# UOUO: Uncontextualized Uncommon Objects for Measuring Knowledge Horizons of Vision Language Models

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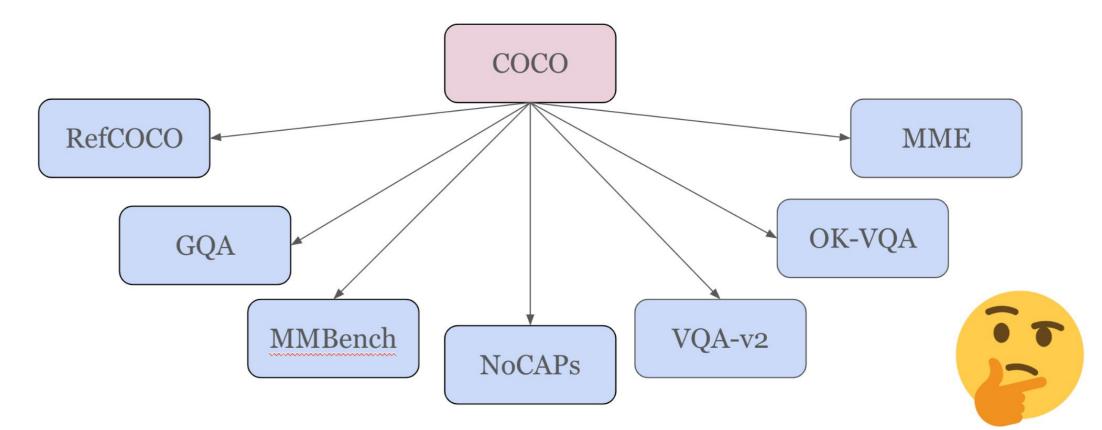
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## COCO vs. UOUO (Long tail, Uncommon)

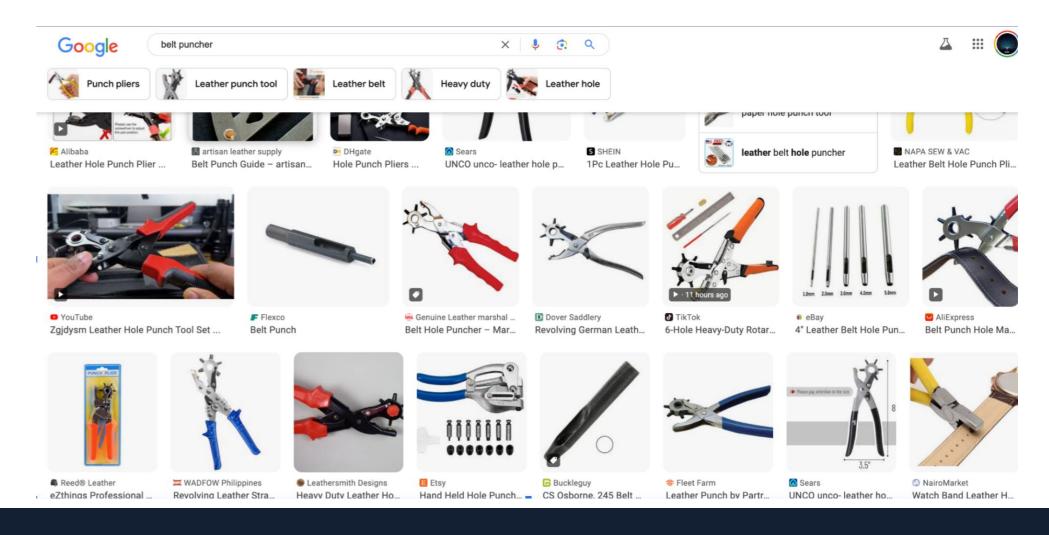




#### No VLM evaluation in uncommon



### Web scraping -> uncommon object



Score

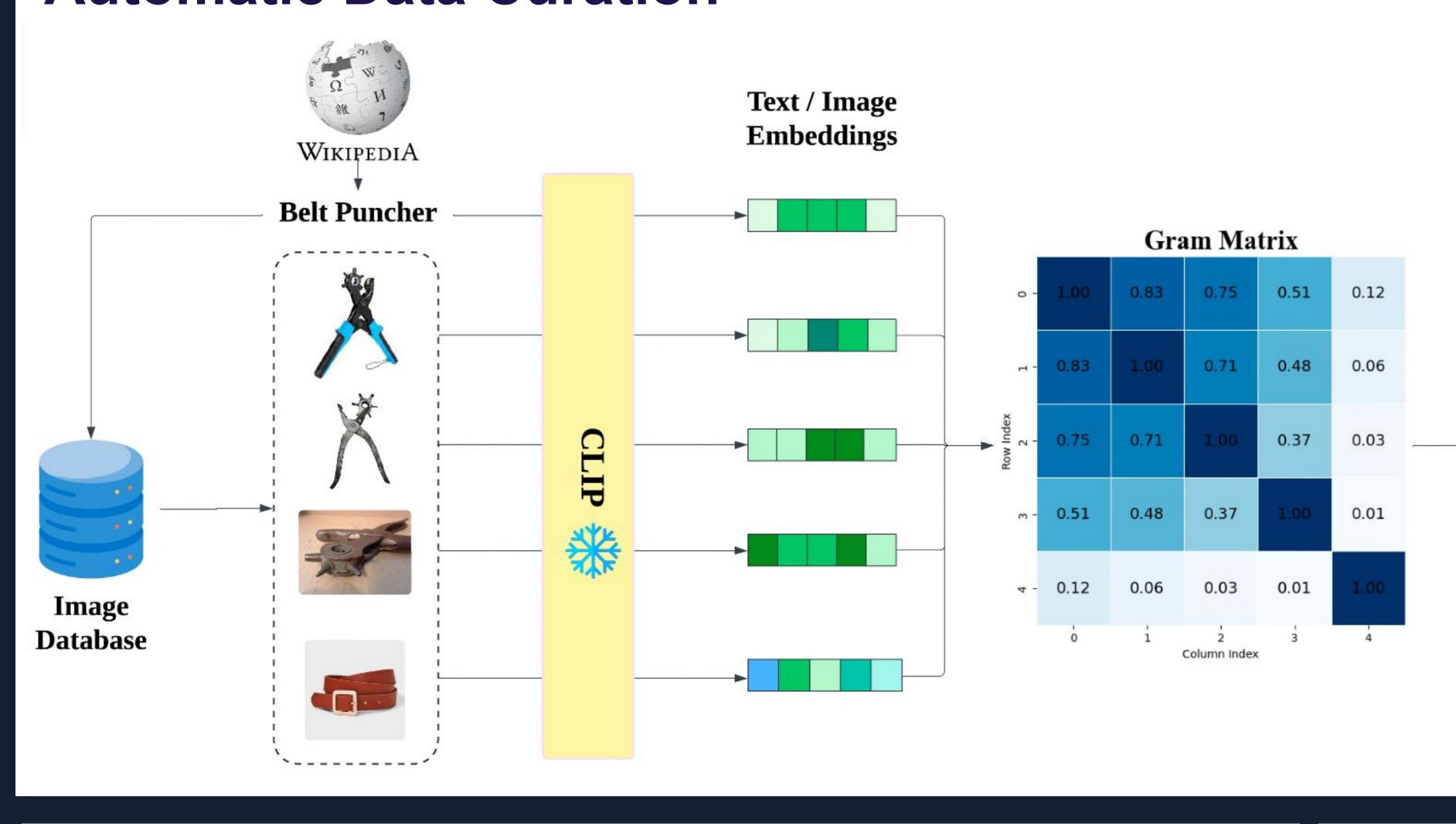
0.98

0.87

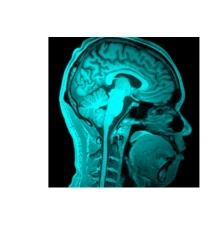
0.59

0.06

#### **Automatic Data Curation**



## **Profile Photo**





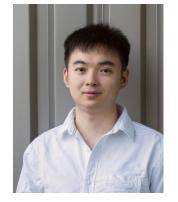




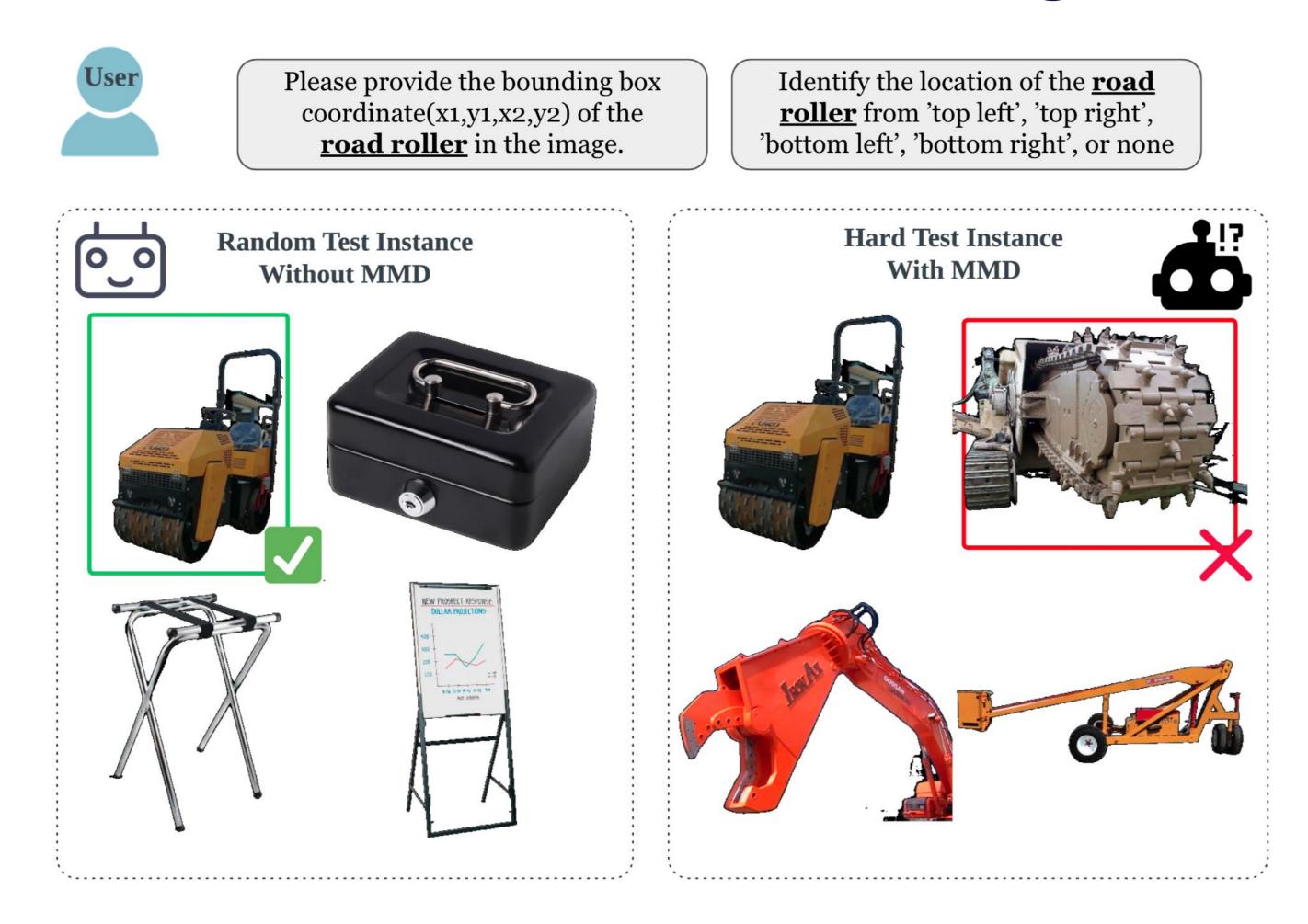








## **Question Generation -> Grounding**



 $MMD(\mathbf{x}, \mathbf{y}) = k(\mathbf{x}, \mathbf{x}) + k(\mathbf{y}, \mathbf{y}) - 2 \cdot k(\mathbf{x}, \mathbf{y})$ 

Where x and y be the sets of CLIP embeddings for two different object categories, where k is a kernel function (We adopt Gaussian RBF)

## **Experiment Results**

Model Name			
llava-v1.5-7b			
llava-v1.5-13b			
llava-v1.6-vicuna-7b			
llava-v1.6-vicuna-13b			
llava-v1.6-34b			
cogvlm-llama3-chat-19b			
gemini-1.5-pro			
gpt-4-turbo			
gpt-4o			

mloU-mmd	mloU-rand	acc-mmd	acc-rand
0.18	0.41	0.42	0.70
0.23	0.47	0.44	0.73
0.28	0.48	0.49	0.75
0.28	0.49	0.52	0.78
0.38	0.55	0.57	0.83
0.49	0.69	0.43	0.60
0.27	0.27	0.63	0.80
0.34	0.38	0.67	0.90
0.33	0.35	0.68	0.88

#### Take away

- Smaller VLMs struggle with uncommon objects, especially in low MMD mosaics.
- A specialized dataset for benchmarking VLMs on uncommon objects.
- An automatic pipeline for web data scraping, curation, filtering, and generating challenging test instances for domain-specific objects