

"A dataset inside a robot's head" by Stable Diffusion

Datasets in Robotics: Past and Future. An open discussion

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TEXAS
Robotics





Meta: Why this Presentation?

- Collaboration in the “Good Systems” team between Roboticists and experts in Information Science and Data Management
- For Information Science and Data Management:
Understanding the use of datasets in Robotics (and other fields) will help creating better information / dataset protocol
- For Robotics:
Explicit discussion and introspective understanding of the reasons and the goals for dataset generation



What is a Dataset?

And some other definitions

Data are observations or measurements (unprocessed or processed) represented as text, numbers, or multimedia.

A **dataset** is a structured collection of data generally associated with a unique body of work.

A **database** is an organized collection of data stored as multiple datasets. Those datasets are generally stored and accessed electronically from a computer system that allows the data to be easily accessed, manipulated, and updated.

Uses of Datasets in AI

Three main purposes



Record an event so that it can be analyzed later →
extract regularities, understand events, gain knowledge



Serve as shared experimental evaluation for solutions to
the same problem → Fair comparison



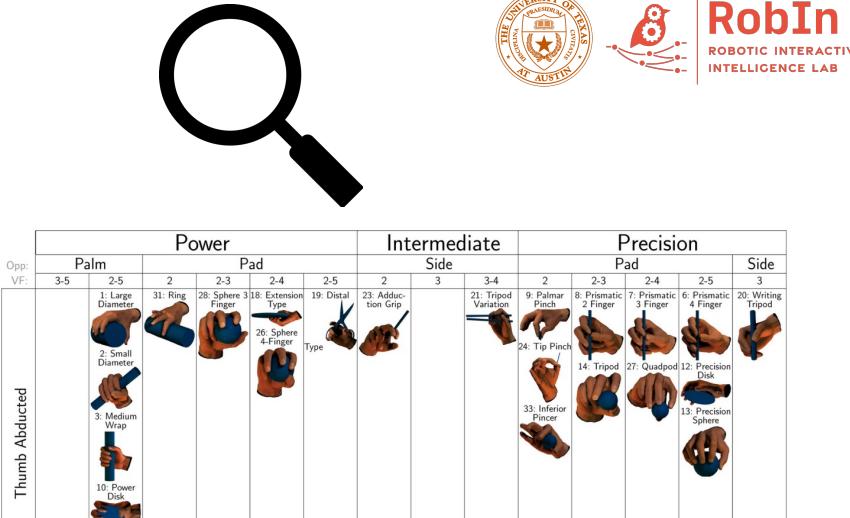
[RECENT!] Train solutions for a problem using data
driven methods

Gaining Knowledge from a Dataset

The Yale human grasping dataset



- Dataset of humans performing common tasks
 - Helps to create a comprehensive taxonomy of possible grasp types
 - The taxonomy is used to analyze frequency and types of grasps in activities



Nr.	Name	Type	Opp. Type	Th. Pos.	VF	Lit. Ref. [54]	Prev. [55]	Mass [g] [57]	Size [cm] [57]	Rigidity [57]	Force [58]
1	Large Diameter	Pow.	Palm	Abd.	VF 2:	2-5	F: 1.7 %	 D: 1.5 %	0 500 1,000 0 5 10 15	ri	wt
					VF 3:	1	D: 0.5 %			sq	
2	Small Diameter	Pow.	Palm	Abd.	VF 2:	2-5	F: 0.7 %	 D: 0.3 %	0 500 1,000 0 5 10 15	ri	wt
					VF 3:	1	D: 0.3 %			sq	in
3	Medium Wrap	Pow.	Palm	Abd.	VF 2:	2-5	F: 12.7 %	 D: 23.8 %	0 500 1,000 0 5 10 15	ri	wt
					VF 3:	1	D: 23.8 %			sq	in
4	Adducted Thumb	Pow.	Palm	Add.	VF 2:	2-5	F: 1.0 %	 D: 0.9 %	0 500 1,000 0 5 10 15	ri	wt
					VF 3:	1	D: 0.9 %				in
5	Light Tool	Pow.	Palm	Add.	VF 2:	2-5	F: 4.8 %	 D: 3.6 %	0 500 1,000 0 5 10 15	ri	wt
					VF 3:	(1)	D: 3.6 %				in
6	Prismatic 4 Finger	Pre.	Pad	Abd.	VF 2:	2-5	F: 4.0 %	 D: 2.0 %	0 500 1,000 0 5 10 15	ri	wt
					VF 3:	1	D: 2.0 %			sq	
7	Prismatic 3 Finger	Pre.	Pad	Abd.	VF 2:	2-4	F: 4.2 %	 D: 3.5 %	0 500 1,000 0 5 10 15	ri	wt
					VF 3:	1	D: 3.5 %				
8	Prismatic 2 Finger	Pre.	Pad	Abd.	VF 2:	2-3	F: 6.4 %	 D: 4.3 %	0 500 1,000 0 5 10 15	ri	wt
					VF 3:	1	D: 4.3 %				
9	Palmar Pinch	Pre.	Pad	Abd.	VF 2:	2	F: 3.9 %	 D: 2.3 %	0 500 1,000 0 5 10 15	ri	wt
					VF 3:	1	D: 2.3 %			fi	

[The Yale human grasping dataset: Grasp, object, and task data in household and machine shop environments, IJRR 2015, Ian M. Bullock, Thomas Feix and Aaron M. Dollar]

Evaluating in Fair Conditions using a Dataset

Setting a common goal for the community



- Dataset of images annotated with category labels
- Organized in a hierarchical taxonomy
- Used to benchmark image recognition solutions



Participation in 2011

96 registrations

15 submissions

Top Entries

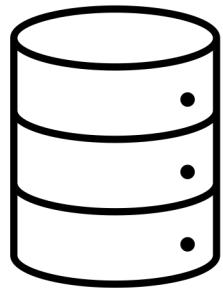
Xerox Research Centre Europe
Univ. Amsterdam & Univ. Trento
ISI Lab Univ. Tokyo
NII Japan

[Deng, Jia, Wei Dong, Richard Socher, Li-Jia Li, Kai Li, and Li Fei-Fei. "Imagenet: A large-scale hierarchical image database." In 2009 IEEE conference on computer vision and pattern recognition, pp. 248-255. ieee, 2009.]

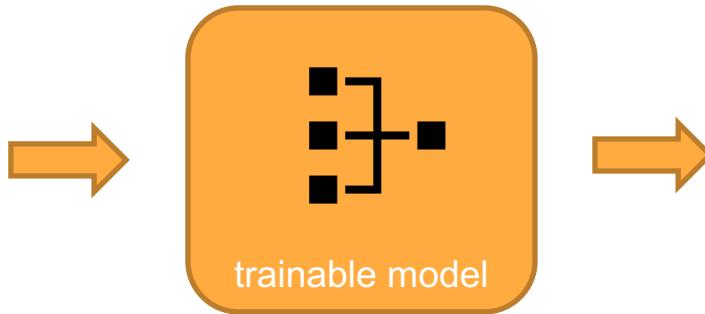


Training Machine Learning Based Solutions

A recent(-ish) trend



dataset

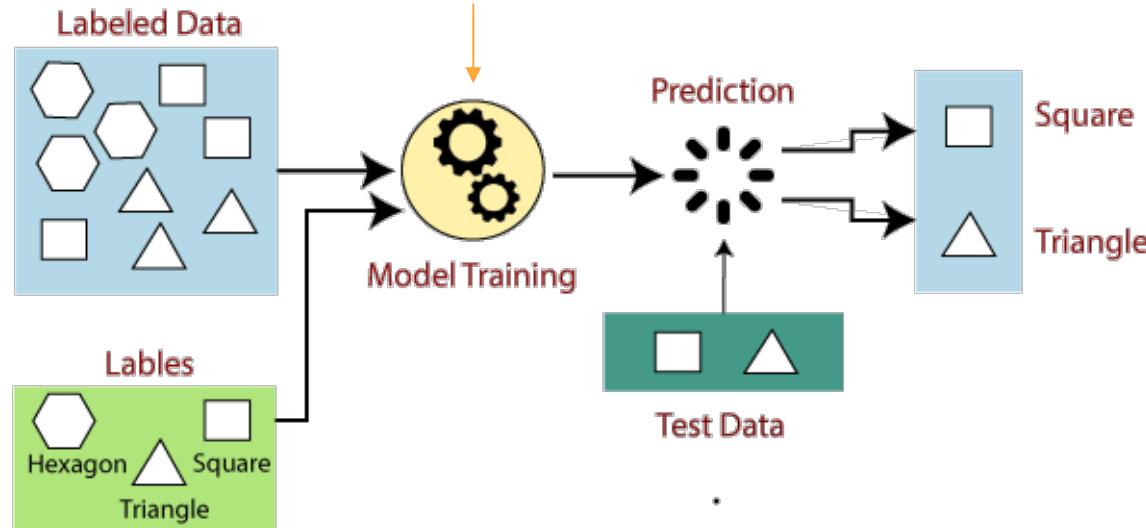


Background: Supervised Learning

How to consume datasets with machine learning

(x, label)

$$f(x) = \text{label}$$

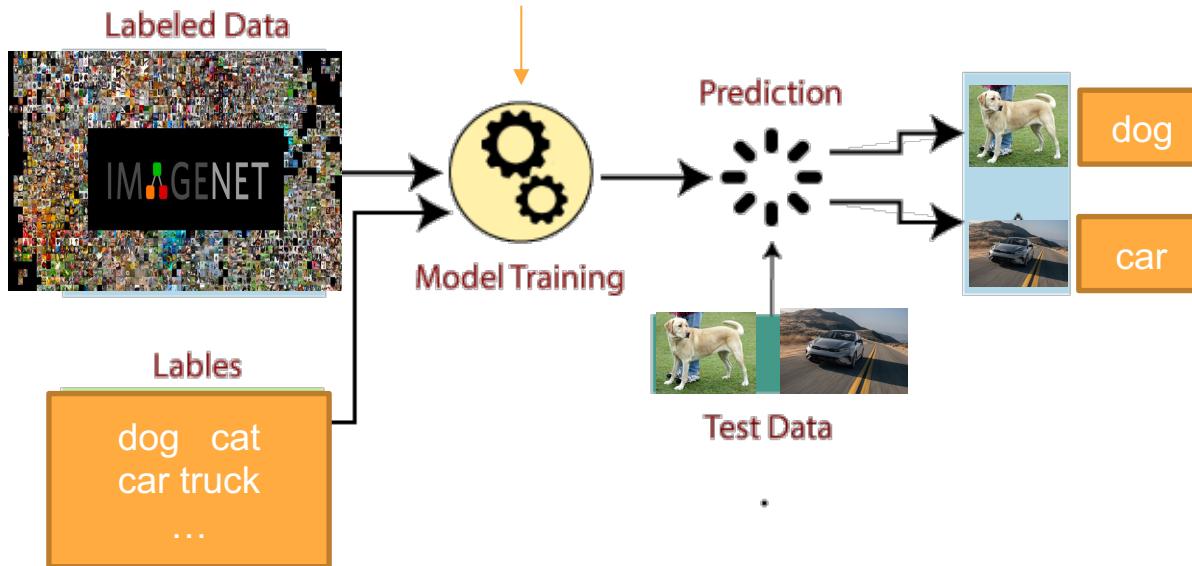


Supervised Learning + Datasets

Encode the information in the dataset into a model

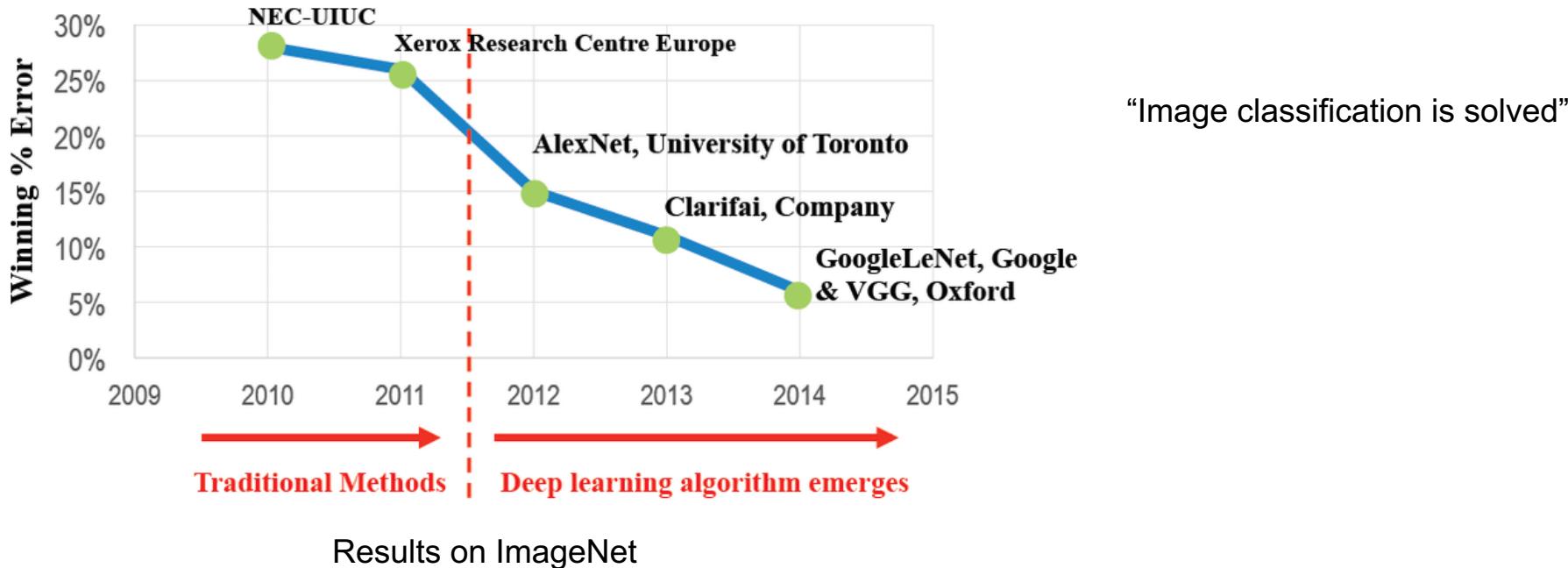
$$f(x) = \text{label}$$

(x, label)



The Dataset Revolution in Computer Vision

Data-driven methods win over classical (hardcoded) solutions



The Ingredients of the Revolution

It is not (just) the algorithm; it is the data(set)!

- Large Dataset
- Model with enough capacity → Deep Neural Network
- Hardware to train the large model using the large dataset



[Krizhevsky, Alex, Ilya Sutskever, and Geoffrey E. Hinton. "Imagenet classification with deep convolutional neural networks." Communications of the ACM 60, no. 6 (2017): 84-90.]

Image Datasets

The fuel for the AI revolution



Image Classification



Image Segmentation



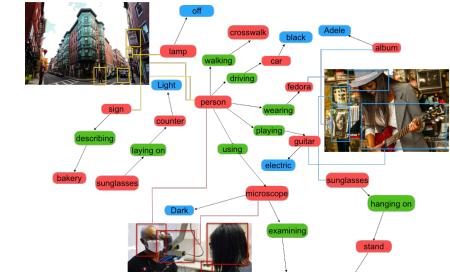
Scene Understanding



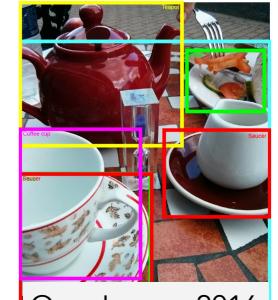
Static



Passive



VisualGenome, 2016 [Krishna et al.]



Extending the First Successes to other CV Tasks

Bringing in the temporal dimension



Activity Recognition



Motion Understanding



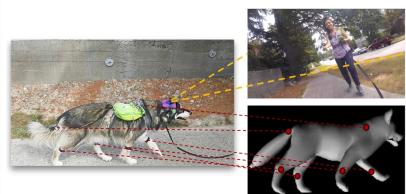
Human-Object Interaction



Dynamic



Passive



Decade-Dog, 2018 [Ehsani et al.]



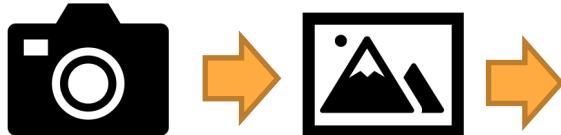
Charades-Ego, 2018 [Sigurdsson et al.]



Meanwhile, in Robotics...

Difficulties to find common ground

In Computer Vision:



- Image Classification
- Object Detection
- Segmentation
- Action Recognition
- Optical Flow Estimation
- ...

In Robotics:



Robotics: Active Datasets?

From datasets to simulation environments?



Robot Task Learning



Benchmark for Solutions



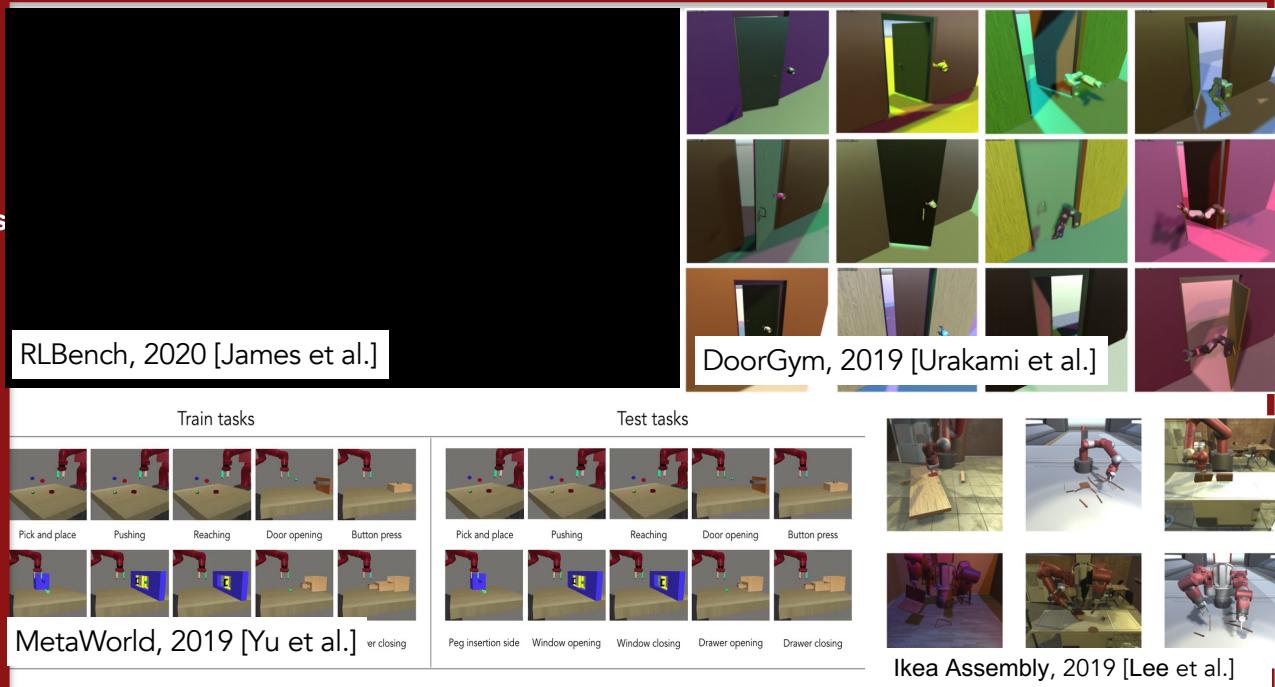
Task-Specific



Dynamic



Active



The slide displays a grid of simulation environments for robotics research, categorized by their characteristics:

- Robot Task Learning:** Represented by a large black square placeholder.
- Benchmark for Solutions:** Represented by a large black square placeholder.
- Task-Specific:** Examples shown include RLBench (2020 [James et al.]), DoorGym (2019 [Urakami et al.]), MetaWorld (2019 [Yu et al.]), and Ikea Assembly (2019 [Lee et al.]).
- Dynamic:** Examples shown include MetaWorld and Ikea Assembly.
- Active:** Examples shown include MetaWorld and Ikea Assembly.

Specific tasks shown in the environments include:

- RLBench, 2020 [James et al.]:** Train tasks: Pick and place, Pushing, Reaching, Door opening, Button press; Test tasks: Pick and place, Pushing, Reaching, Door opening, Button press.
- DoorGym, 2019 [Urakami et al.]:** Train tasks: Pick and place, Pushing, Reaching, Door opening, Button press; Test tasks: Peg insertion side, Window opening, Window closing, Drawer opening, Drawer closing.
- MetaWorld, 2019 [Yu et al.]:** Train tasks: Pick and place, Pushing, Reaching, Door opening, Button press; Test tasks: Pick and place, Pushing, Reaching, Door opening, Button press.
- Ikea Assembly, 2019 [Lee et al.]:** Train tasks: Pick and place, Pushing, Reaching, Door opening, Button press; Test tasks: Pick and place, Pushing, Reaching, Door opening, Button press.

Some Examples of Robotics Datasets

Two types of Datasets

Datasets that are useful for Robots



20Bn-smth-smth

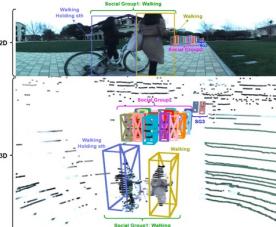


RBO Dataset

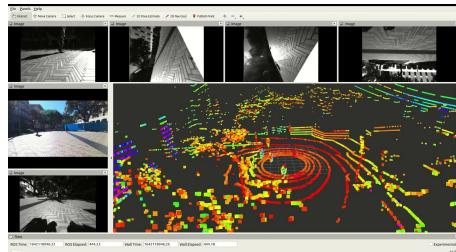


YCB Dataset

Datasets from Robots



JRDB



ScanD



RoboTurk
Real Robot
Dataset

111 hours of robot demonstrations
1 week of data collection
3 dexterous manipulation tasks
54 non-expert users
2144 demonstrations



What does a Good Dataset in Robotics need?

Meta information

Robot Embodiment:

- Actuation
- Sensing
- Morphology



Environment:

- Layout
- Type
- Other agents
- Context



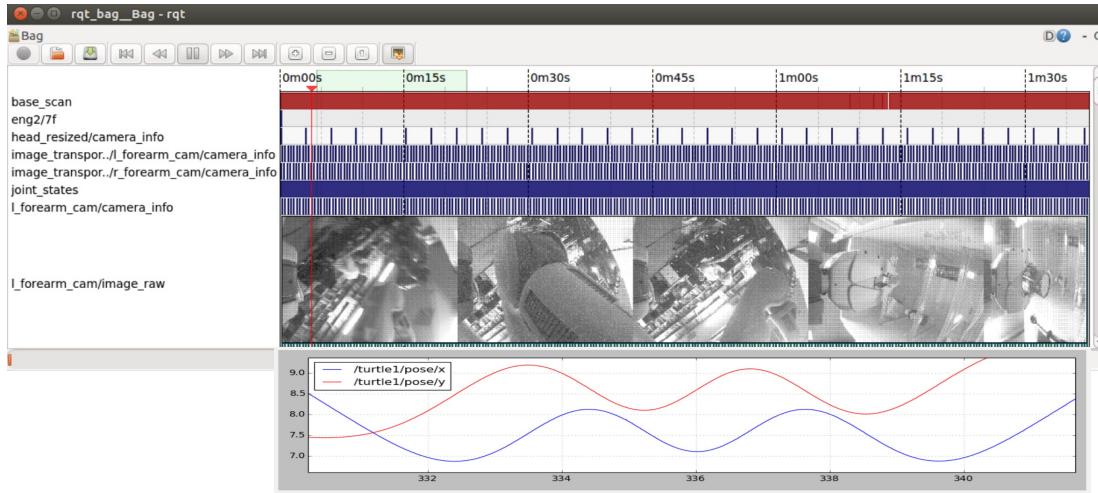
Event:

- Task/Goal
- Objectives
- Conditions



What does a Good Dataset in Robotics need?

Temporal information



- Timestamps of each signal
 - ideally synced
- Values in a protocoled format
 - Careful with compression!
 - Better raw
- Annotations



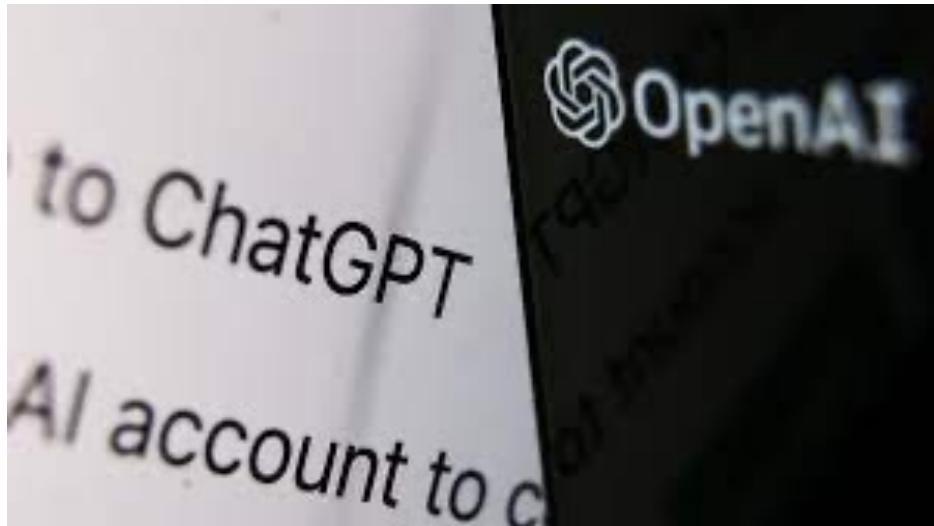
Resources

Where to find datasets that are relevant for Robotics

- IJRR Data Papers: <https://journals.sagepub.com/topic/collections-ijr/ijr-3-datapapers/ijr>
 - "Dry" description of the datasets with focus on reproducibility
- NeurIPS Dataset track: <https://neurips.cc/Conferences/2022/CallForDatasetsBenchmarks>
 - New track focused on Datasets and Benchmarks for Robotics and AI in general
- Awesome Datasets for Robotics: <https://github.com/mint-lab/awesome-robotics-datasets>
 - Collection of links collected by a lab in Korea

The Future in Other Fields

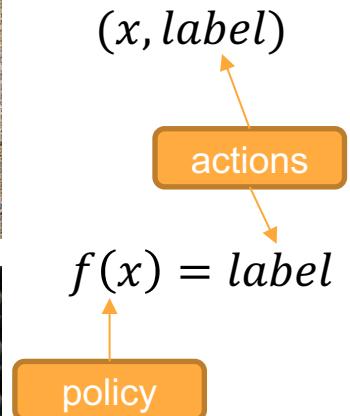
Removing the need for (costly) annotations



"The University of Texas at [redacted] (UT Austin, UT, or Texas) is a public research [redacted] in Austin, Texas. It was founded in 1883 and is the oldest [redacted] in the University of [redacted] System.

The Future in Robotics

Learning to act from large datasets?





Summary

- Datasets have three usages in AI/Robotics
 - Analyze/understand events
 - Benchmark solutions
 - Train solutions with ML
- This last usage is becoming more and more important in robotics
- Robotics datasets are harder than in CV due to the unclear/variable interface
- Datasets will still play an important role in AI/robotics but researchers want to avoid manual annotations

Open Questions:

- What type of datasets did/do you use for robotics?
 - What data
 - What annotations
 - What size
- What do you think are the main current and future uses of datasets in robotics?
- What information do our non-robotics colleagues need from us when helping us creating datasets?

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