of step visual models

Input

→ A set of tasks

ex: *"Make Meringue"*, *"Make Pancakes"*, "Change a car tire", ...

For each task, **a list of steps**:

Make Pancake

- 1) pour egg
- 2) add milk
- 3) whisk mixture

For each task, a set of narrated videos:



"... now we pour the egg ..."

Output What and When?



A visual **classifier** for each step

Localize each steps in all videos



Our assumptions

Temporal ordering. Steps always occur in the order given by the list of steps.



At least once. We assume that for each video, each step occurs once.

Video and narration. Correlation between video and language.

The approach

TL;DR: We jointly (i) learn step classifiers over pretrained visual features and (ii) localize where the steps happen in the video.

Formally: This is done by an alternate optimization between the parameters of the step classifier **(F)** and the localization variable **(Y)** under specific constraints that reflects our assumptions.



Component based model for steps



[Ferrari and Zisserman, 2007], [Farhadi et al, 2009], [Liu et al, 2011], [Yao et al, 2011], ...

The CrossTask dataset

- Designed to assess the benefit of sharing knowledge across tasks:
 - 18 primary tasks, 2750 videos with full temporal annotation
 - 65 related tasks, 1950 videos without annotation
- Diverse set of tasks: Car maintenance, gardening, cooking, home repair



Results: gains from sharing

Localization metric (recall) on CrossTask



Results: novel task transfer

We train only on related task and transfer to the unseen primary task.



Qualitative results

Source Steps From Related Tasks

Cut Steak







AddAdd CherriesTomatoto Cake





Unseen Task: Make French Strawberry Cake

Cut Strawberry



Add Strawberry To Cake





HowTo100M: Learning a Text-Video Embedding by Watching Hundred Million Narrated Video Clips, *ICCV19*







M. Tapaswi



I. Laptev



J. Sivic

J. 51VIC

*equal contribution

A. Miech*

D. Zhukov*

How much can we scale Instructional Video dataset and what can we get from that?

The HowTo100M dataset in numbers

- 23K human tasks scrapped from WikiHow
- 1.2M unique YouTube videos (duration 15 years)
- 136M clips with narration transcribed into text (mostly from ASR)
- Larger than any existing manually annotated captioning dataset

Dataset	Clips	Captions	Videos	Duration	Source	Year
Charades [48]	10k	16k	10,000	82h	Home	2016
MSR-VTT [58]	10k	200k	7,180	40h	Youtube	2016
YouCook2 [67]	14k	14k	2,000	176h	Youtube	2018
EPIC-KITCHENS [7]	40k	40k	432	55h	Home	2018
DiDeMo [15]	27k	41k	10,464	87h	Flickr	2017
M-VAD [52]	49k	56k	92	84h	Movies	2015
MPII-MD [43]	69k	68k	94	41h	Movies	2015
ANet Captions [26]	100k	100k	20,000	849h	Youtube	2017
TGIF [27]	102k	126k	102,068	103h	Tumblr	2016
LSMDC [44]	128k	128k	200	150h	Movies	2017
How2 [45]	185k	185k	13,168	298h	Youtube	2018
HowTo100M	136M	136M	1.221M	134,472h	Youtube	2019

How to collect HowTo100M? Step 1 : WikiHow



Result: list of 130k tasks

How to be healthy How to cook quinoa in a Rice Cooker How to Sew an Apron How to Break a Chain How to April Fool your Girlfriend

Annotation cost: 0

How to collect HowTo100M?

Step 2 : Filter task by verb to keep visual tasks

Result: list of 23k tasks



Annotation cost: 8 hours for Antoine

How to collect HowTo100M? Step 3 : YouTube queries for videos with captions

Result: 1.2 M unique videos

😑 🏼 🖻 YouTube	how to change tire
크는 FILTER	
	How to Change a Tire Change a flat car tire step by step Howdini • 1.9M views • 11 years ago Nothing takes the joy out of a road trip like a flat tire. Do you know how to change it? We didn??t, but we've learned from Allan
E A S Y	How to Change a Tire (plus jacking it up) ChrisFix I 1.2M views • 4 years ago How to take off a car wheel. One of the most fundamental parts of working on a car is properly and safely taking the wheel off and
HOW TO REMOVE & REPLACE A TIRE & TUBE Parti Tool	How to Remove and Install a Bicycle Tire & Tube Park Tool © 1.1M views • 2 years ago This video will guide you through the complete process of removing and installing a tire and tube on a bicycle rim. We will 4K CC

Annotation cost: 0

How to collect HowTo100M? Step 4 : Create clips

Result: 136M narrated clips



14K

Transcript		:	×		
00:18	down then give it a couple good wrap				
00:21	make sure it's full of air worst thing		- 1		
00:24	you want to do is put on a flat spare		- 1		
00:25	then you'll remove the tire careful this				
00:28	can be quite heavy well we have our				
00:30	spare tire in our Jack we've set our				
00:32	reflective warning signal out				
00:34	we've also dropped the wheel on the				
00:36	opposite side of the flat tire so we're				
00:38	ready to start changing it the first				
00:40	thing we'll do is jack the vehicle up				
00:42	you'll have to loosen the jack a little				
English (auto-generated)					

How to Change a Tire | Change a flat car tire step by step

2,216,300 views • Jan 31, 2008

¶ 799 → SHARE =+ SAVE ····



Annotation cost: 0

Learning a visual-text embedding on HowTo100M



DeViSE: A Deep Visual-Semantic Embedding Model, Frome et al. NeurIPS2013

Learning a visual-text embedding on HowTo100M

 $S_{i,j} = S(X_i, Y_j)$ (dot product) $\forall (i, j), \ j \neq i, S_{i,i} > S_{i,j}, S_{i,i} > S_{j,i}$ $L = \frac{1}{R} \sum_{i=1}^{n} \sum_{j=1}^{n} \left[\max(0, m + S_{i,j} - S_{i,i}) + \max(0, m + S_{j,i} - S_{i,i}) \right]$ i=1 $j\neq i$... these nice plants... ... fill pot water fill pot water ...

Evaluation procedure

Text to video retrieval: YouCook2, MSRVTT, LSMDC

Answering the phone





Beauty of having a joint text and video embedding: In both cases, we can evaluate without finetuning!

Within domain: YouCook2 retrieval (YouTube cooking videos)

YouCook2 (R@10)



Trained on HowTo100M

PT HowTo100M + Finetune YouCook2

Within domain: CrossTask action localization

Localization metric (recall) on CrossTask



Out of domain: MSRVTT (popular & generic YouTube videos)

R@10



JSFusion: A joint sequence fusion model for video question answering and retrieval, Yu et al. ECCV2018

Out of domain ++: LSMDC (movies)

R@10



JSFusion: A joint sequence fusion model for video question answering and retrieval, Yu et al. ECCV2018

Coming back to the original question: scale matters!



Online web demo



http://howto100m.inria.fr/

HowTo100M Text-to-Video Search Engine

Enter your search term...



Summary

1) Leveraging the structure of narrated videos

Making Meringue Pour egg Add sugar Whisk mixture



Making Pancakes Pour mixture





<u>Cross-task weakly supervised learning from</u> <u>instructional videos</u>, Dimitri Zhukov, Jean-Baptiste Alayrac, Ramazan Gokberk Cinbis, David Fouhey, Ivan Laptev, Josef Sivic, *CVPR2019*

2) Leveraging the scale of narrated videos



HowTo100M: Learning a Text-Video Embedding by Watching Hundred Million Narrated Video

Clips, Antoine Miech, Dimitri Zhukov, Jean-Baptiste Alayrac, Makarand Tapaswi, Ivan Laptev, Josef Sivic, *ICCV2019*

Future directions

Dealing with the noise. In 50% of the cases, video and narration are not matching. Something should be done!





Still relying on pretrained features (obtained from Kinetics or ImageNet) the
story is not complete.

The dream: end to end learning directly from HowTo100M.