

# Product recognition AI

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#### **Problem statement**

#### **Manual Tagging**

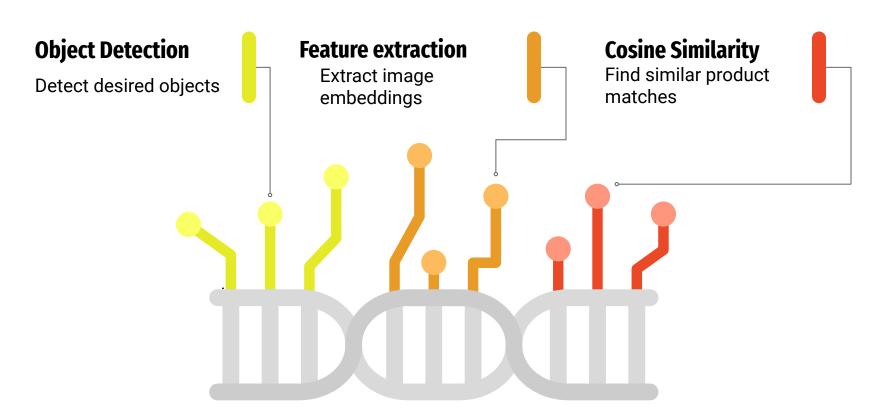
- Manual Product tagging is a time consuming task, it involves:
  - Searching for each product label
  - Searching for similar products

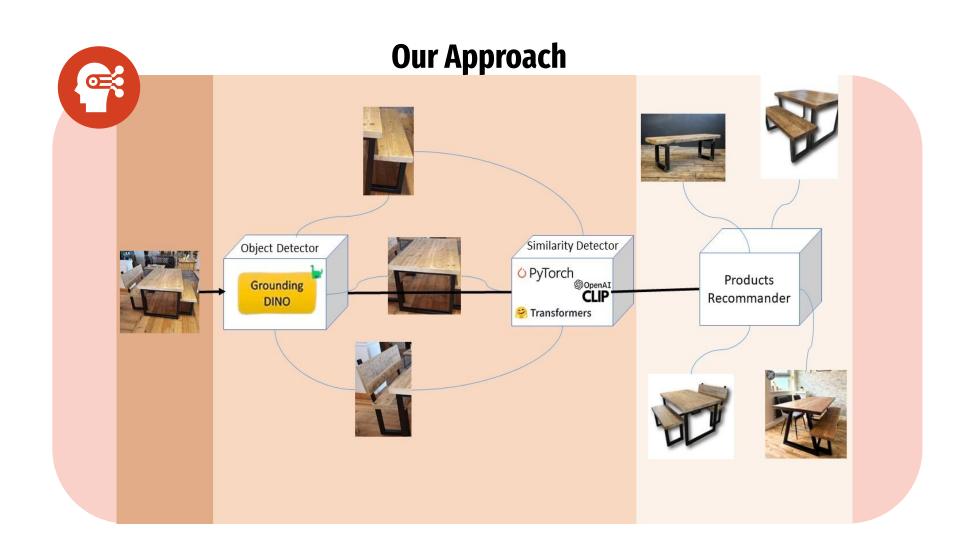
#### **Automatic Tagging**

- Streamline business needs
- More efficient

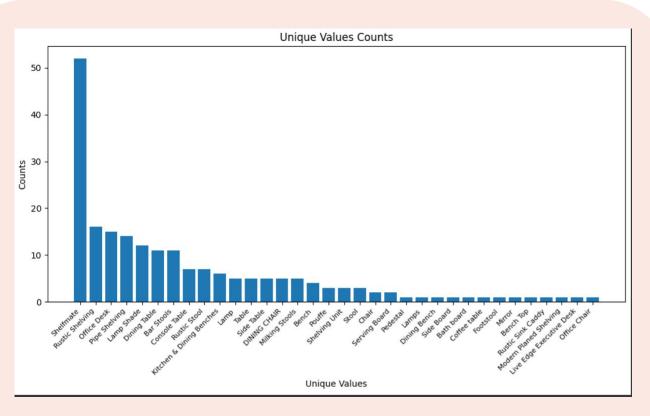
#### **Oculizm** Acumen Collection (83953) × $\equiv$ $\otimes$ ✓ Home Q Search by name or Product ID... ALL PRODUCTS Summary Sales 0 -Orders Content Posts ^ All posts Galleries ① Curate ^ Facebook Instagram acumencollection acumencollection acumencollection acumencollection Upload Products All Products **Product Feeds** Reviews Waiting for app.oculizm.com...

#### **Proposed Solution**

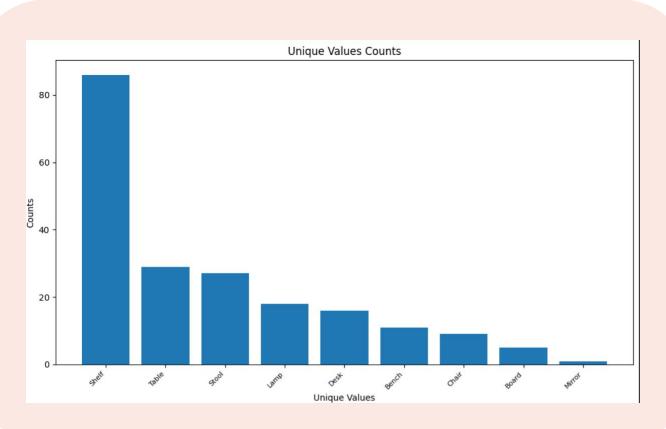


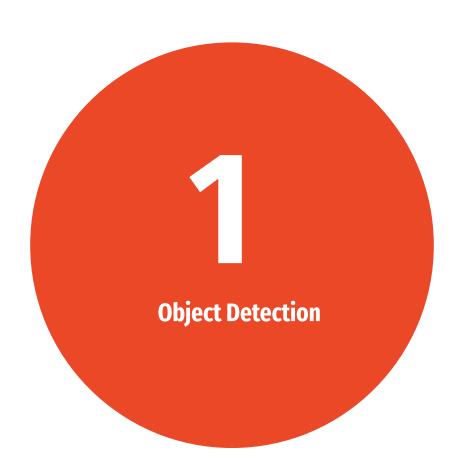


#### **Data Before Cleaning**



#### **Data After Cleaning**





## **What's grounding DINO**

#### **Grounding Dino**



**Zero-Shot Object Detection** 

detecting objects even when they are not part of the predefined set of classes in the training data

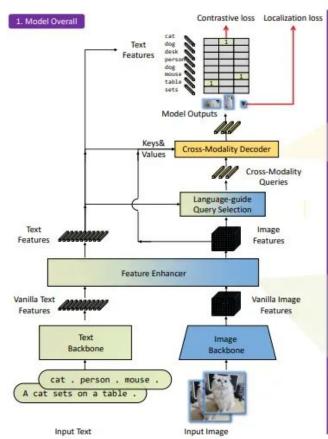
Referring Expression Comprehension (REC)

Identifying and localizing a specific object or region within an image is based on a given textual description

Elimination of Hand-Designed Components

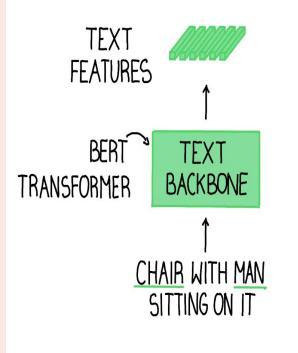
simplifies the object detection pipeline by removing the need for hand-designed components **Grounding dino architecture** 

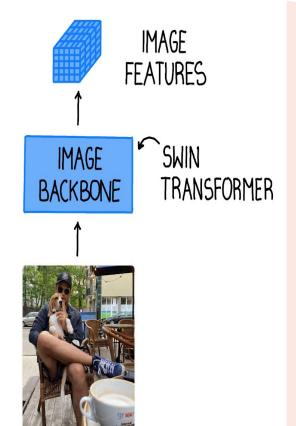






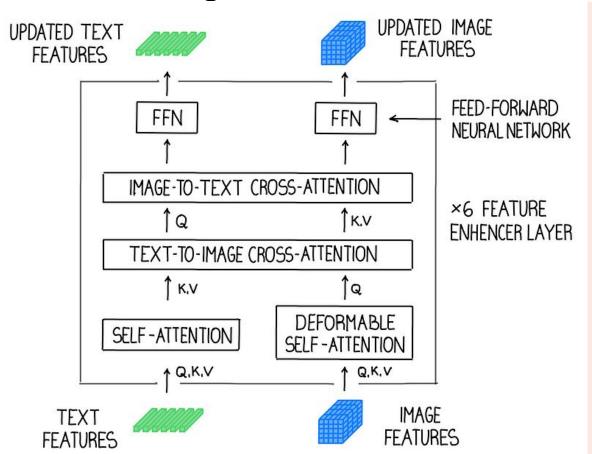
#### **Grounding dino Text and Image Backbone**



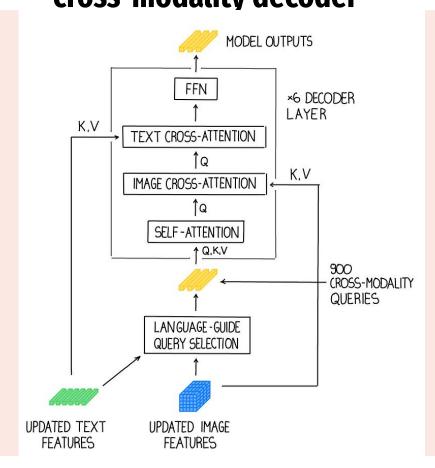




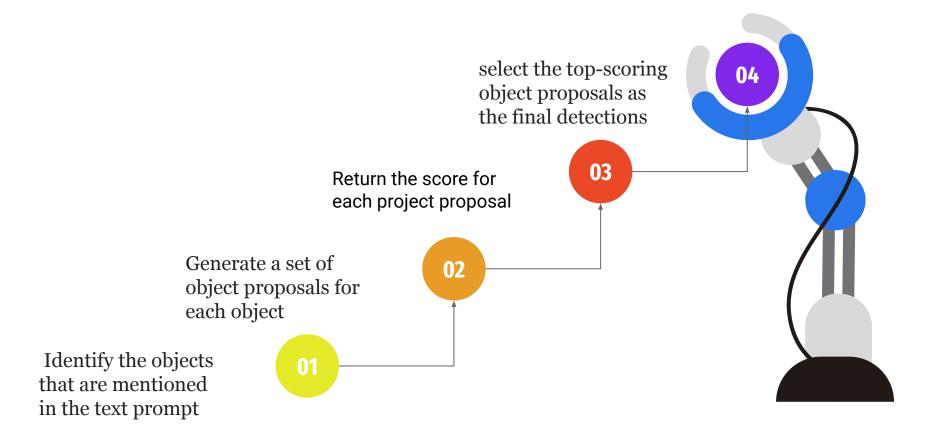
#### **Grounding dino Feature enhancer**



Grounding dino language-guided query selection and cross-modality decoder



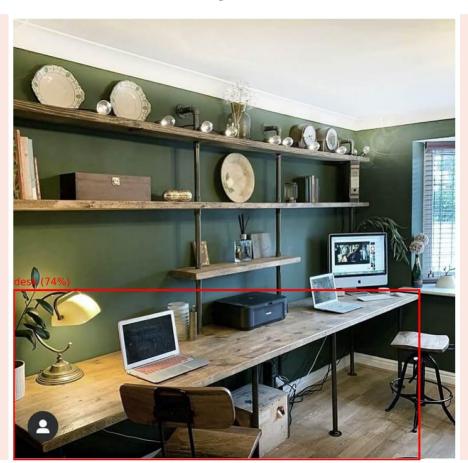
#### **How it works**



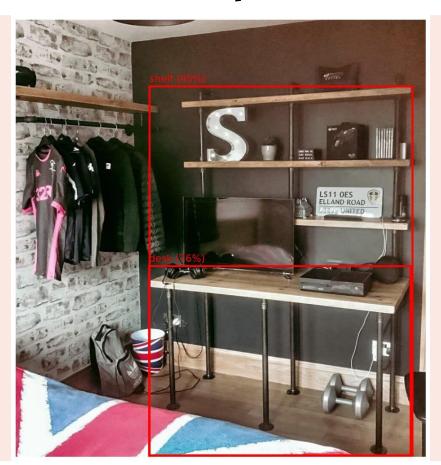












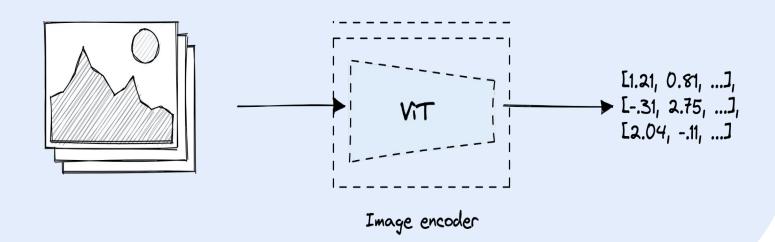


## What's Clip(Contrastive Language-Image Pre-training)

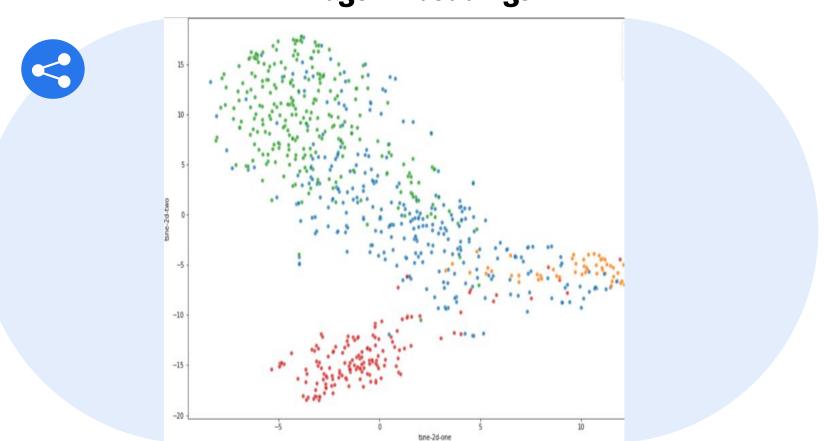


#### **Clip image embeddings**





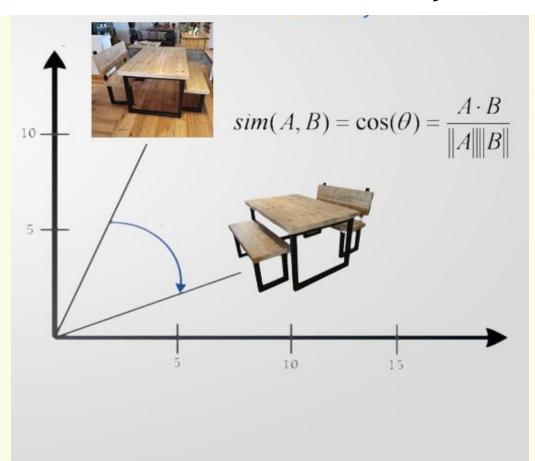
## **Image Embeddings**



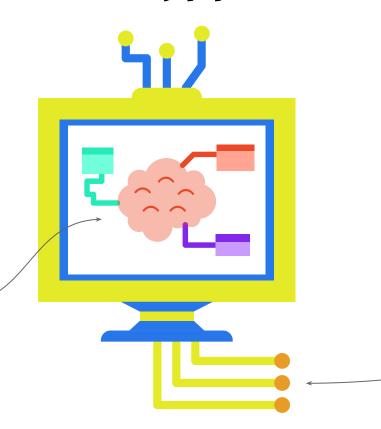


## What's cosine similarity





## **Similarity pipeline**



- Recom 1
- Recom 2
- Recom 3

- transform
- Embedding
- Label



Similarity: 79% ID: 4632199233591



Similarity: 77% ID: 6622128472119



Similarity: 75% ID: 4290573697079



Similarity: 75% ID: 4465363648567





Similarity: 79% ID: 6624599736375



Similarity: 79% ID: 6871472537655



Similarity: 77% ID: 6820334108727



Similarity: 77% ID: 3528439038016





Similarity: 87% ID: 6624603111479



Similarity: 84% ID: 4632992907319



Input Image table (table)

Similarity: 84% ID: 4453882724407



Similarity: 83% ID: 4467025641527





Similarity: 87% ID: 6624603111479



Similarity: 84% ID: 4632992907319



Input Image table (table)

Similarity: 84% ID: 4453882724407



Similarity: 83% ID: 4467025641527



#### **Deployment**





#### **Future Work**



#### Our roadmap includes:

- Expanding the model to diverse product categories like clothing.
- fine-tuning the grounding DINO approach through semi-supervised learning, effectively bridging labeled and unlabeled data.
- Enhance evaluation metrics by introducing a new dimension for measuring similarity, ensuring a more comprehensive assessment.

