



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

# Datasets in Robotics: Past and Future. An open discussion

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ROBOTIC INTERACTIVE  
INTELLIGENCE LAB

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.

# What is a Dataset?

“A Dataset if an opportunity  
for the research community  
to set a North Star”



# 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



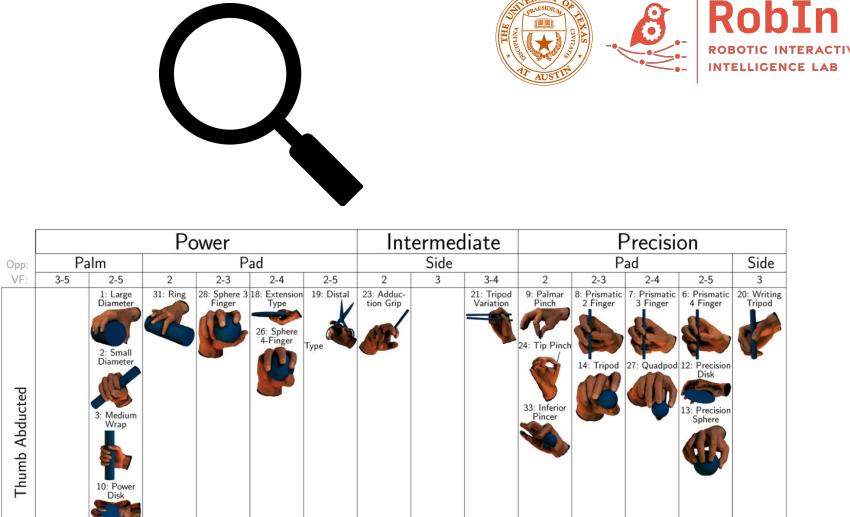
[RECENT!] Train solutions for a problem using data  
driven methods

# 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:					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:					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:					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					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)					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:					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:						
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:						
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:					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

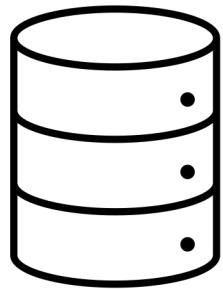
Xerox Research Centre Europe  
Univ. Amsterdam & Univ. Trento  
ISI Lab Univ. Tokyo  
NII Japan

[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.]

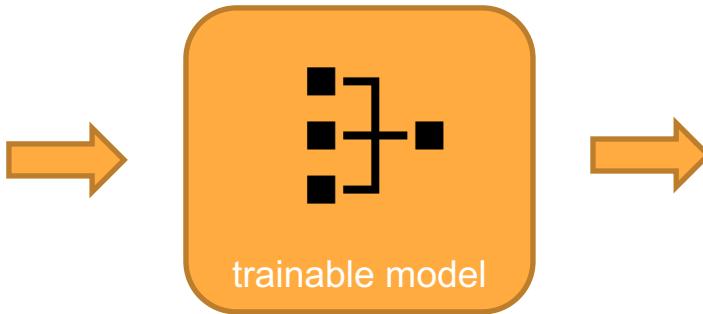


# Training Machine Learning Based Solutions

A recent(-ish) trend



dataset

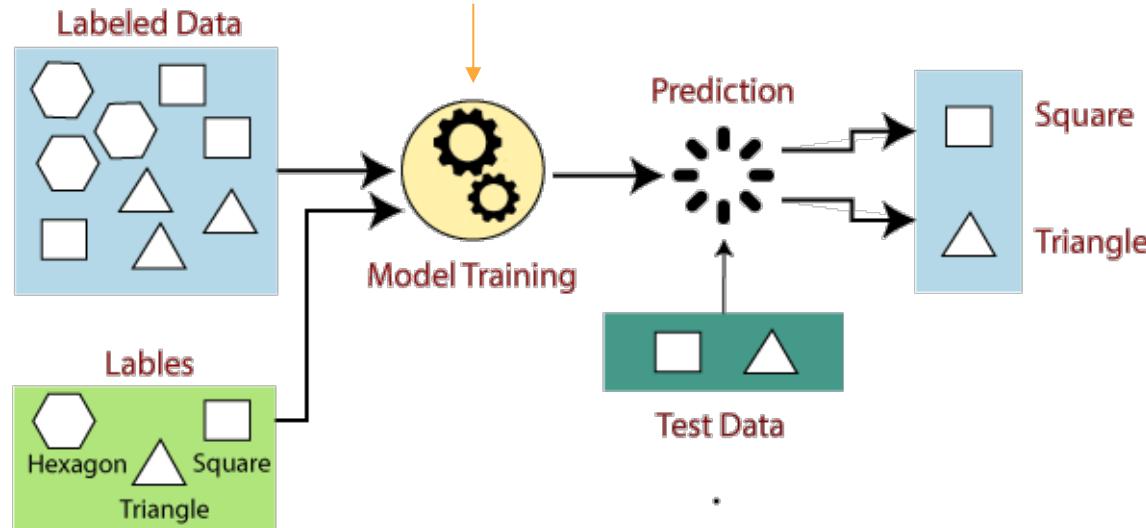


# Background: Supervised Learning

How to consume datasets with machine learning

$(x, \text{label})$

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

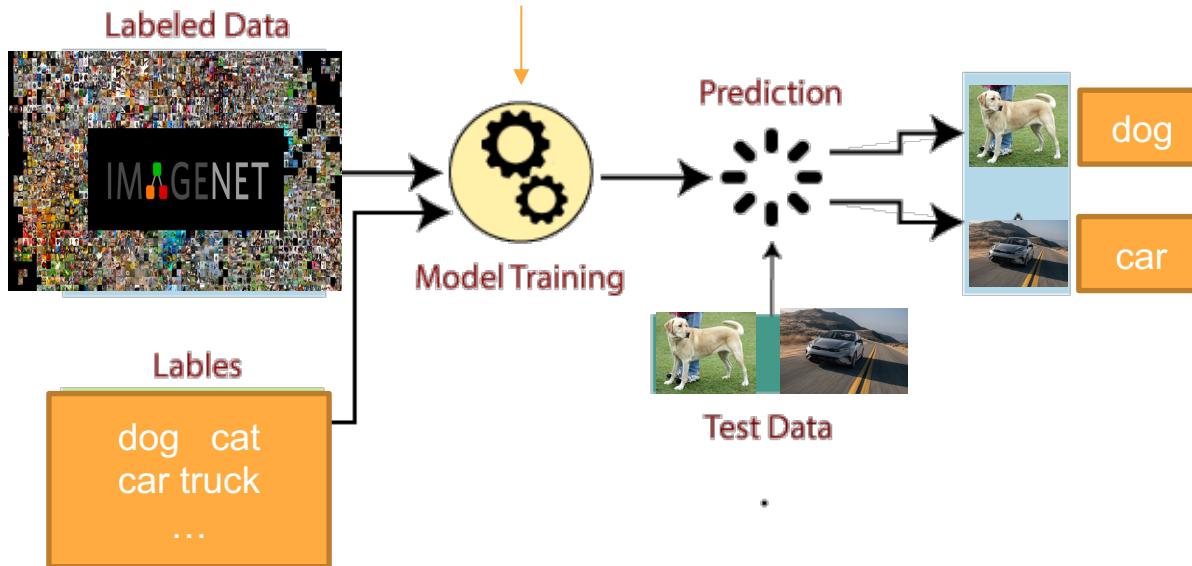


# Supervised Learning + Datasets

Encode the information in the dataset into a model

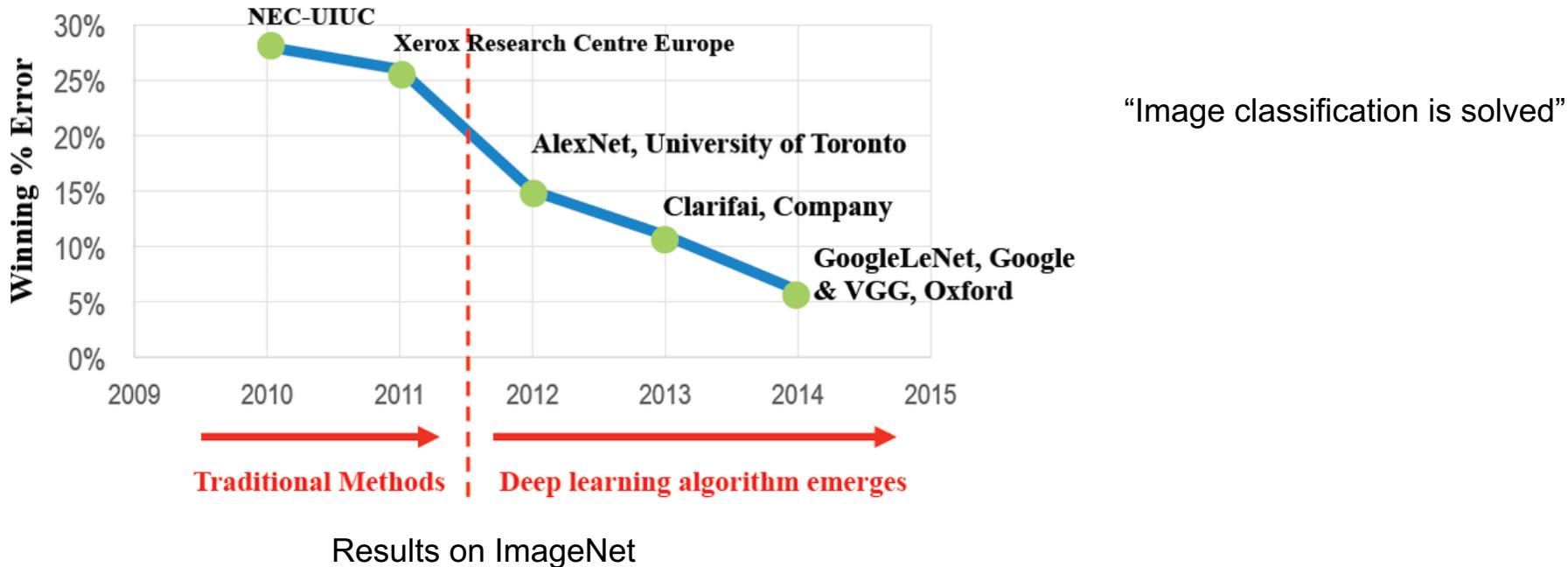
$$f(x) = \text{label}$$

$(x, \text{label})$



# 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



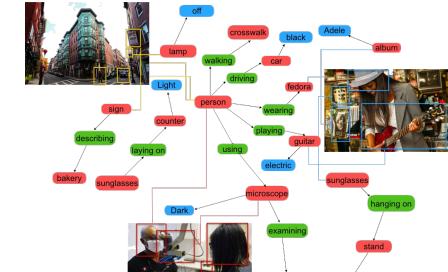
Scene Understanding



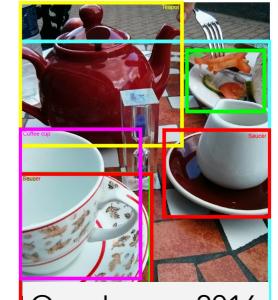
Static



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



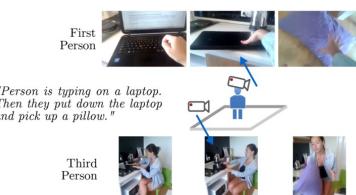
Dynamic



Passive



Decade-Dog, 2018 [Ehsani et al.]



Charades-Ego, 2018 [Sigurdsson et al.]

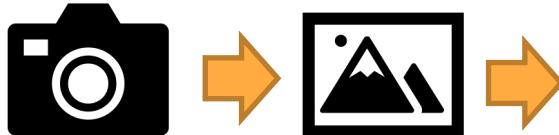


Grasping Dataset, 2014 [Bullock et al.]

# 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



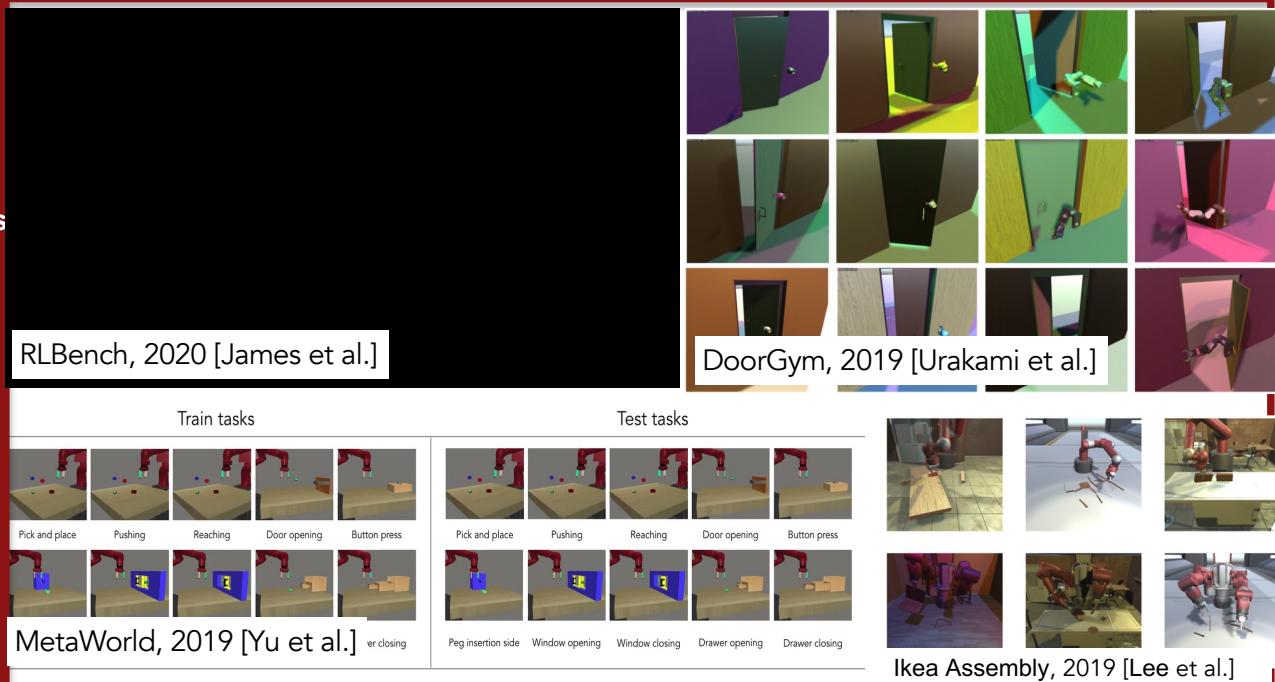
Task-Specific



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

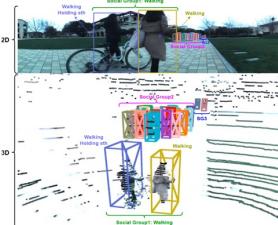


RBO Dataset

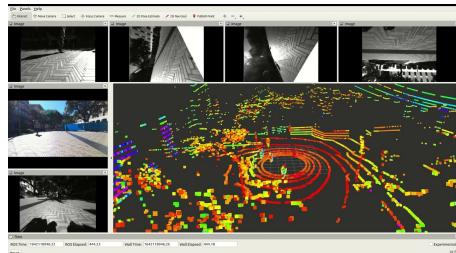


YCB Dataset

Datasets from Robots



JRDB



ScanD



111 hours of robot demonstrations  
**RoboTurk**  
1 week of data collection  
**Real Robot Dataset**  
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



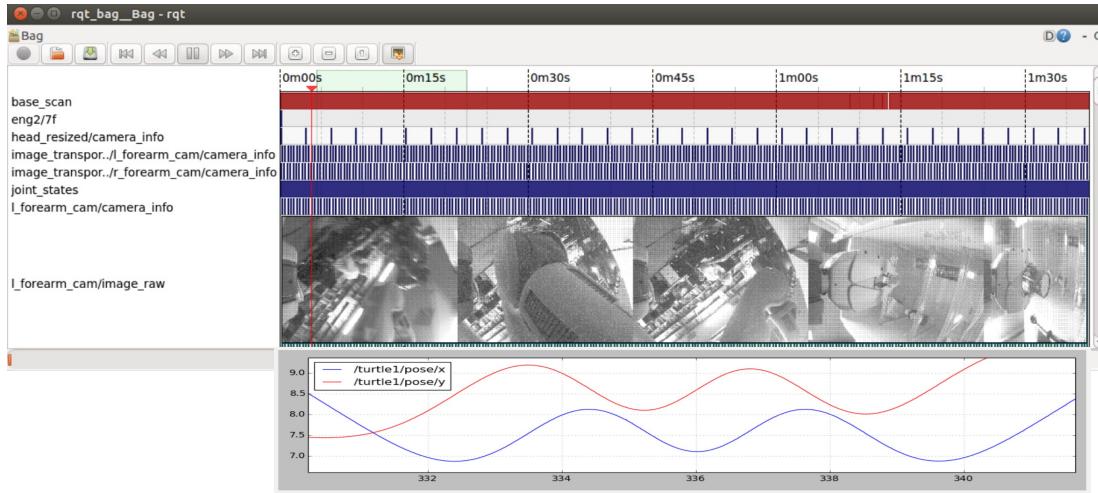
Event:

- 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



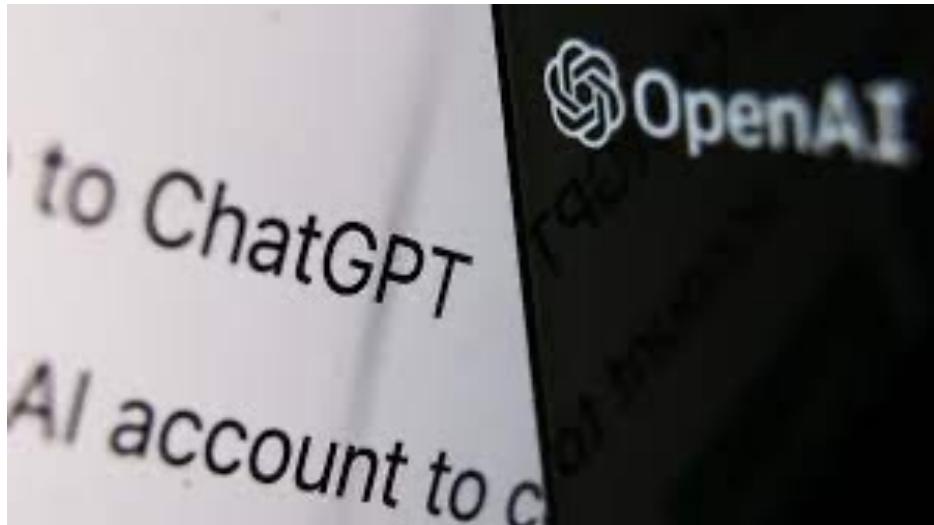
# Resources

*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

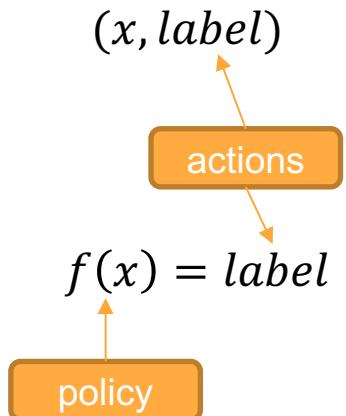
*Removing the need for (costly) annotations*



"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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