ManipVQA: Injecting Robotic Affordance and Physically Grounded Information into Multi-Modal Large Language Models

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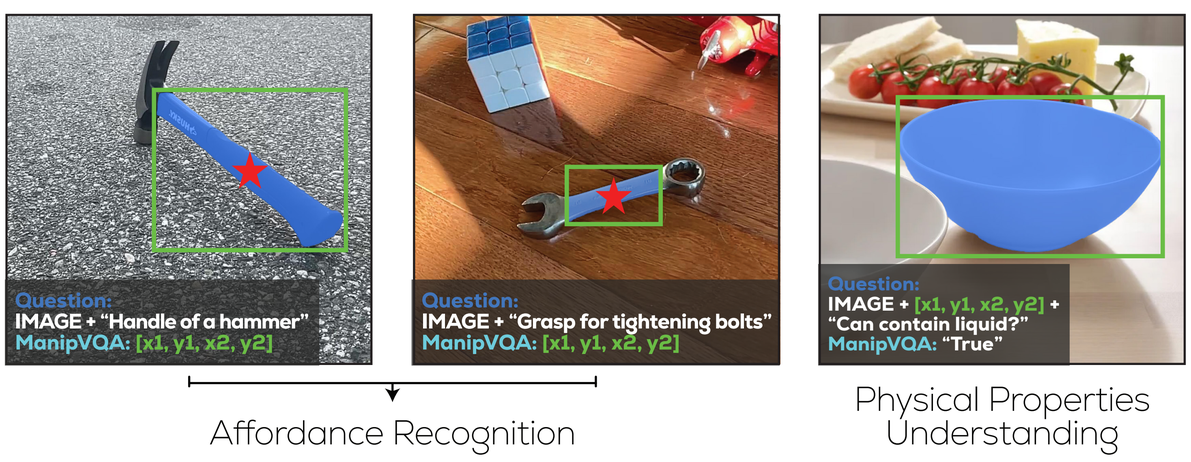
Introduction

Current MLLMs, while proficient in general vision tasks, encounter significant challenges in robotic manipulation. These limitations arise from their struggle to recognize affordances and physical properties of objects, which are essential for robotic manipulation.

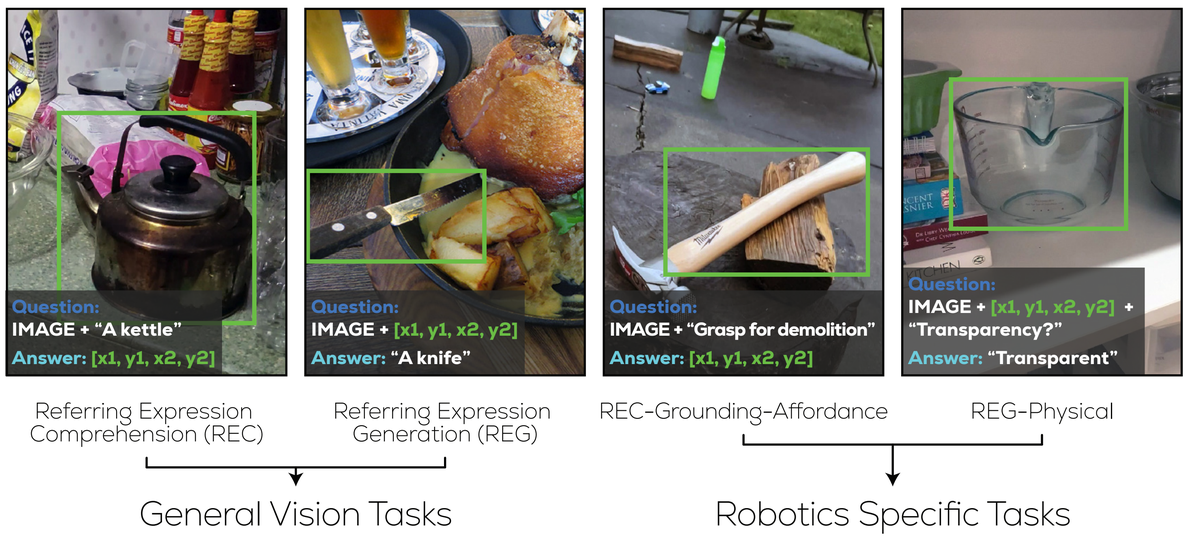
ManipVQA overcomes these limitations by infusing MLLMs with robotics-specific knowledge.

Demonstration



Method & Key Contributions

1 Unified VQA Format



2 Training on Specialized Datasets

General Visual Reasoning Datasets

PACO, RefCOCO, and Visual Genome
Rich sources of information on parts and attributes of common objects

Physically Grounded Dataset

PhysObjects Dataset
We use annotations for liquid storage suitability, seal-ability, and transparency

Robotic Affordance Datasets

HANDAL Dataset
212 hardware and kitchen tools with annotated handle locations

RGB-D Part Affordance Dataset
105 kitchen, workshop, and garden tools with 7 pre-defined affordances

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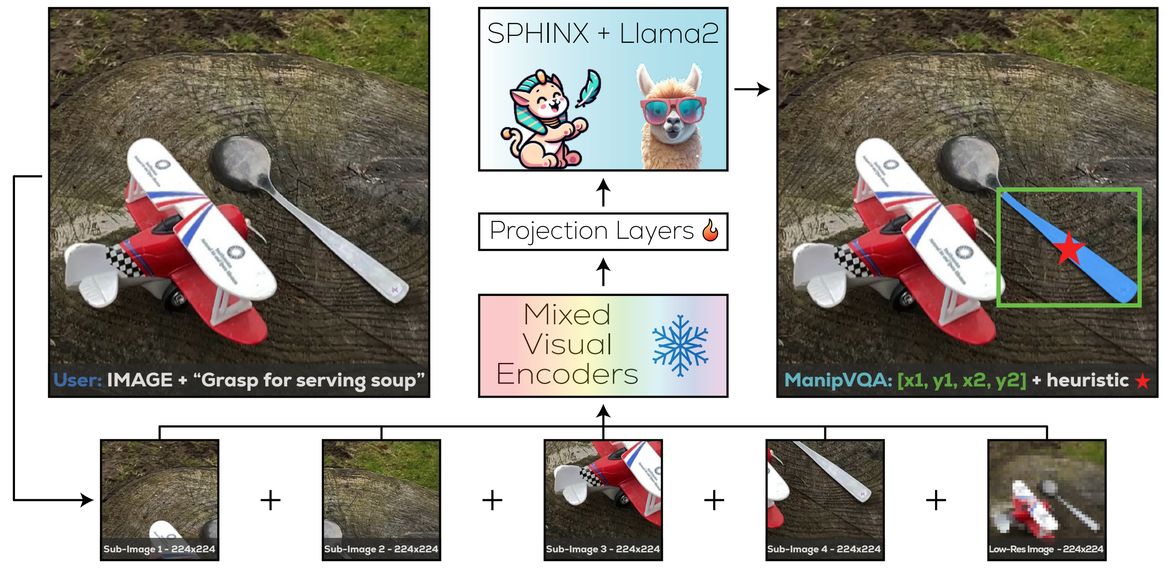
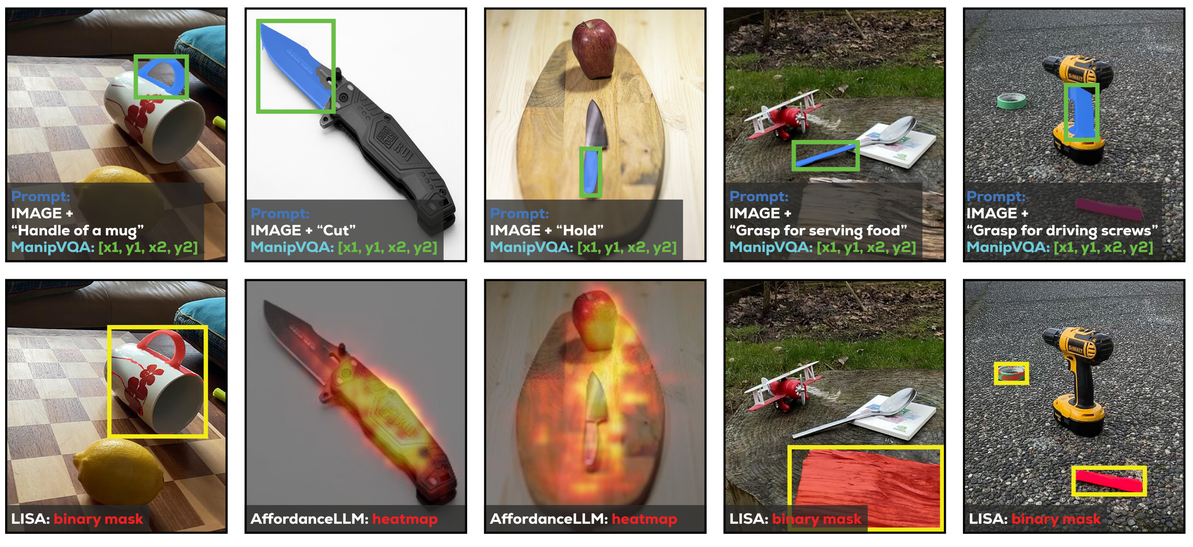
Augmented Instructions with GPT-4
Contextually rich affordance-based tasks

3 Built on SPHINX and Llama2



Experiments

ManipVQA Outperforms Previous Models in Robotic Specific Vision Tasks



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