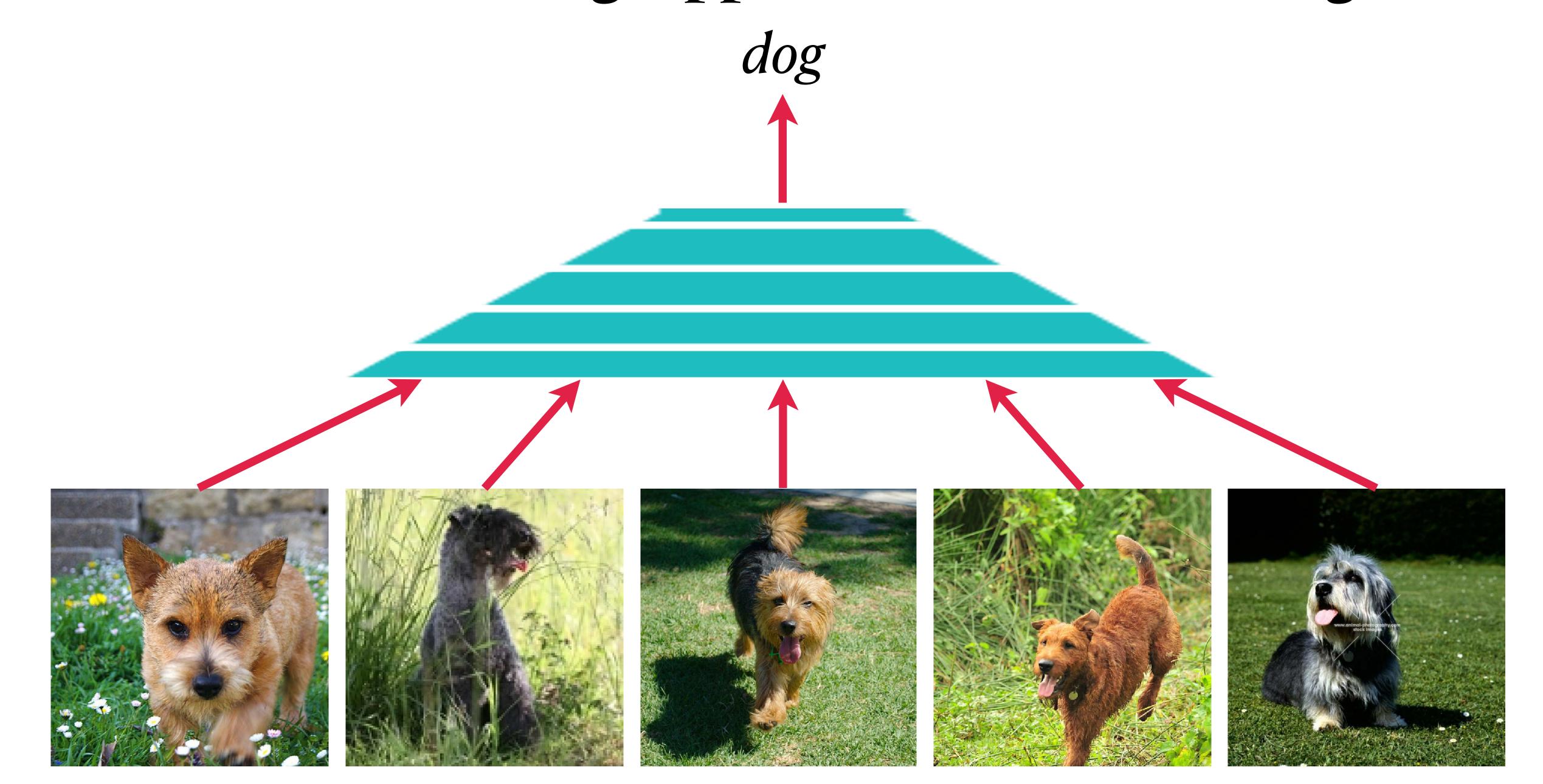
## Introduction to EECS 598: Action and Perception

Stella Yu

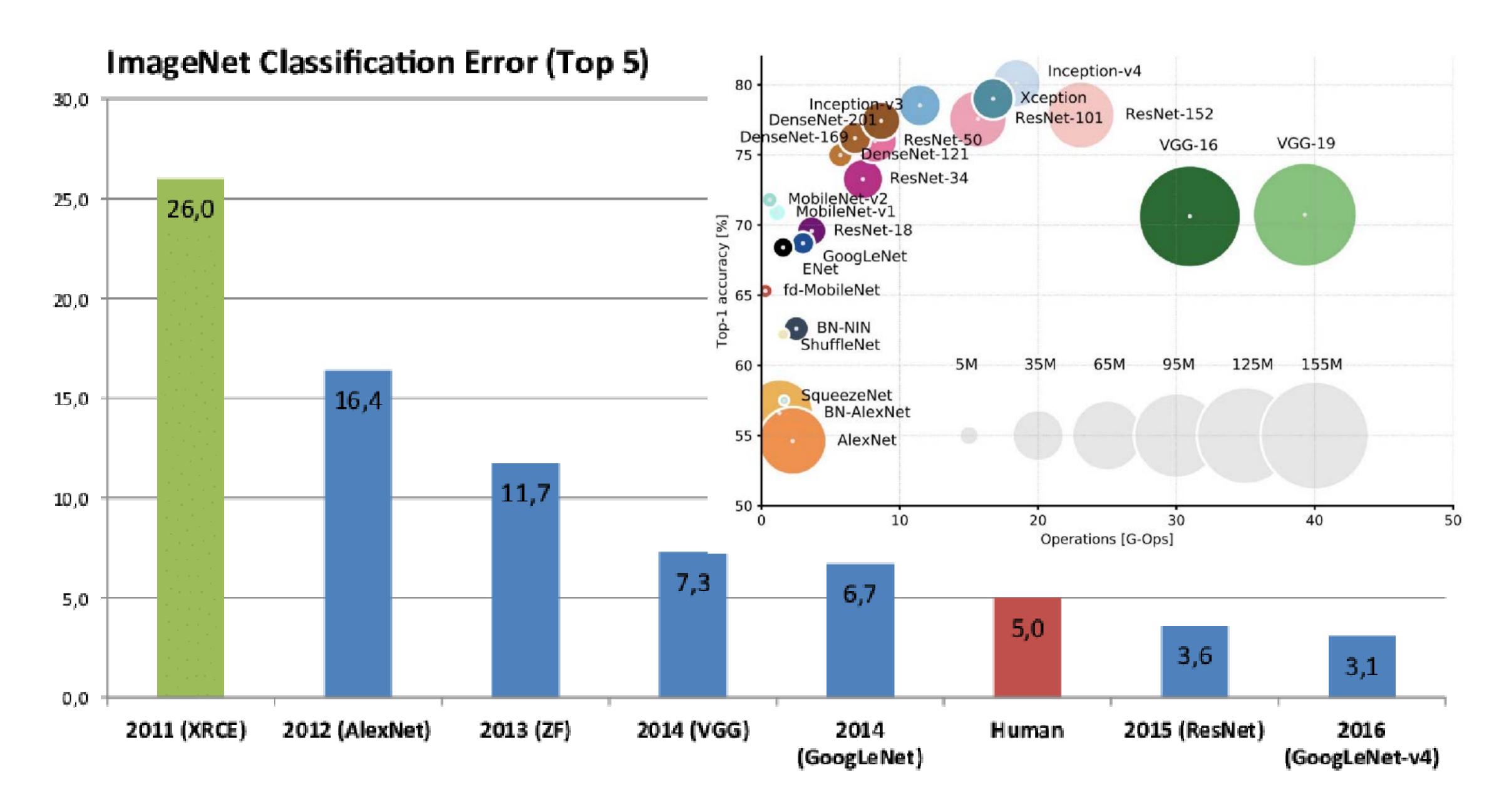
University of Michigan

4 January 2023

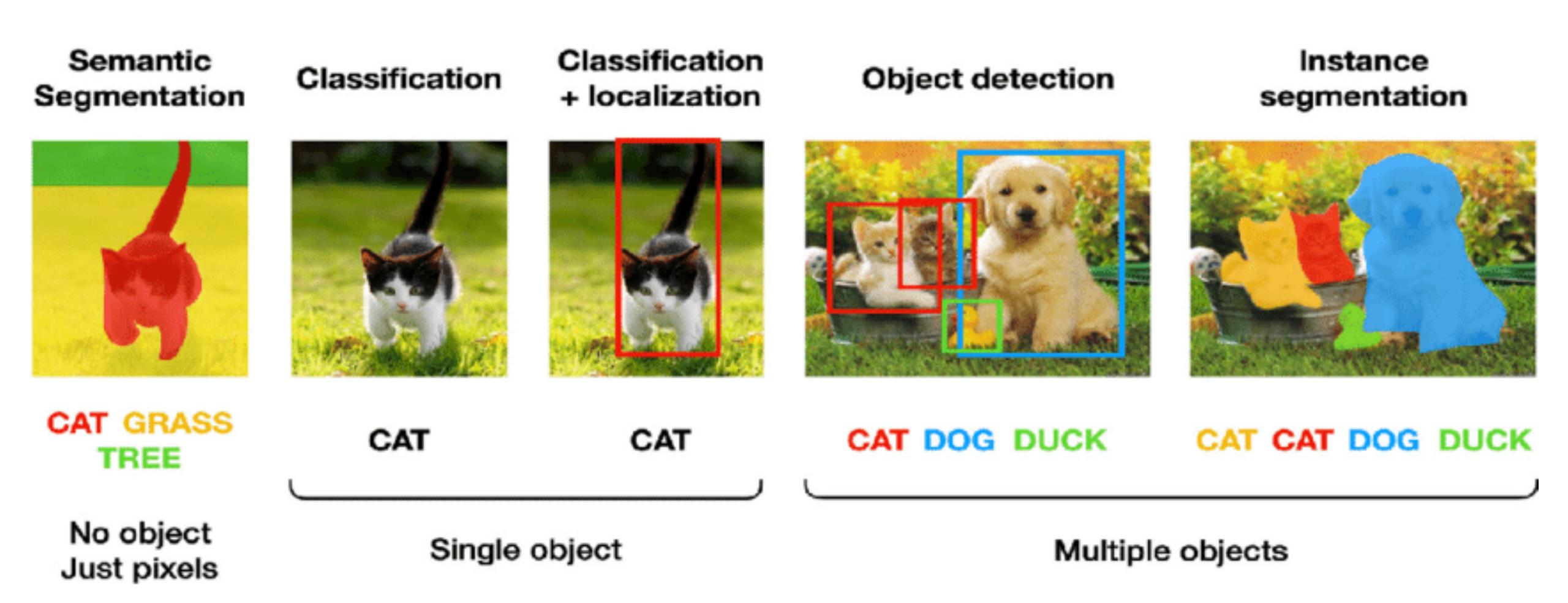
## End-to-End Learning Approach to Visual Recognition



## Supervised Learning Reaches Superhuman Performance



# Deep Learning Is Successful But Too Specialized



High-level: Models trained for one task do not work for another.

## Deep Learning Is Successful But Too Specialized



Low-level: Models trained on one data kind do not work for another.

## Deep Learning Is Successful But Too Specialized

"A Car Parked On The Side of The Road"



Low-level: Models trained on one data kind do not work for another.

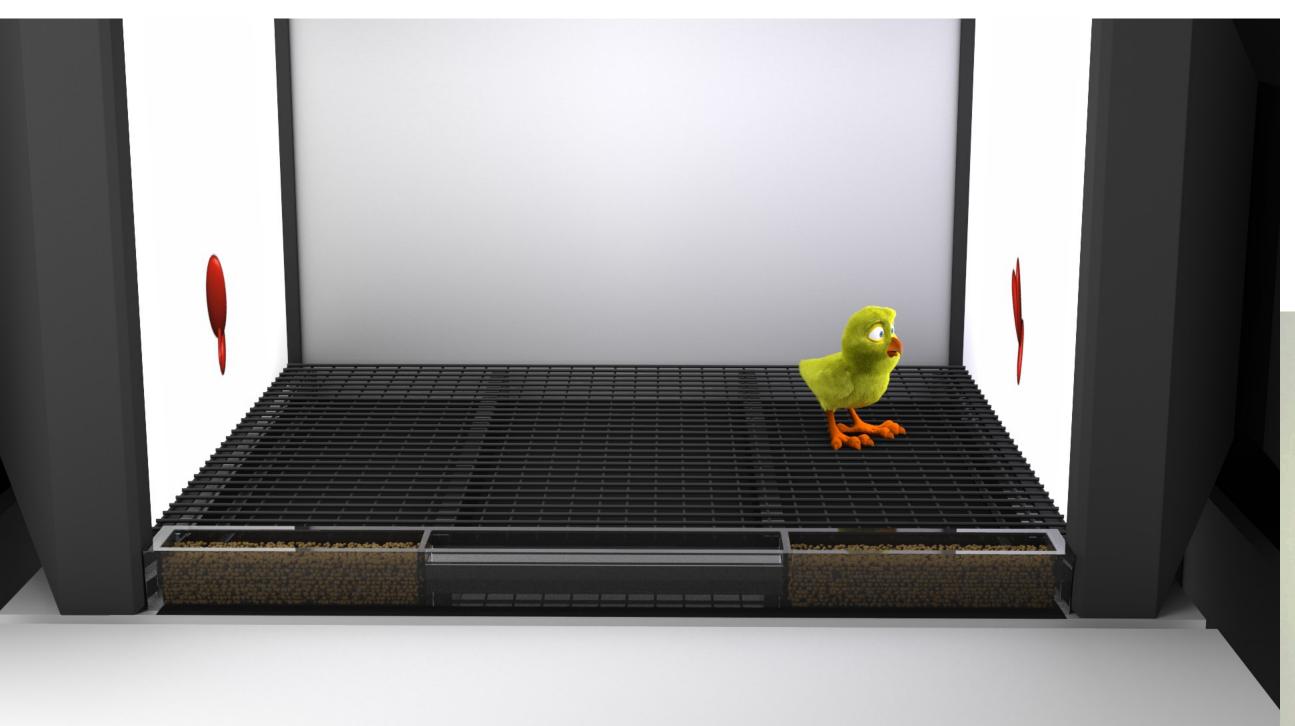
## Natural Learning of Vision: No Semantic Supervision



Linda Smith: [Jayaraman S. et al, PLoS ONE 2015; Clerkin et al, TRSB 2017; Slone et al, DS 2019]

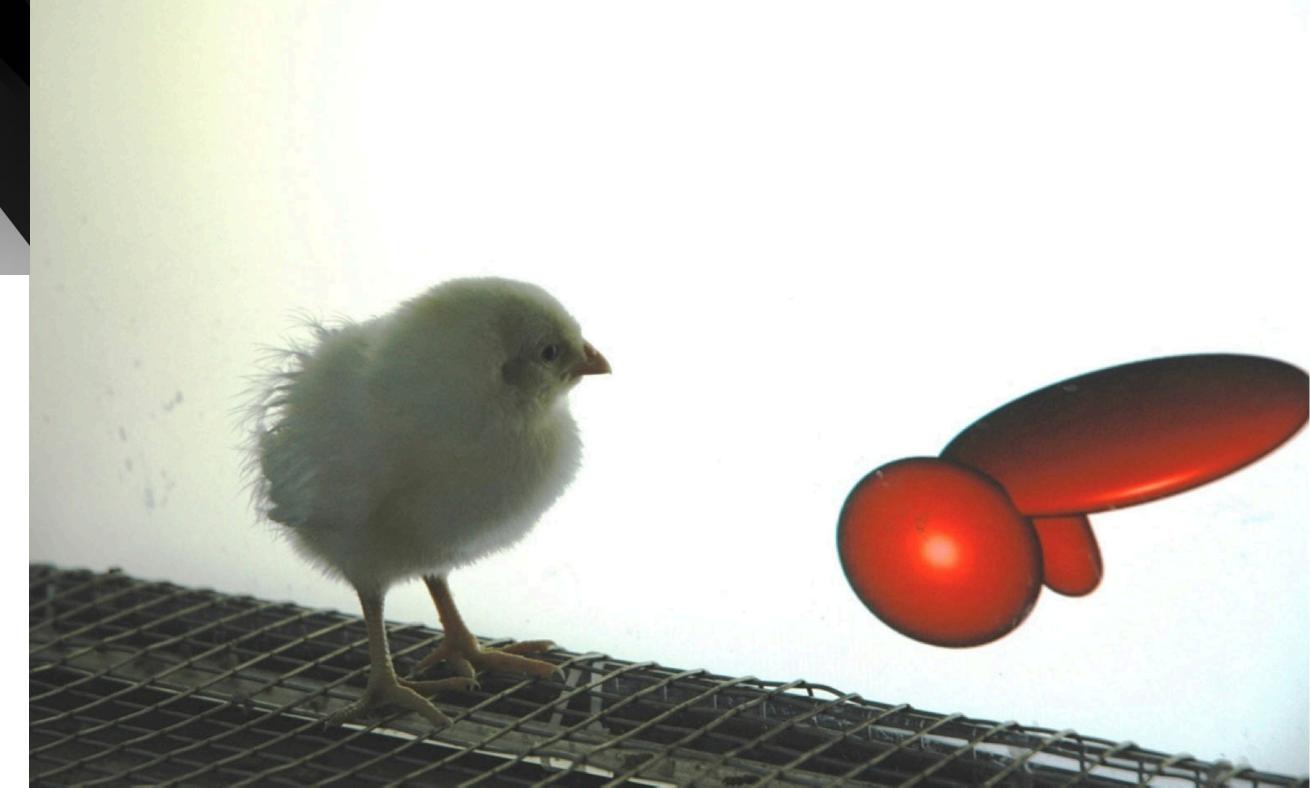


## Newborn Visual Recognition from Slow Smooth Videos

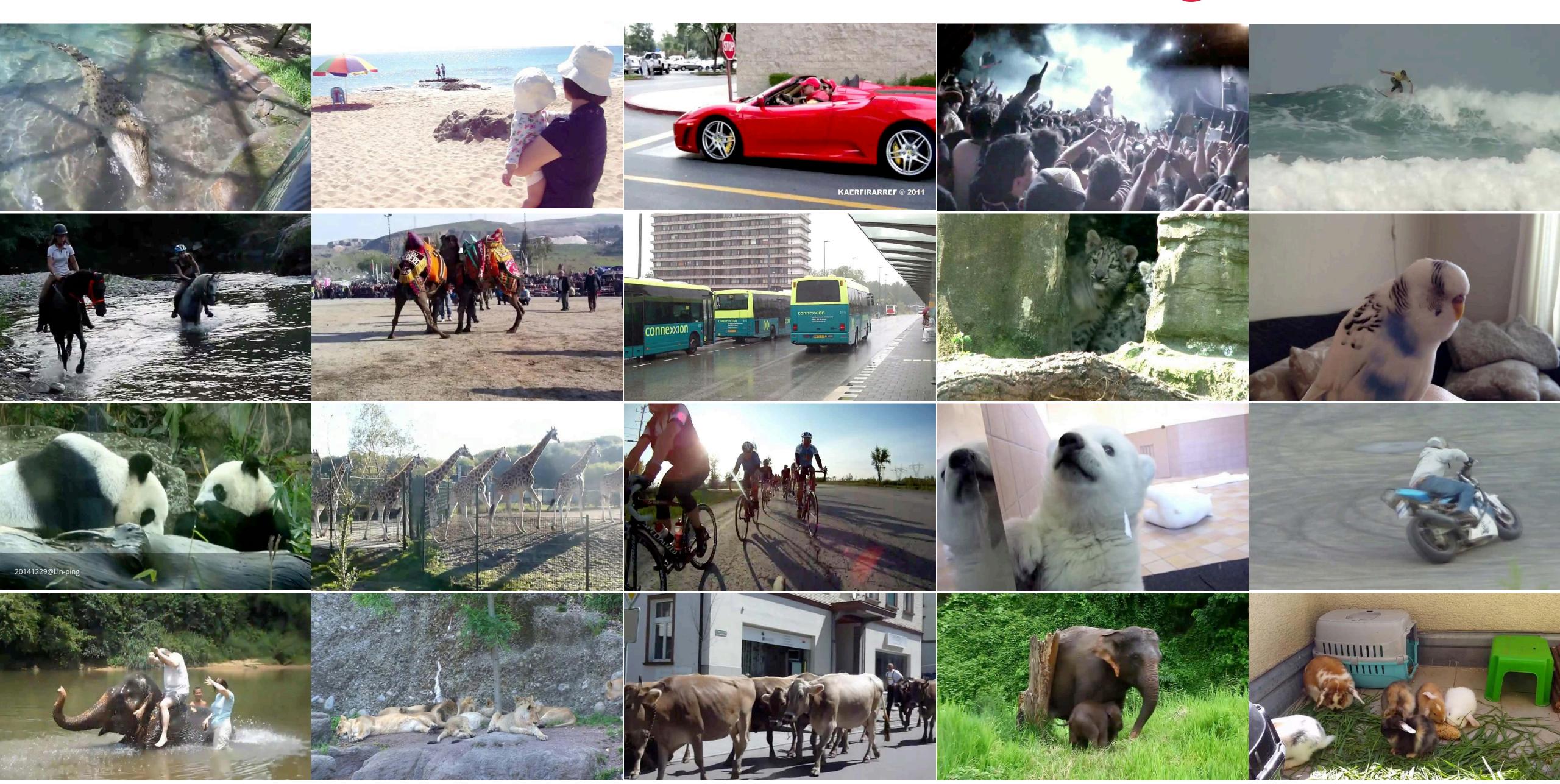


[Justin Wood et al, 2016]: A smoothness constraint on the development of object recognition.

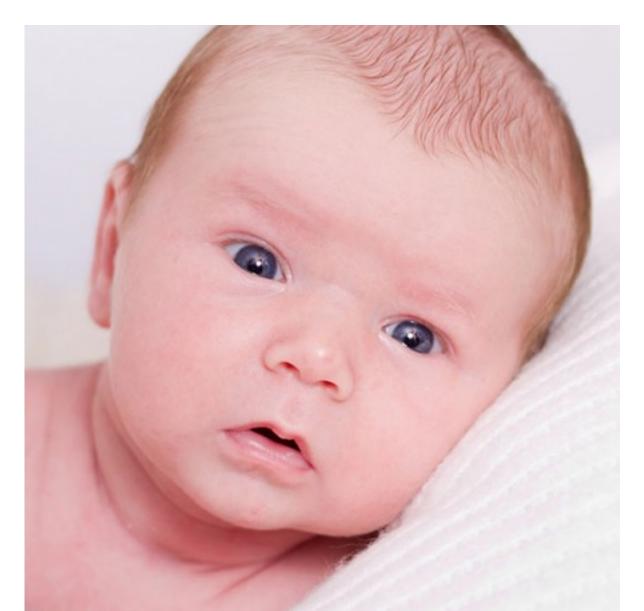
The development of newborn object recognition in fast and slow visual worlds.



# What Can A Model Learn from Nothing but Data?



# Baby Vision vs. Grown Vision









# From Undivided Sensation to Bounded Rationality





blooming buzzing confusion

-William James



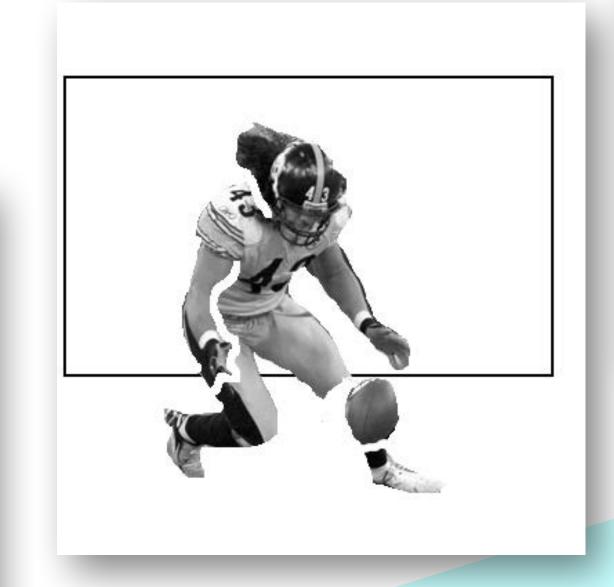
## Bottom-Up Approach to Visual Recognition

universal but brittle

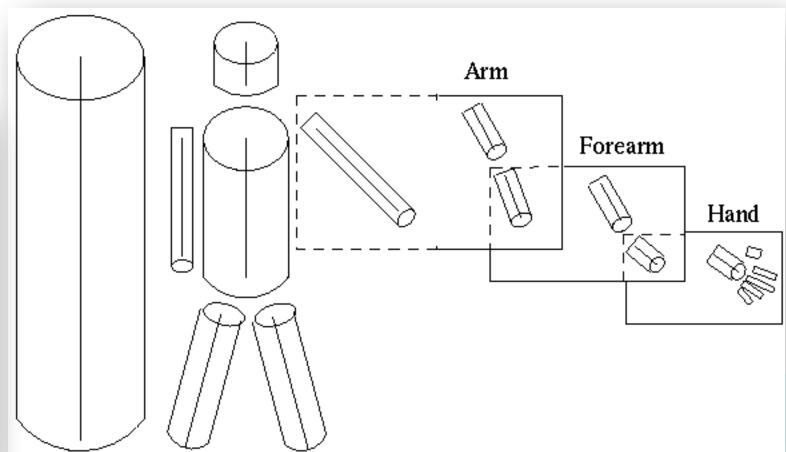
primal sketch



2.5D-sketch



3D models



high-level

mid-level,

The Year Park In the Ye

image

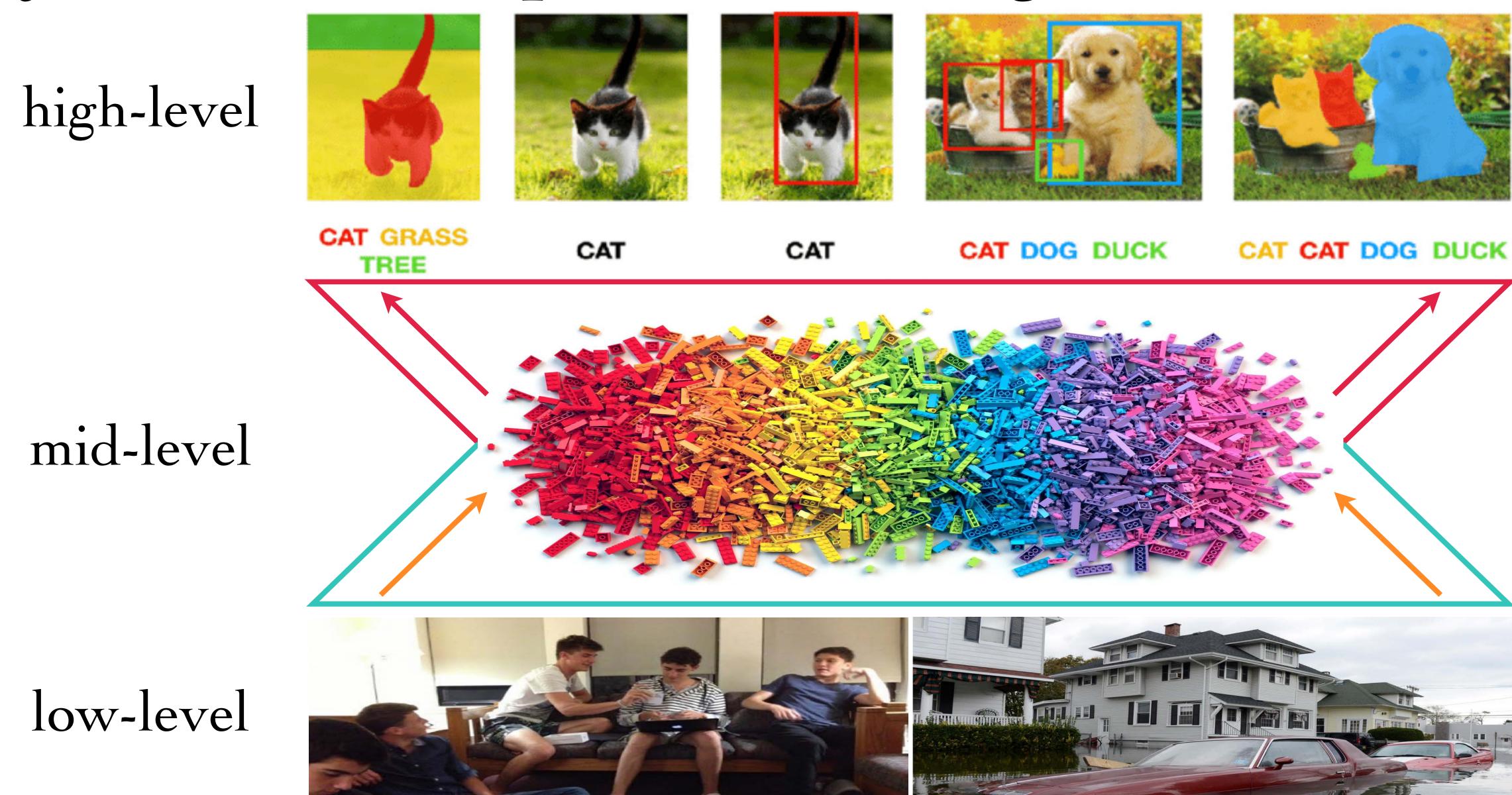
10W-level

[Marr, 1982]

## My Research: Unsupervised Learning of Mid-Level Vision



## My Research: Unsupervised Learning of Mid-Level Vision



#### Unsupervised Learning: Segment Objects, Differentiate Them, Parse into Parts

[ AMD, NeurIPS 2021; NPID, CVPR 2018; CLD, CVPR 2021; HSG, CVPR 2022]



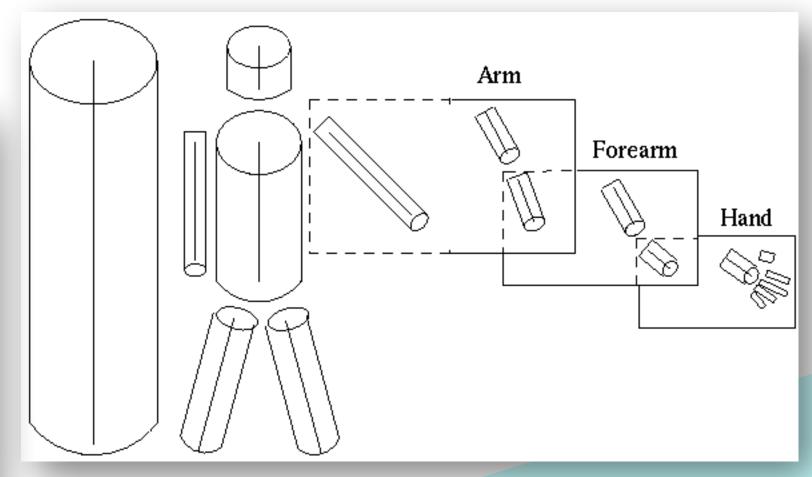
# Mid-Level Vision Is the Key

Jow-level 2.5D-sketch

primal sketch







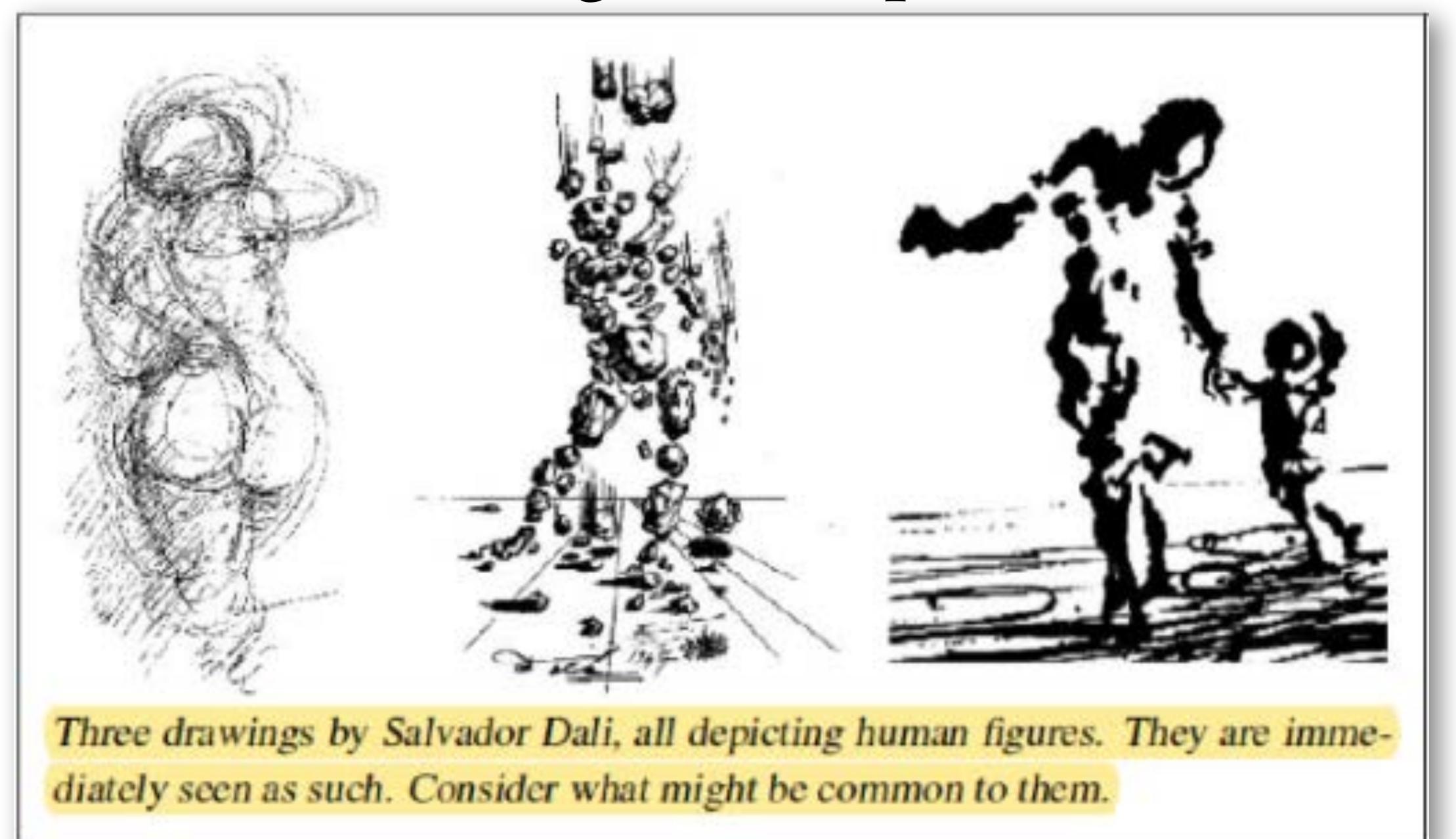
high-level

mid-level

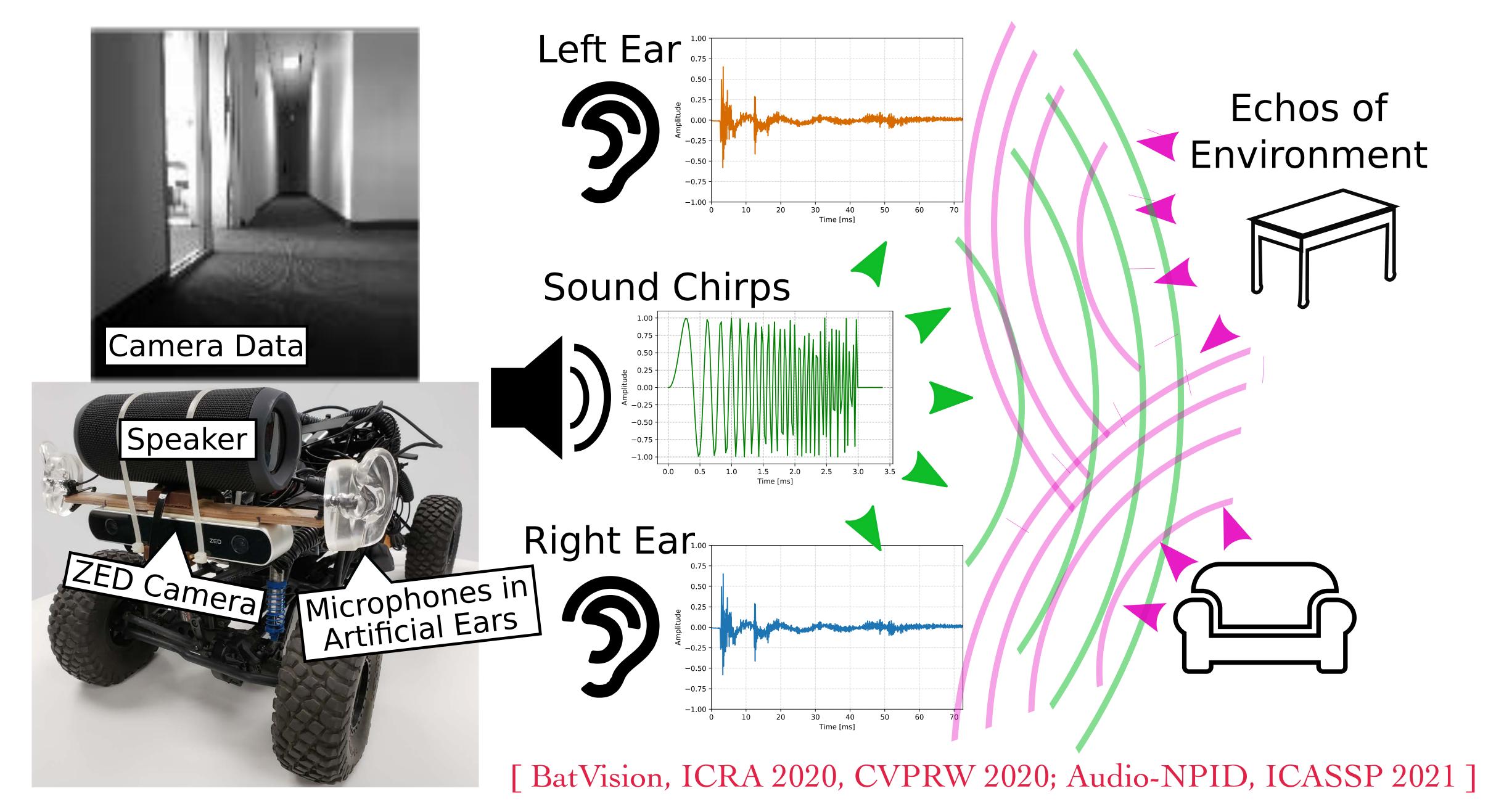




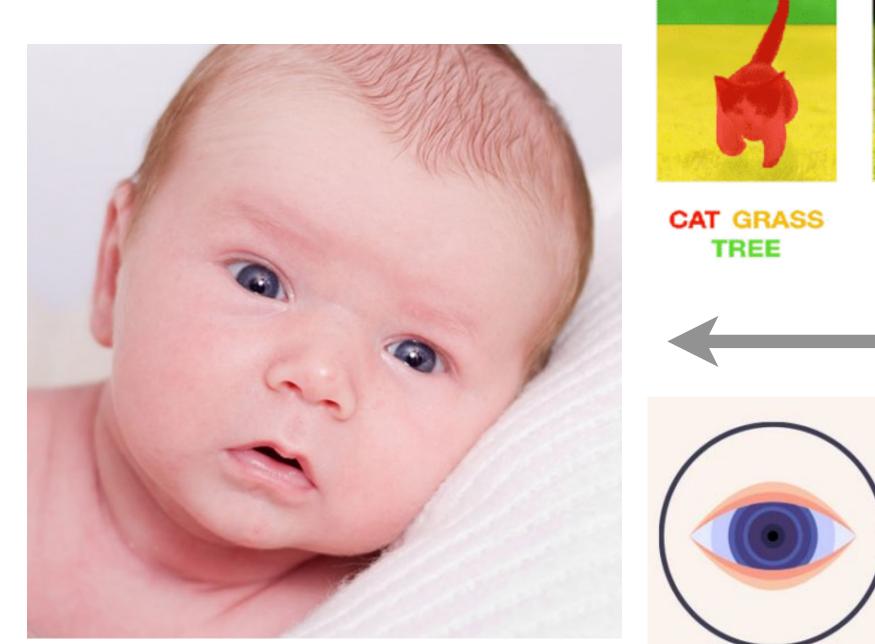
J. J. Koenderink: Edges Are Imposed, Not Detected

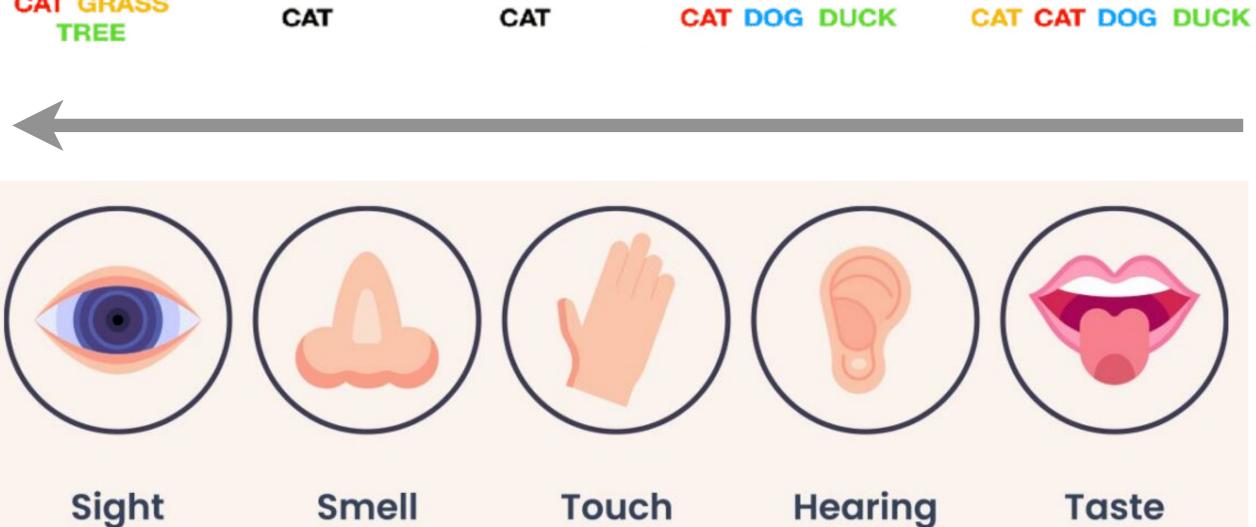


### Unsupervised Ambient Sound Recognition for Localization / Navigation



## Objective: What Is A Baby Supposed to Learn?







Model

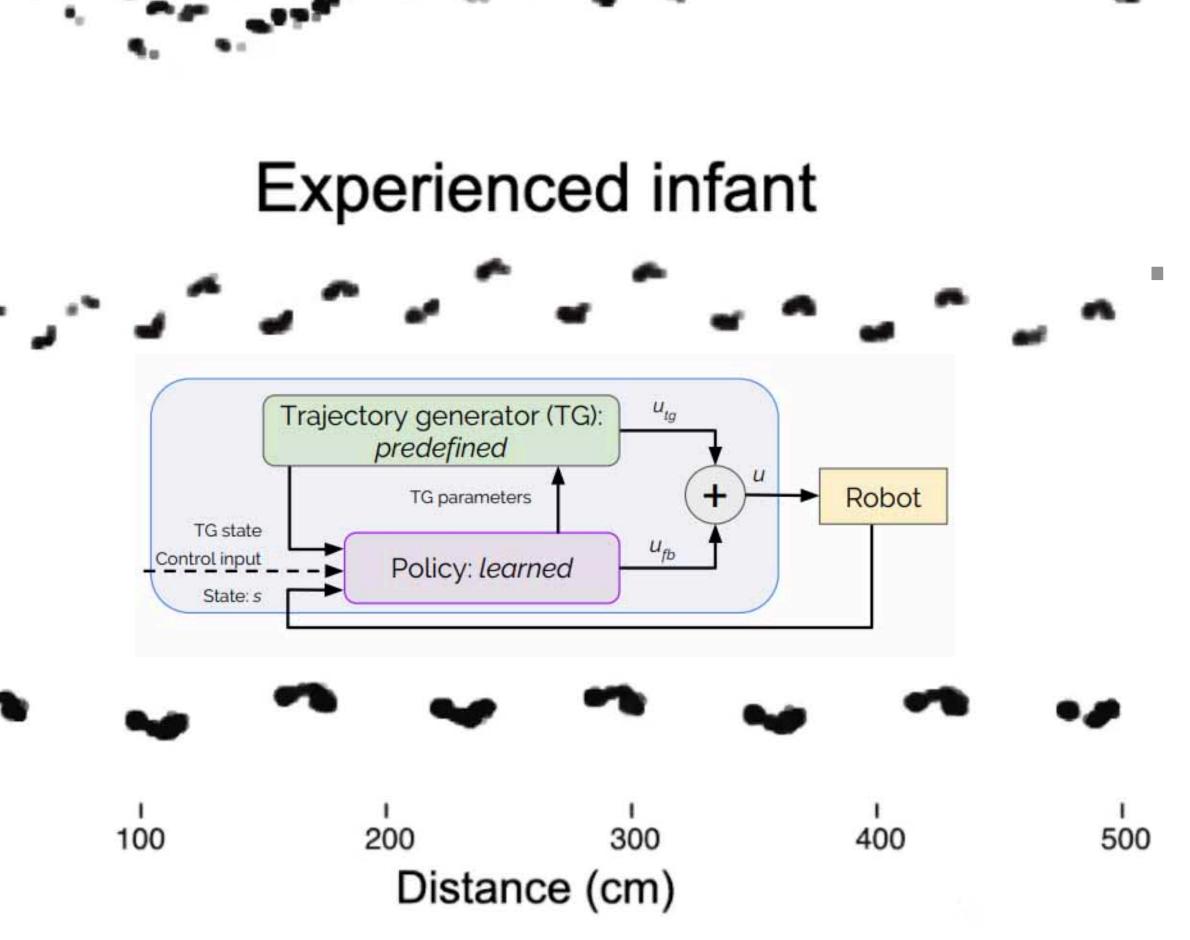
Researcher

## Objective: What Is A Baby Supposed to Learn?

Novice infant

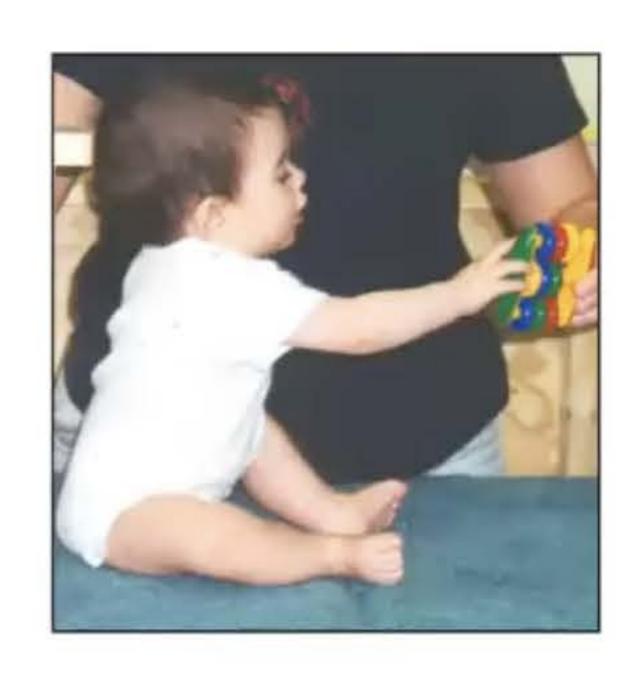


Model

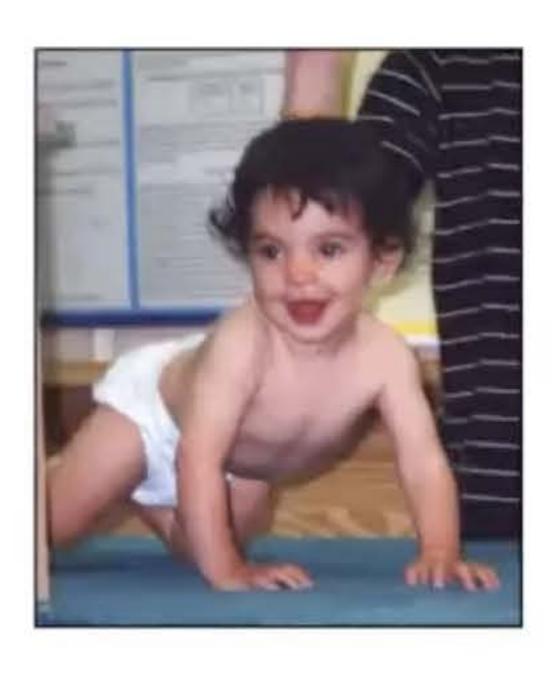


Researcher

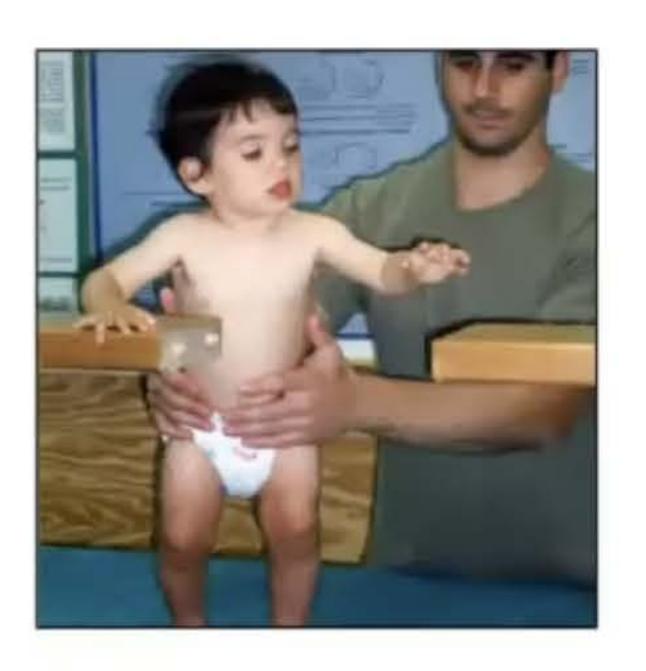
## Different Skills Learned at Different Times



Sit



Crawl



Cruise



Walk

[ Credit: Karen Adolph ]

# Different Skills Learned with Different Bodies

# 12 months

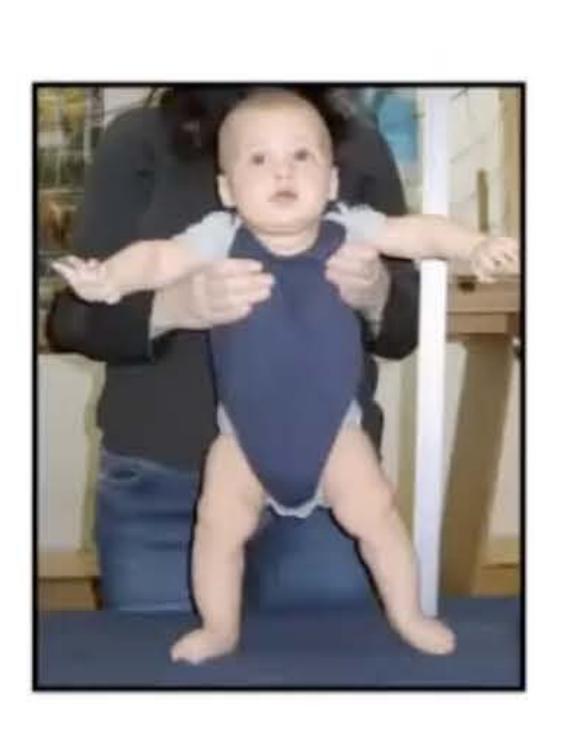




50 cm







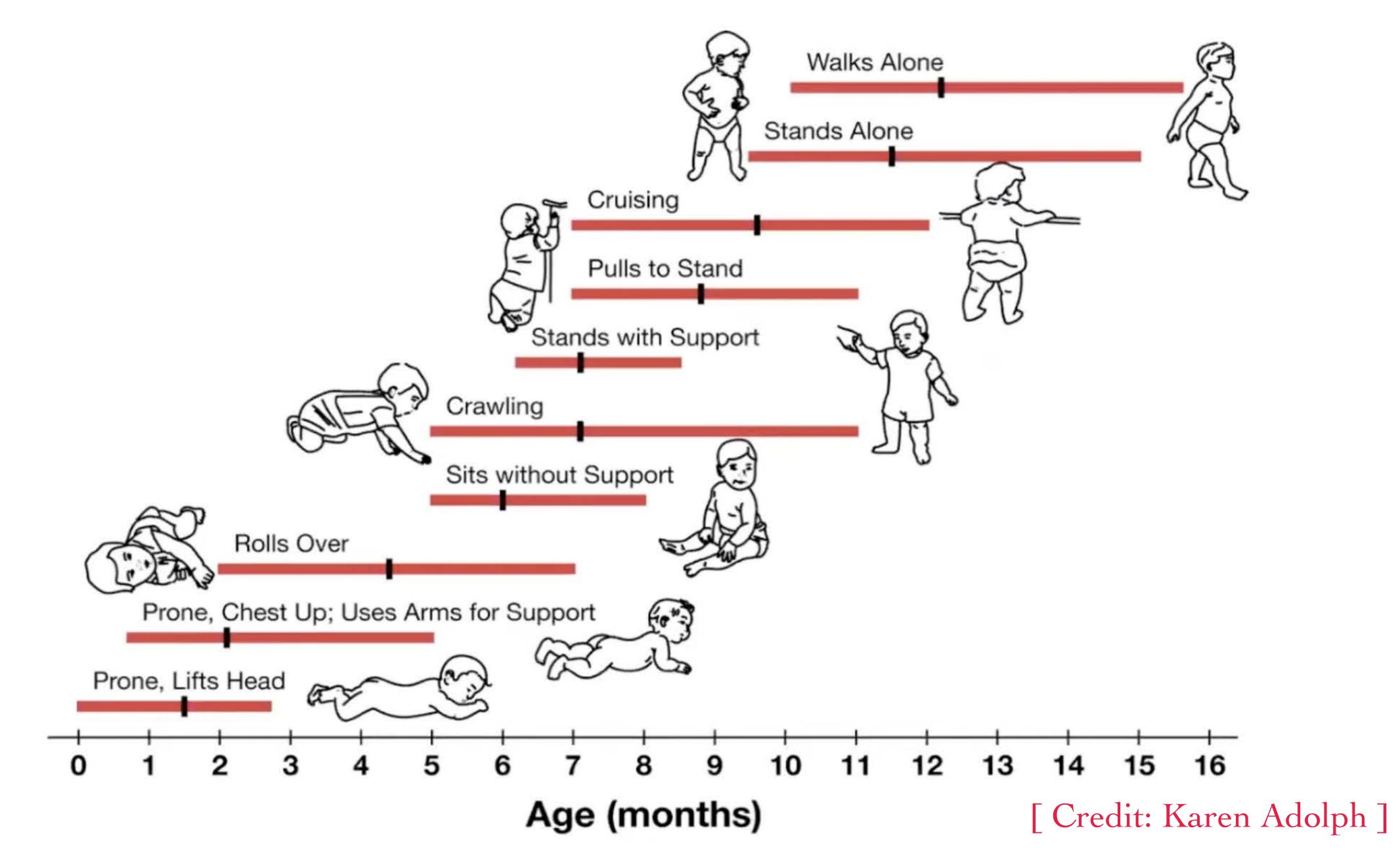




76 cm

[ Credit: Karen Adolph ]

## Different Skills Learned with Different Data



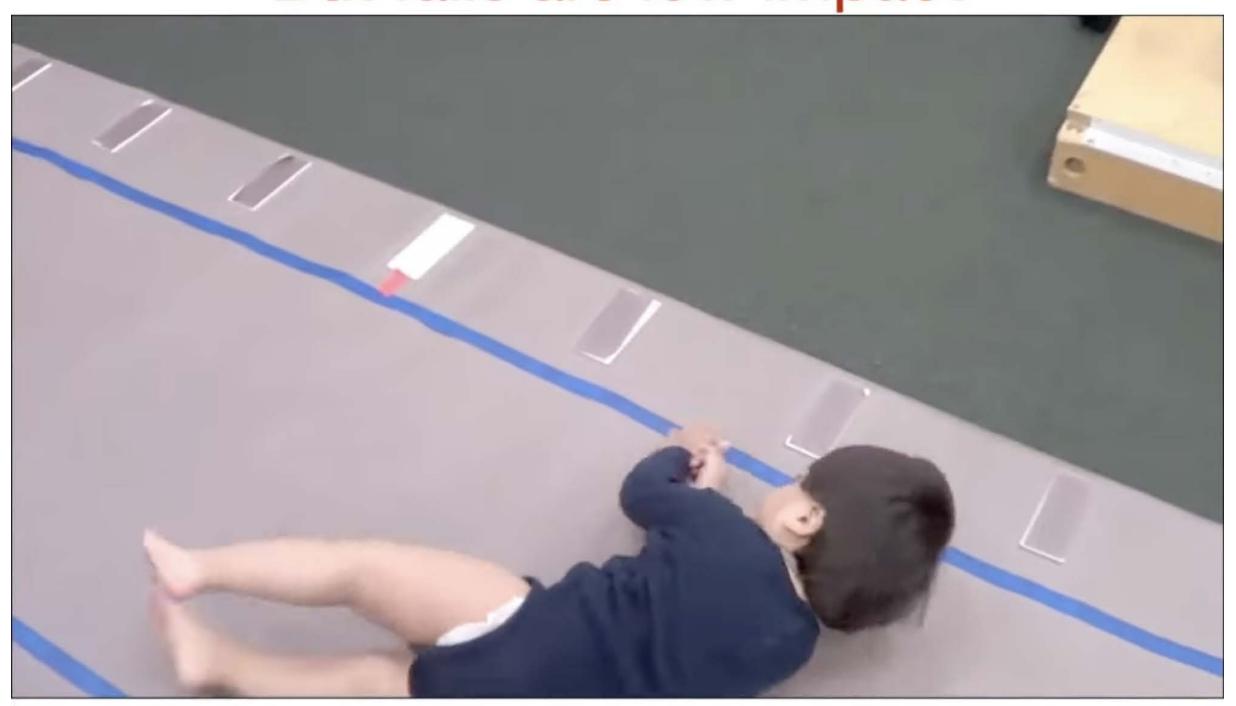
## Skills Learned with Different Sensitivities to Errors

Example #2: High-impact errors (negative reward) shape learning



Robinovitch (2018). Databrary. https://nyu.databrary.org/volume/739

Infant walking is full of errors:
But falls are low impact



Adolph (2012) Psych Sci; Han & Adolph (2021) Dev Sci

## Learn the Simultaneous Development of Action and Perception

We see in order to move

We move in order to see

During the process we learn both how/what to see/move

TABLE 1. Marr's Three Levels of Explanation for Cognitive Capacities (Marr 1982, 24).

Computational Theory	Representation and Algorithm	Hardware Implementation
What is the goal of the computation, why is it appropriate, and what is the logic of the strategy by which it can be carried out?	How can this computational theory be implemented? In particular, what is the representation for input and output, and what is the algorithm for the transformation?	How can the representa- tion and algorithm be realized physically?

## Some Questions to Ponder

- How does our blurry visual system learn to acquire a clear image?
- How do we start to see depth from 2D images?
- How do we start to see colors? And colors of what?
- How do we learn ocular motor control?
- How do we learn reaching and grasping?
- How do we learn locomotion?
- How do we learn manipulation?

•