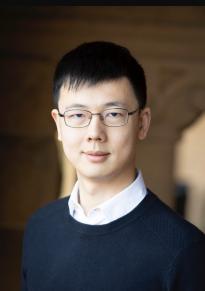
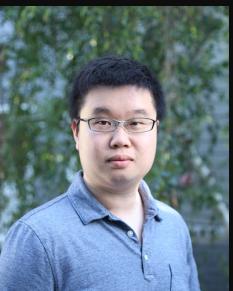
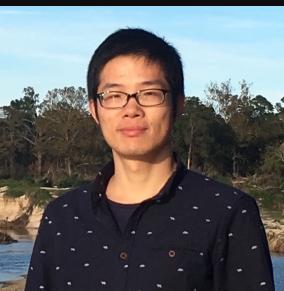


# Bongard-HOI: Benchmarking Few-Shot Visual Reasoning for Human-Object Interactions

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# Visual Concept Learning

Bicycle



...

Ride bicycle



...



Need large amount of training data



Hard to generalize beyond the training concepts

# Bongard-HOI Benchmark

Positive Examples  
**ride bicycle**



Negative Examples  
**not ride bicycle**



Query Images  
**positive**



**negative**



# Hard Negatives in Bongard-HOI

person  
ride  
bicycle



person  
straddle  
bicycle



person  
repair  
bicycle



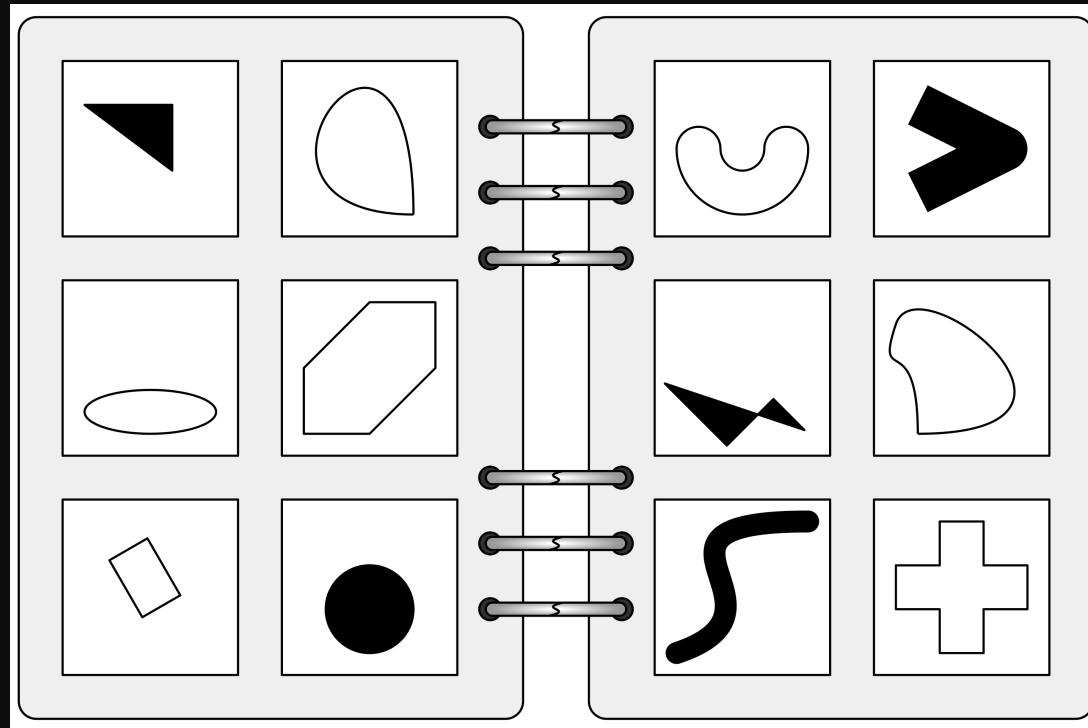
person  
walk  
bicycle



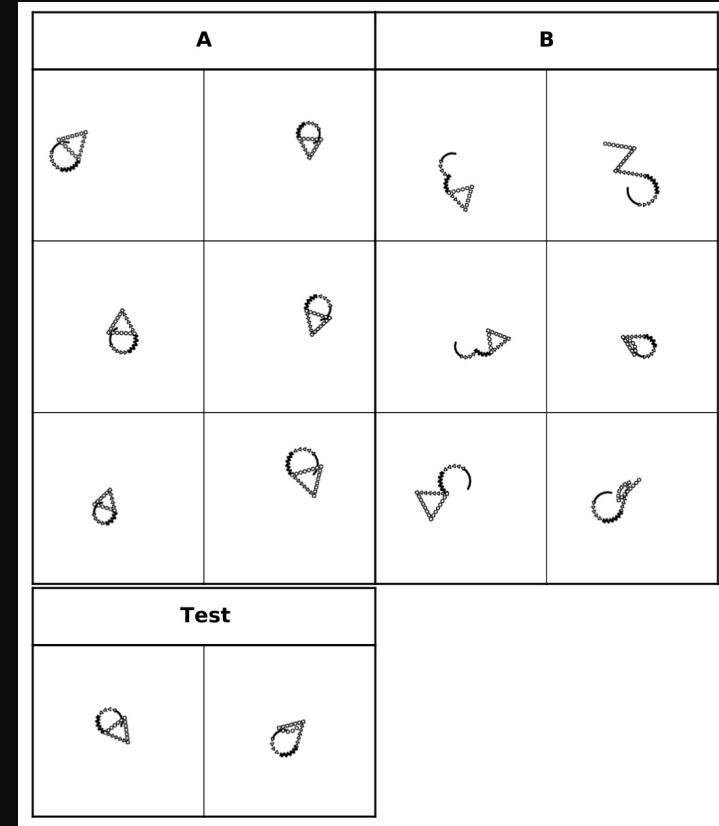
Simple visual recognition is not sufficient.

Visual reasoning (e.g., few-shot learning, context reasoning) of the interactions is needed.

# Inspirations from Cognitive Science



Original Bongard problems  
[Bongard, 1970.]



Bongard-LOGO  
[Nie et al., NeurIPS 2021]

# Different Types of Generalization

Training  
set

sit\_on bed



straddle bicycle



hug person



wash car



wash bicycle



sit\_on bench



greet person



shear sheep



Test  
set

seen action,  
seen object

seen action,  
unseen object

unseen action,  
seen object

unseen action,  
unseen object

Increasing difficulty

# Context-Dependent Reasoning

Positive Examples  
**drink\_with cup**



Hard Negative Examples  
**not drink\_with cup**



Query Images  
**negative**



# Context-Dependent Reasoning

Positive Examples  
**hold cup**



Hard Negative Examples  
**not hold cup**



Query Images  
**positive**

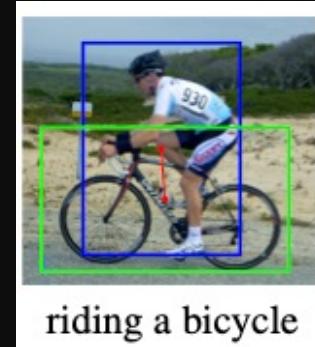


# Comparisons with Other Benchmarks

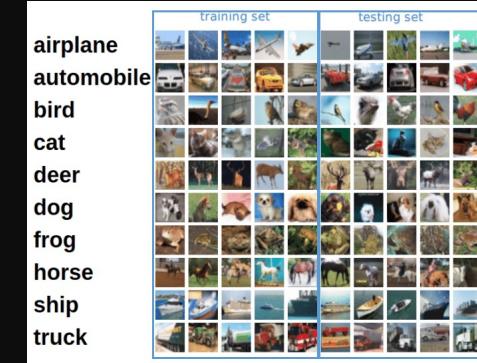
Bongard-  
HOI



HOI  
detection



minilmageNet



Bongard-  
LOGO

A	B
∅	∅
∅	∅
∅	∅
Test	
∅	∅

Natural images



Hard negatives



Compositional concept



Few-shot learning



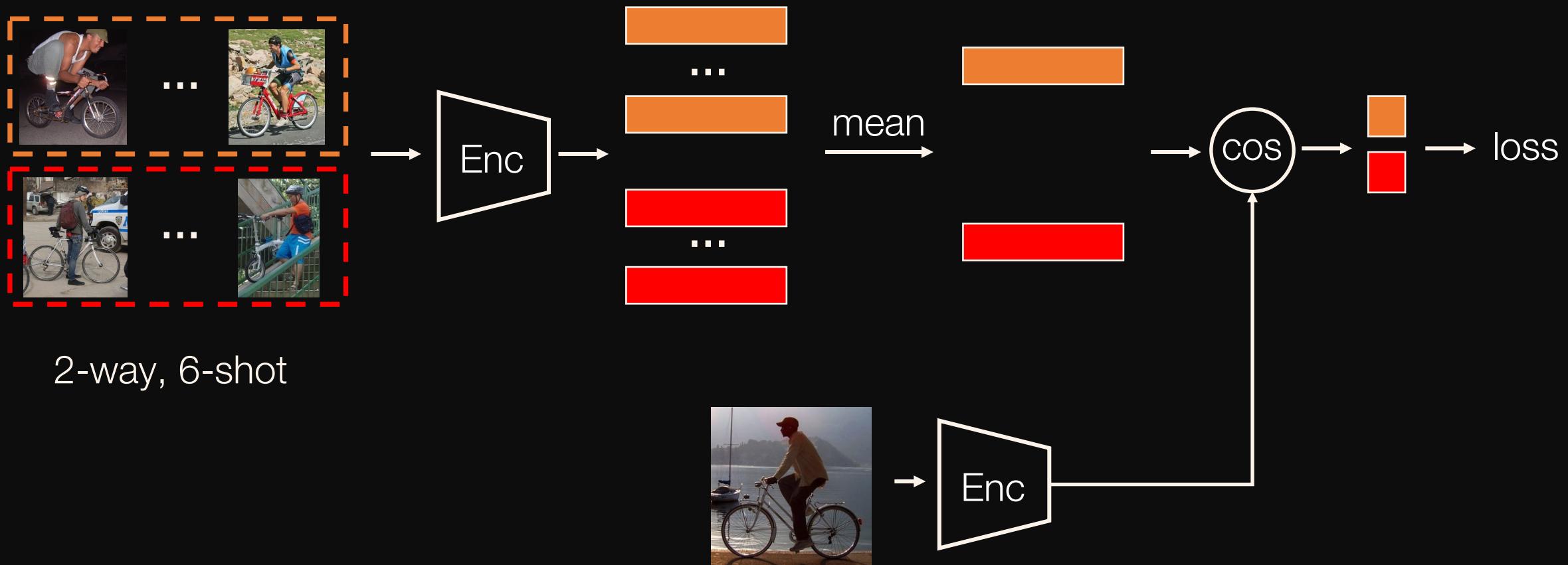
Ctx.-dependent reasoning



Generalization types



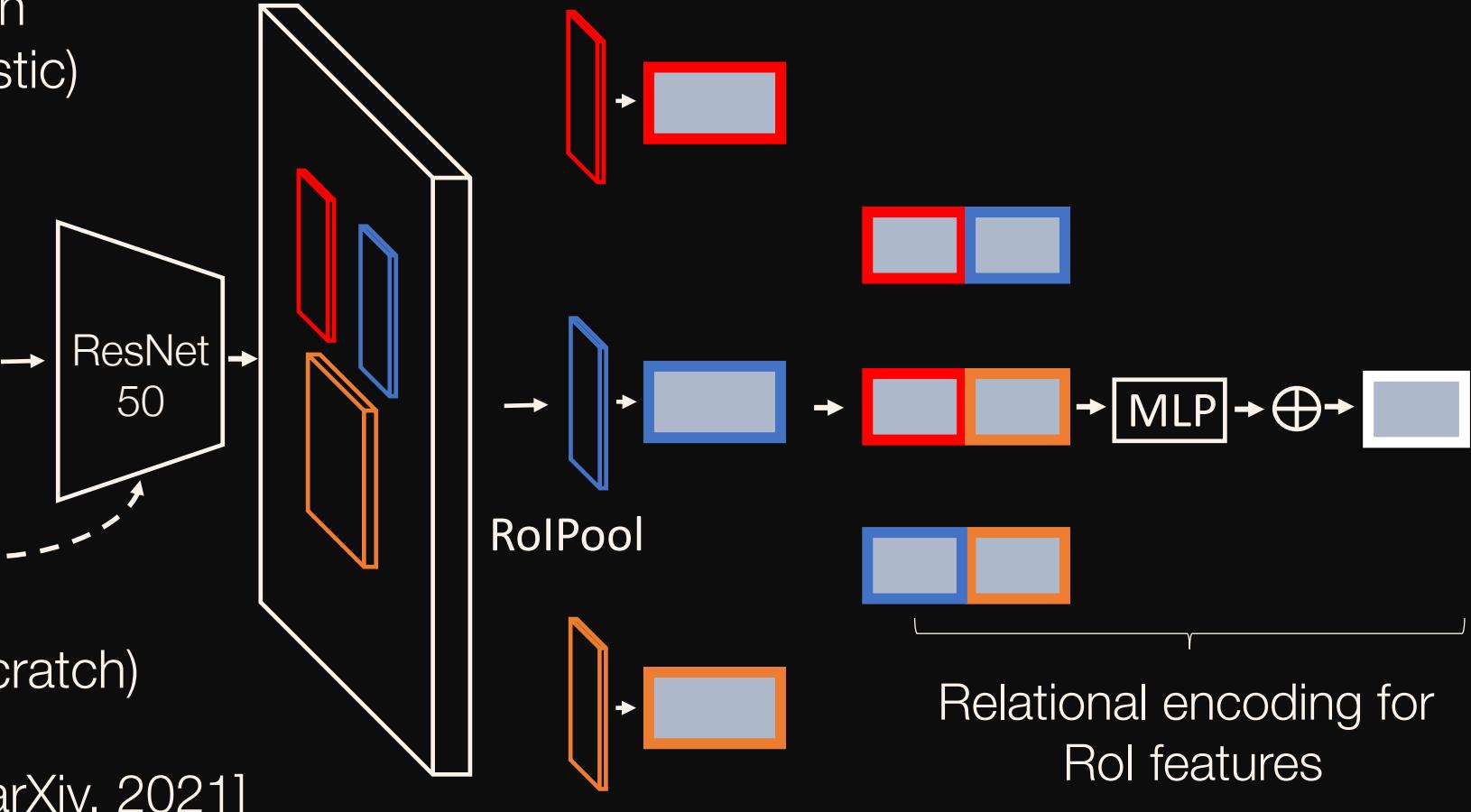
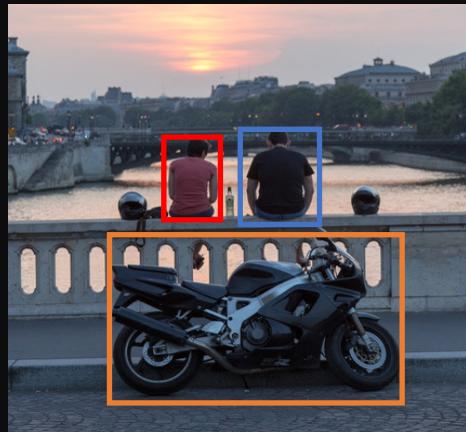
# Meta-Learning for Bongard-HOI



[Chen et al., Meta-Baseline. ICCV 2021]

# Image Encoding with Relational Network

Objectness detection  
(binary category-agnostic)

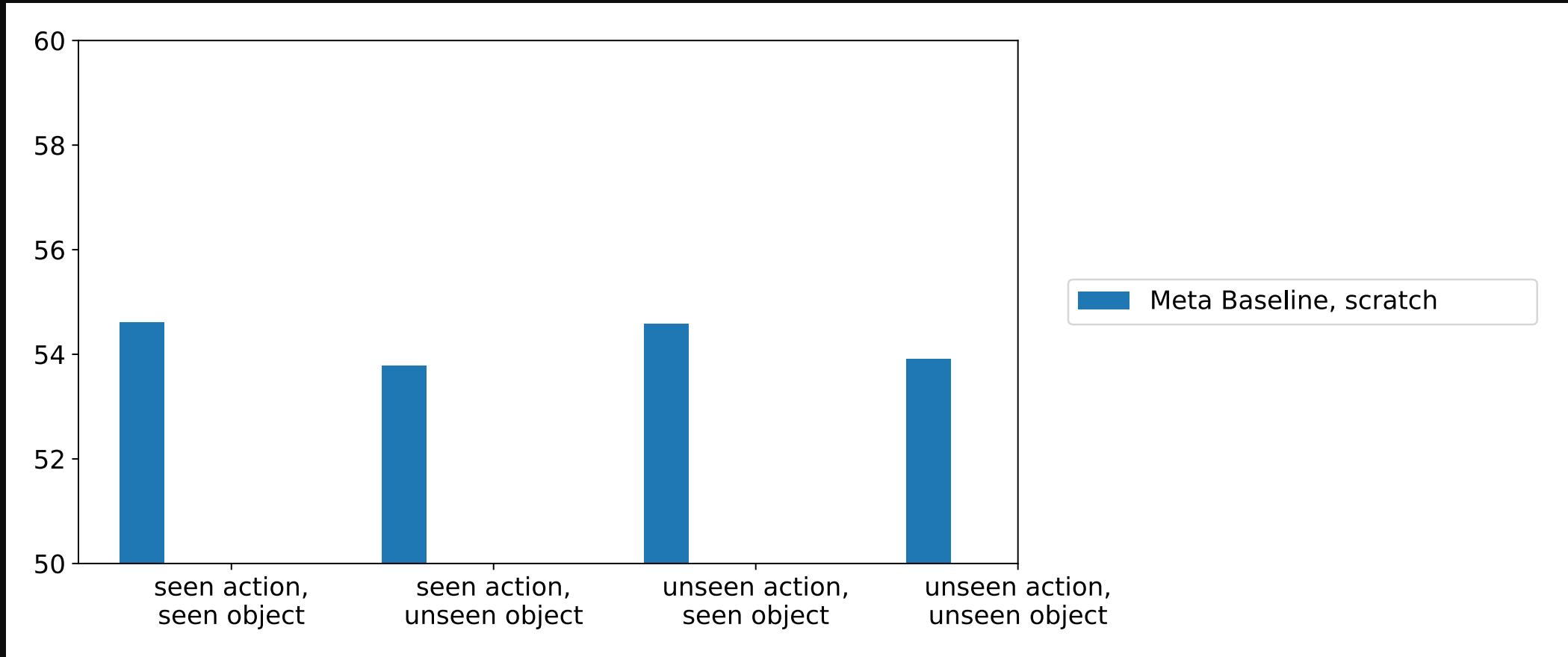


- Random initialization (scratch)
- ImageNet pre-training
- MoCo V2 [Chen et al., arXiv, 2021]

Relational encoding for  
ROI features

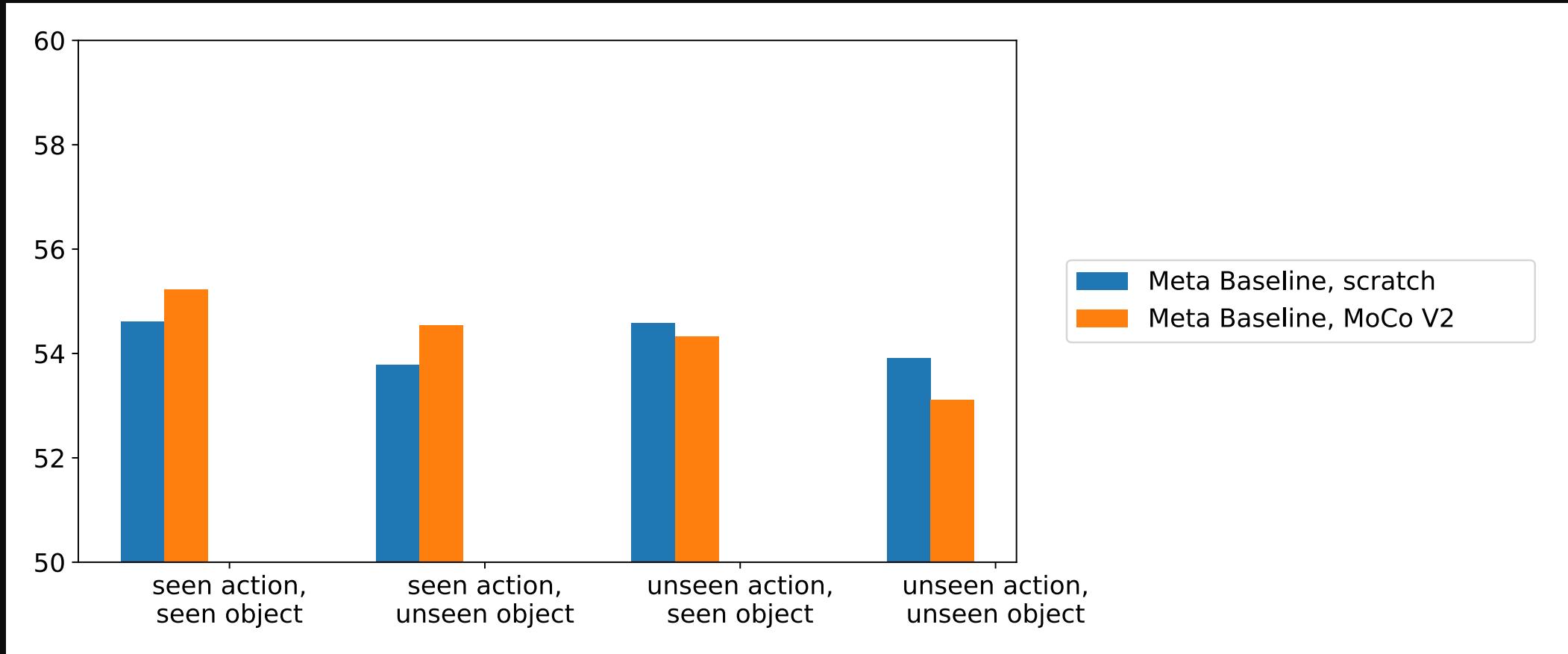
[Santoro et al., NeurIPS 2017]

# Comparisons of Meta Learning Models



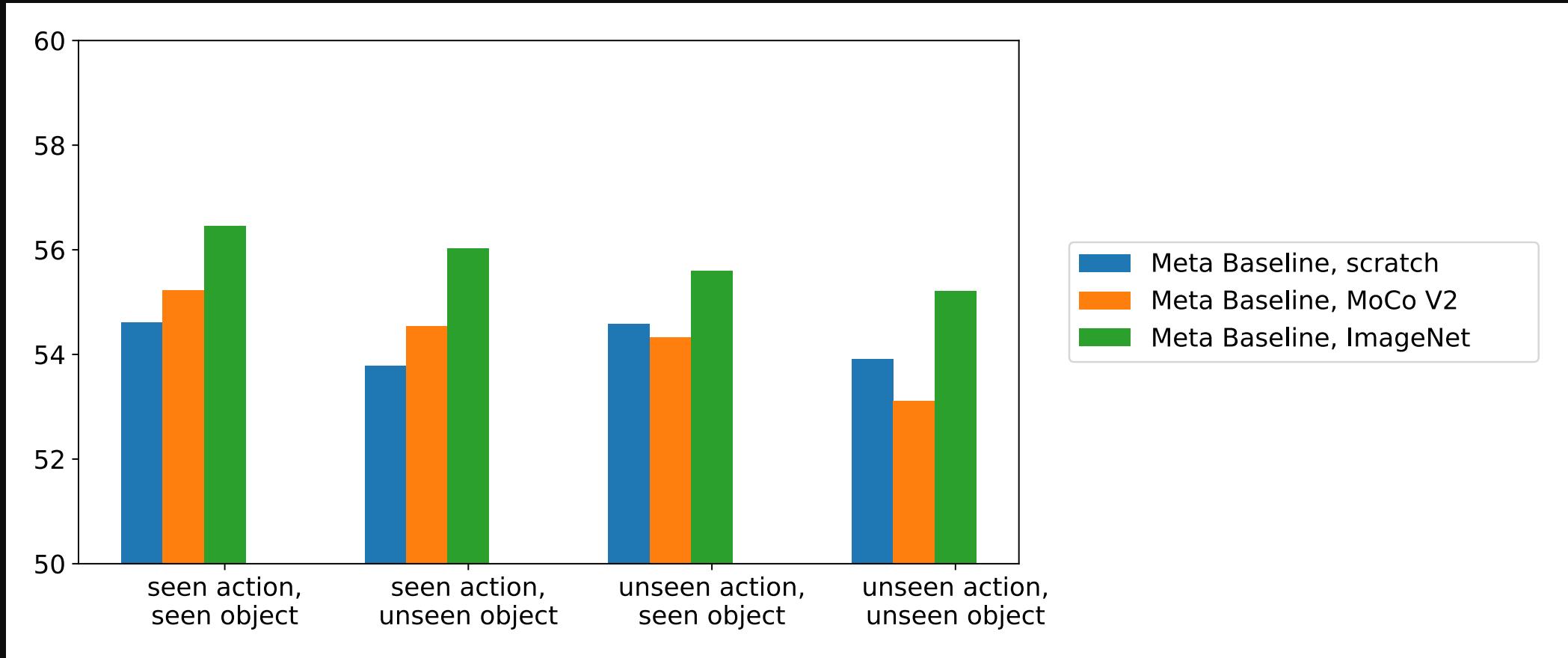
Increasing difficulty

# Comparisons of Meta Learning Models



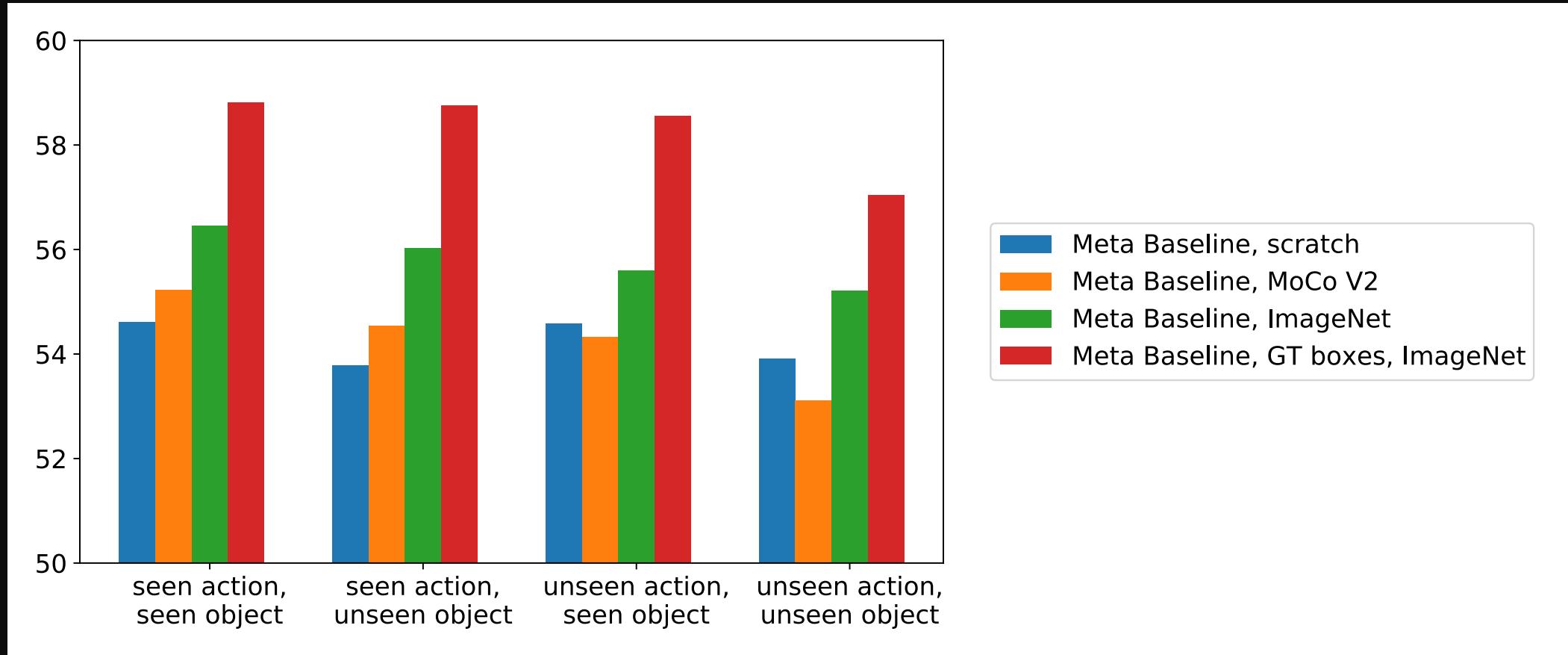
Increasing difficulty →

# Comparisons of Meta Learning Models



Increasing difficulty

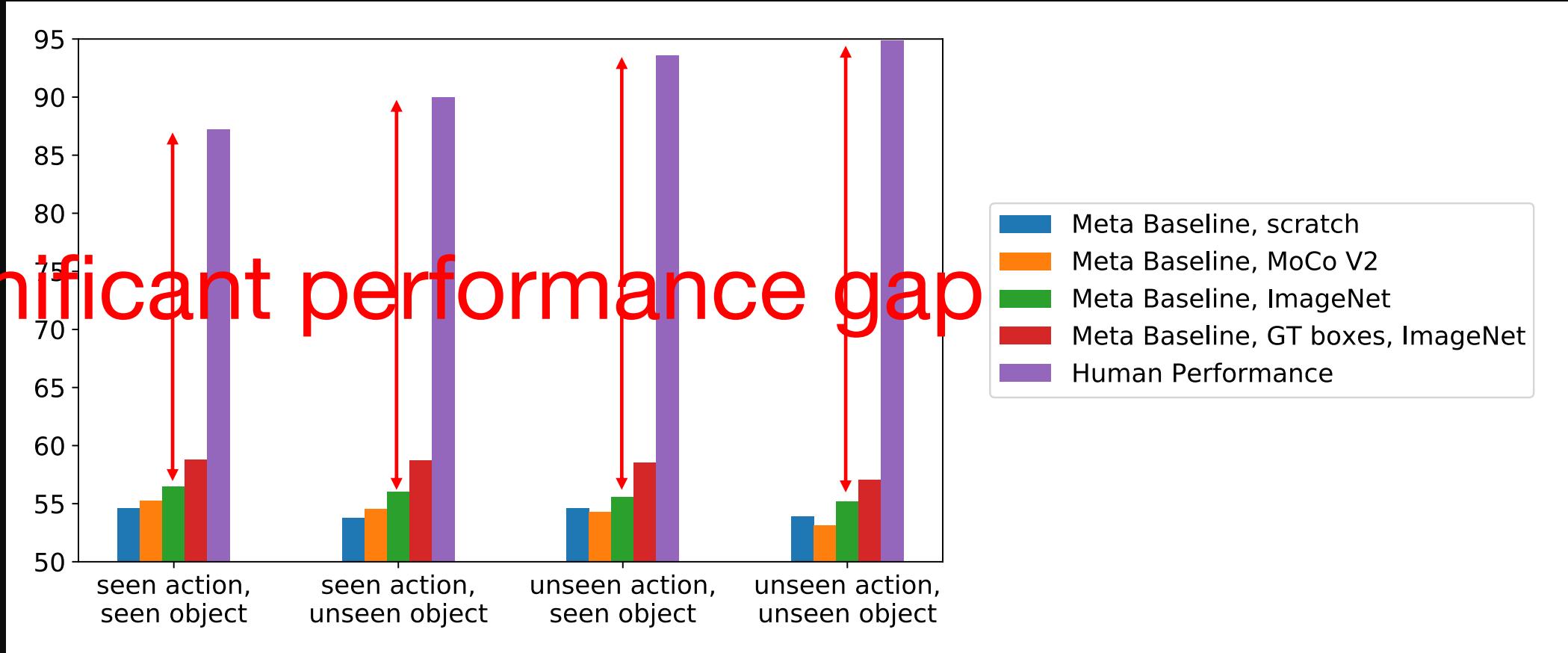
# Importance of Holistic Visual Perception and Reasoning



Increasing difficulty

# Comparisons with Human Accuracy

significant performance gap



# Take-home Messages of Bongard-HOI

- A new benchmark about HOI, highlighting visual reasoning
  - Few-shot learning
  - Context reasoning
  - Generalization beyond training concepts
  - ...
- Meta-learning models do not work well enough
  - Pre-training is helpful
  - Holistic visual perception and reasoning is essential
- There exit huge gap w.r.t human performance

# Data and Code



Poster: 36b