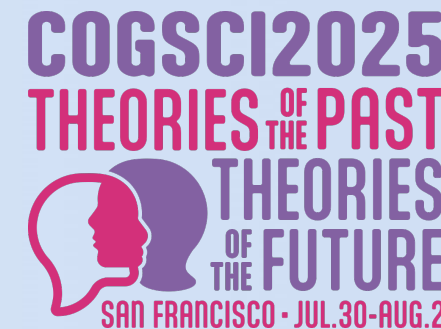




¹PEKING
UNIVERSITY



²BIGAI



Probing and Inducing Combinational Creativity in Vision-Language Models

Yongqian Peng^{*,1}, Yuxi Ma^{*,1}, Mengmeng Wang², Yuxuan Wang²,
Yizhou Wang¹, Chi Zhang², Yixin Zhu^{1,✉}, Zilong Zheng^{2,✉}

*equal contribution ✉corresponding authors

Background



<https://www.bbc.com/news/videos/cn4vwq0v9v5o>

Do these creative products generated by AI emerge from
genuine creative processes or **sophisticated pattern matching**?

Meet Ai-Da: the humanoid robot artist whose painting sold for \$1,1 million

Janice Beckett-Msiza

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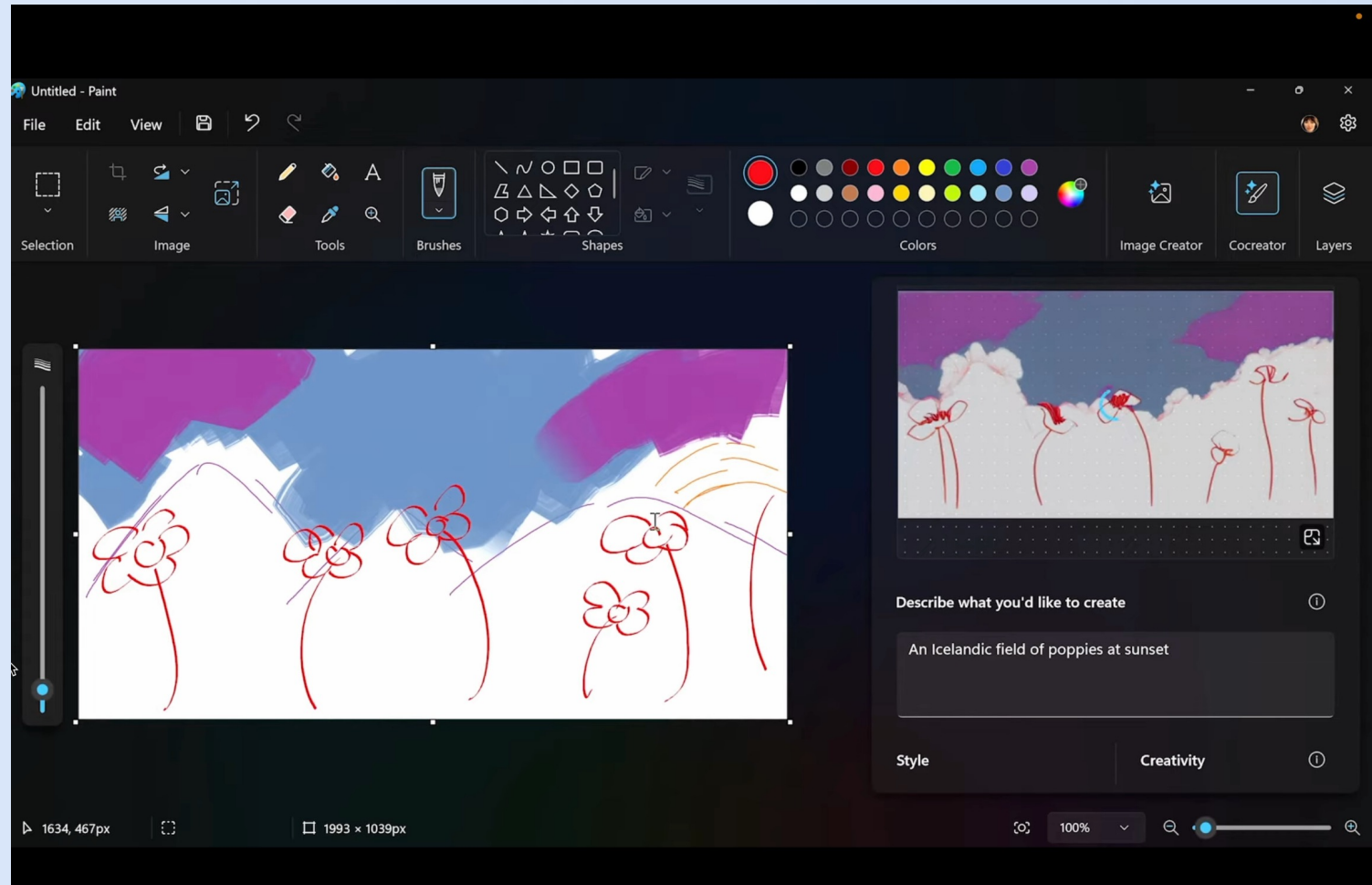
Ai-Da is the first humanoid robot artist to sell artwork at auction. (PHOTO: Gallo Images/Getty Images)

<https://www.news24.com/you/news/international/meet-ai-da-the-humanoid-robot-artist-whose-painting-sold-for-11-million-20241112>

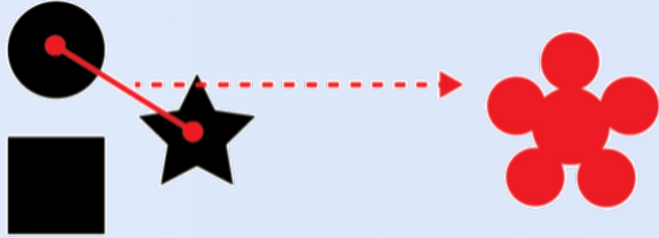
Why Study Machine Creativity?

“Creativity is the ability to produce work that is both novel and useful.”

- **Understanding Human Creativity Better**
- **Enhancing Human Creativity**

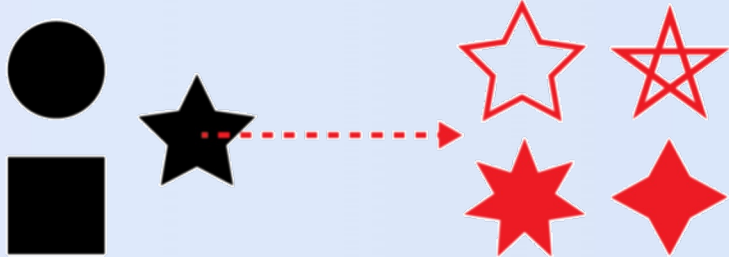


Combinational Creativity in AI



Combinational Creativity

Combining existing ideas and things into something new



Explorative Creativity

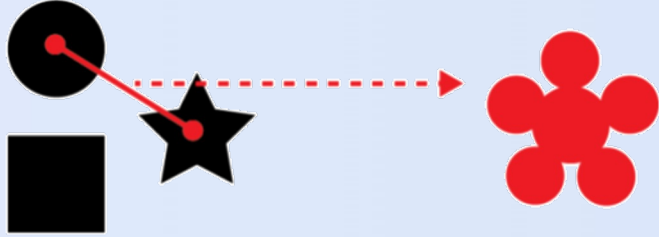
Exploring possibilities within a domain



Transformational Creativity

Radically new ideas that redefine the domain and existing rules

Combinational Creativity in AI

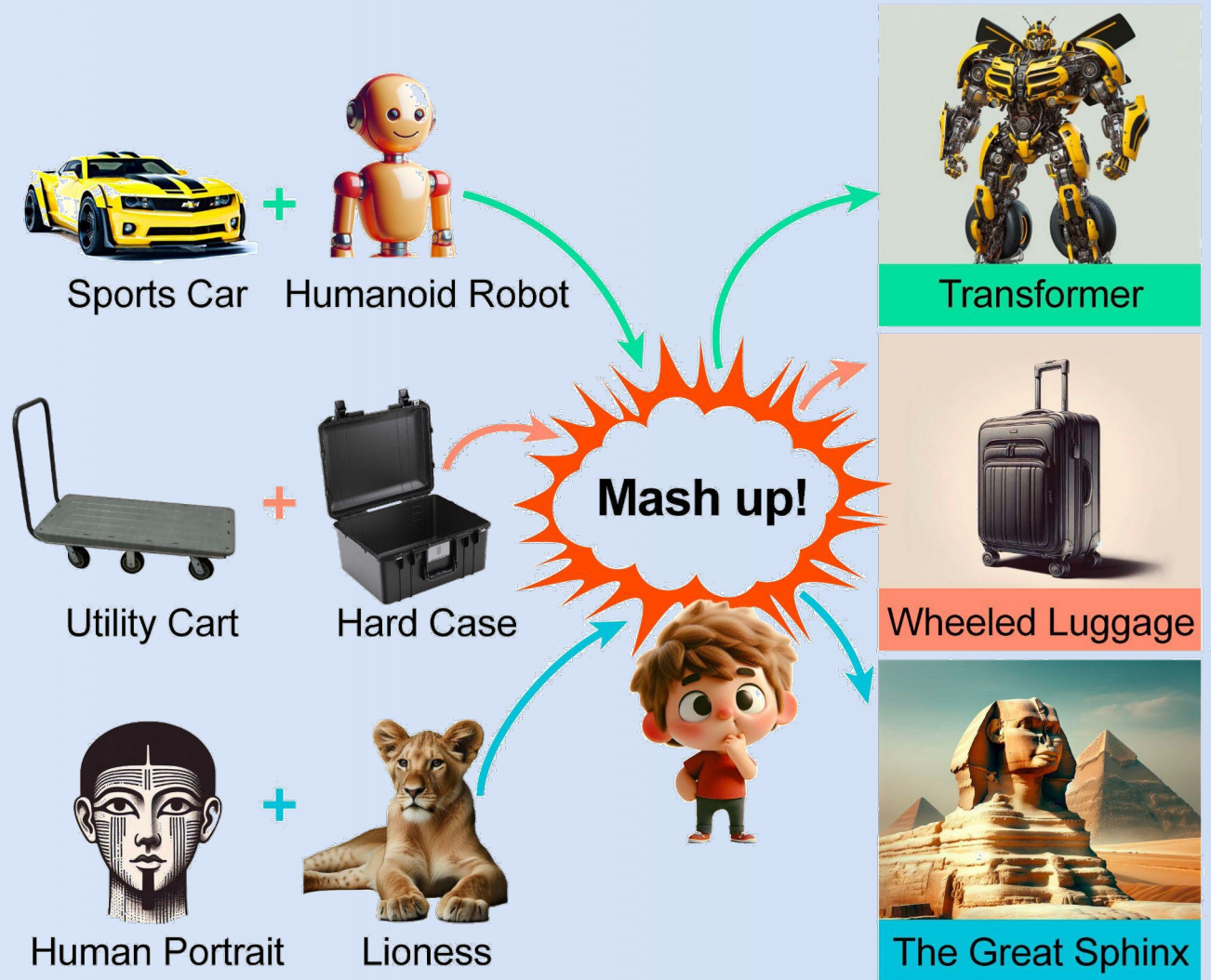


Boden, M. A. (2007). *Creativity in a nutshell. Think*, 5 (15), 83–96.

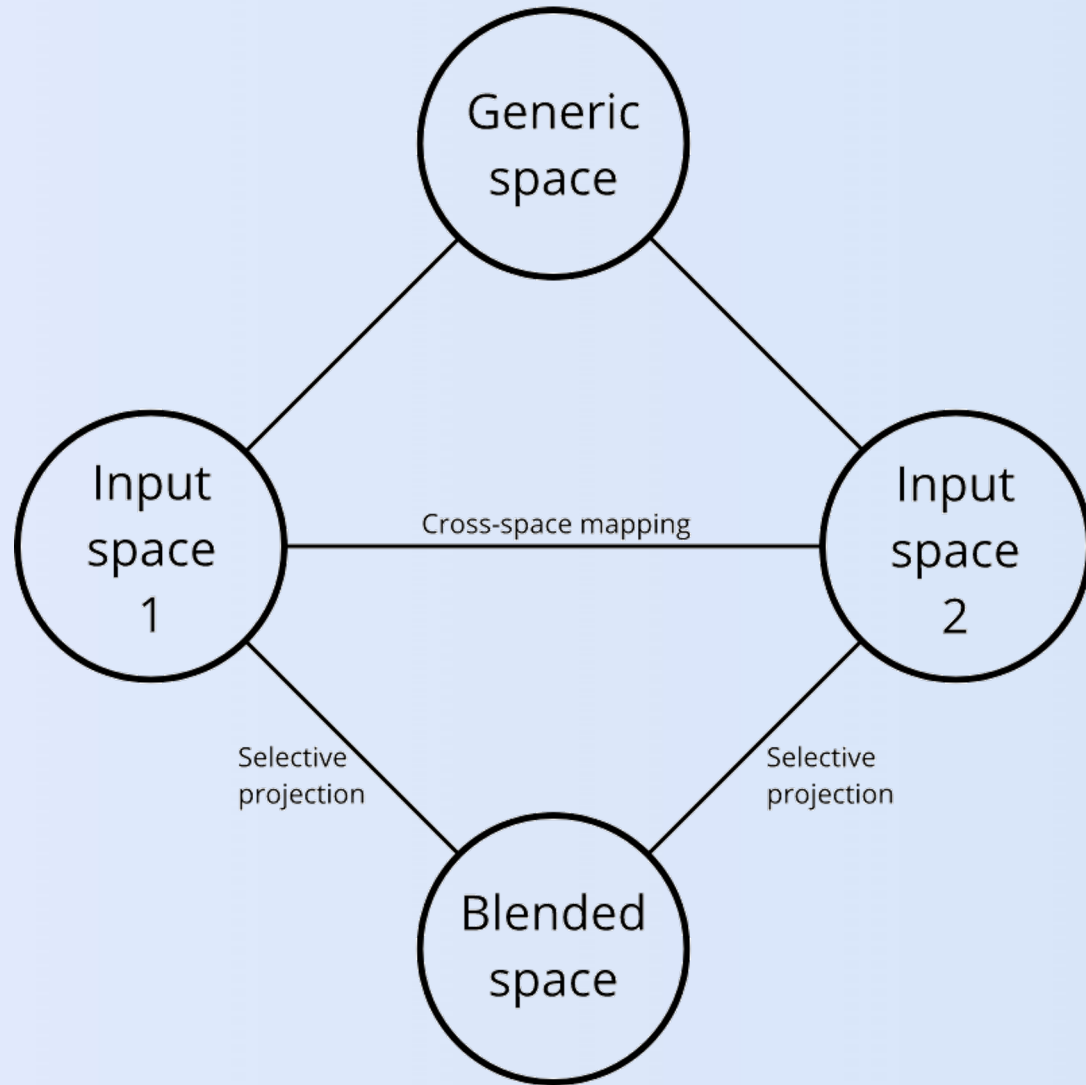
Combinational Creativity

Combining existing ideas and things into something new

- Well-defined
- Easy to implement
- Empirically dominant



Conceptual Blending Theory



How It Works: The Four-Space Model

1. Two Input Spaces

Each contains a separate scenario or concept.

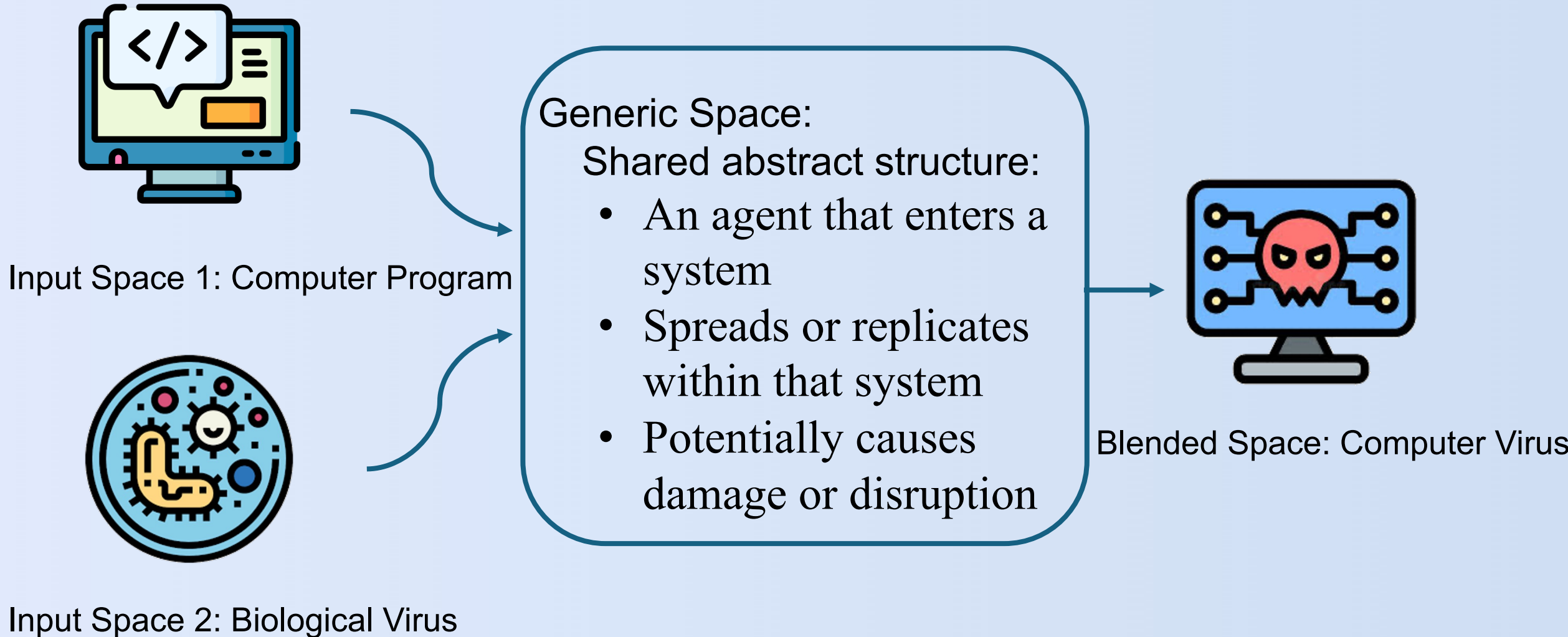
2. Generic Space

Captures what is common across both inputs—shared structure or schema.

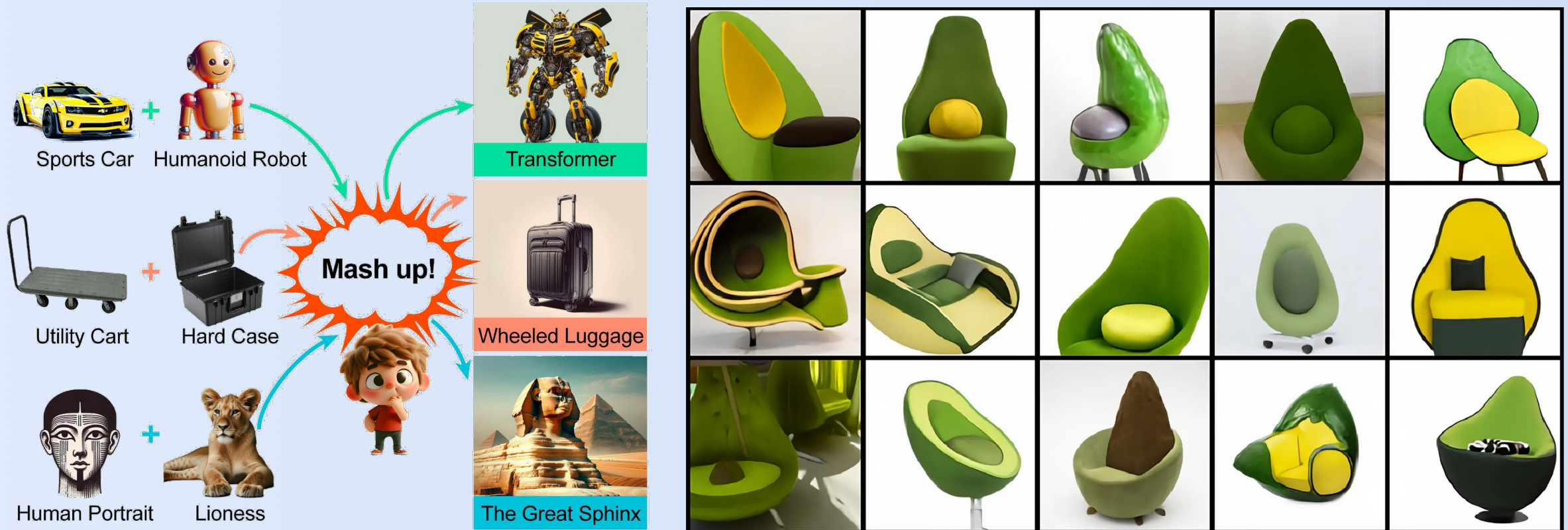
3. Blend Space

Selectively projects parts from the inputs into a **new space**, yielding emergent meaning.

Conceptual Blending Theory – An Example



Research Questions

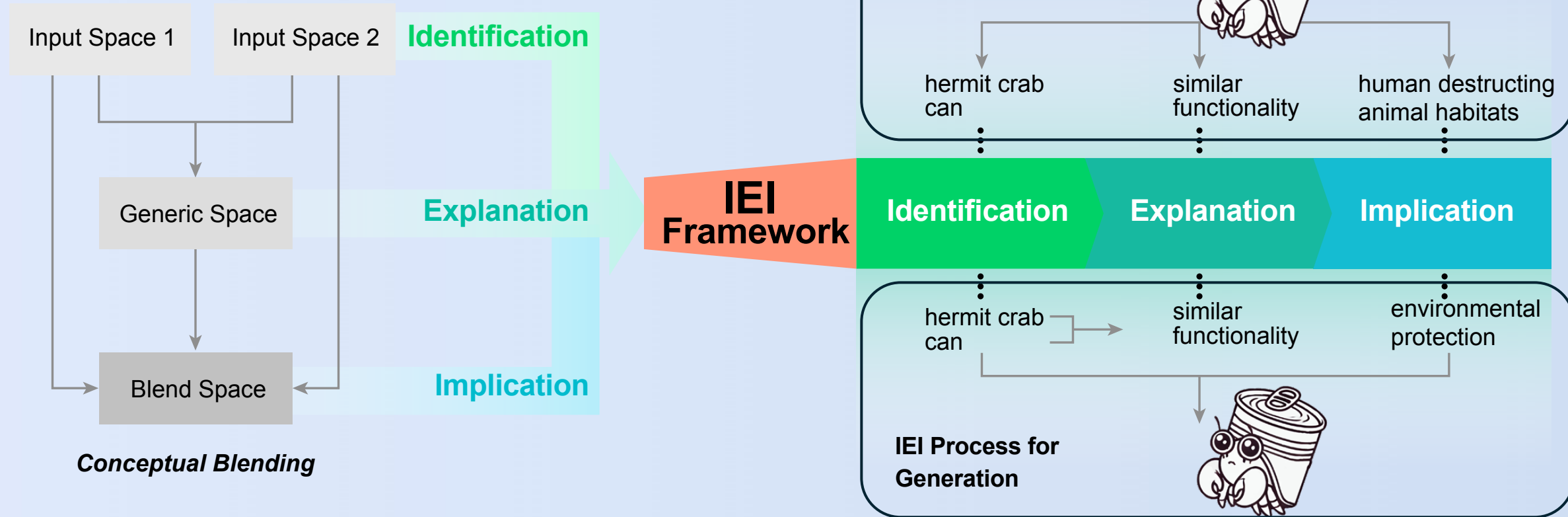


Peng, Y., Ma, Y., Wang, M., Wang, Y., Wang, Y., Zhang, C., ... & Zheng, Z. (2025). Probing and inducing combinational creativity in vision-language models. *arXiv preprint arXiv:2504.13120*.

<https://openai.com/index/dall-e/>

- To what extent can VLMs comprehend combinational creativity?
- Can the explicit integration of the combination process enhance models' ability to generate more creative products?

Overview



- We built a three-level framework of combinational creativity and a novel dataset containing mashup images for comprehensive analysis of how VLMs understand combinational creativity.
- We study whether explicitly incorporating this three-level framework into the generation process can enhance text-to-image models' ability to generate creative mashup images.

Three Levels of Combinational Creativity



Shared abstract structure:

- An agent that enters a system
- Spreads or replicates within that system
- Potentially causes damage or disruption



Identification-level — Input Space

Identify the objects used in the combination from the final product, answering: What objects are used for a combination?

Explanation-level — Generic Space

Explain the principles behind the combination, delving into relationships between objects, and answering: How does the combination work?

Implication-level — Blended Space

Examine the underlying meaning behind the combinational creativity product, answering: What is the meaning of the combination?

CreativeMashup Dataset

(a) Understanding Task



Identification

? Identify the primary objects in the image.

📷 Fish
Garbage

Explanation

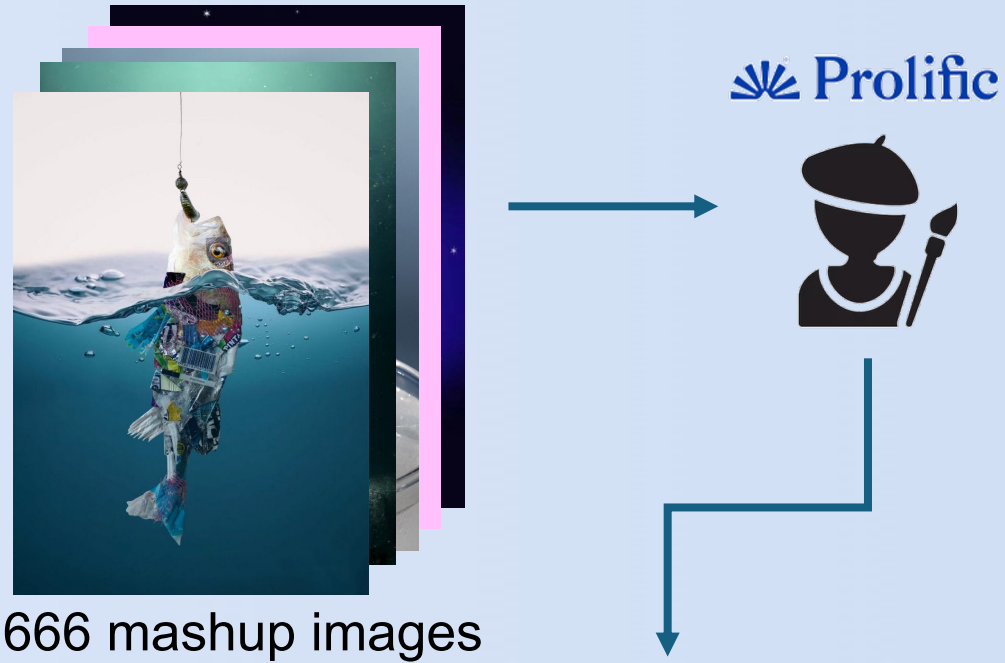
? Choose the relevant attributes that make the combination feasible.

📷 Shape

Implication

? Interpret what the combination of objects might be intended to convey.

📷 This image symbolizes how marine life is increasingly ingesting and being affected by human-made waste. This artwork likely aims to raise awareness about the severe consequences of environmental pollution and the urgent need for action to protect our oceans.



666 mashup images

```
{
  "image": "338.jpg",
  "answer": {
    "identification": [
      "fish",
      "garbage"
    ],
    "explanation": [
      "shape"
    ],
    "interpretation": [
      "This image symbolizes how marine life is increasingly ingesting and being affected by human-made waste. This artwork likely aims to raise awareness about the severe consequences of environmental pollution and the urgent need for action to protect our oceans."
    ]
  }
}
```

Annotated data

Understanding Task

(a)

Understanding Task



Identification

? Identify the primary objects in the image.

📷 Fish
Garbage

Explanation

? Choose the relevant attributes that make the combination feasible.

📷 Shape

Implication

? Interpret what the combination of objects might be intended to convey.

📷 This image symbolizes how marine life is increasingly ingesting and being affected by human-made waste. This artwork likely aims to raise awareness about the severe consequences of environmental pollution and the urgent need for action to protect our oceans.

Identification



Ground Truth

Mapping

Response

Explanation

Ground Truth

['shape', 'color',
'functionality' , ...]



Selected Option(s)

Implication



Implication
from Model I

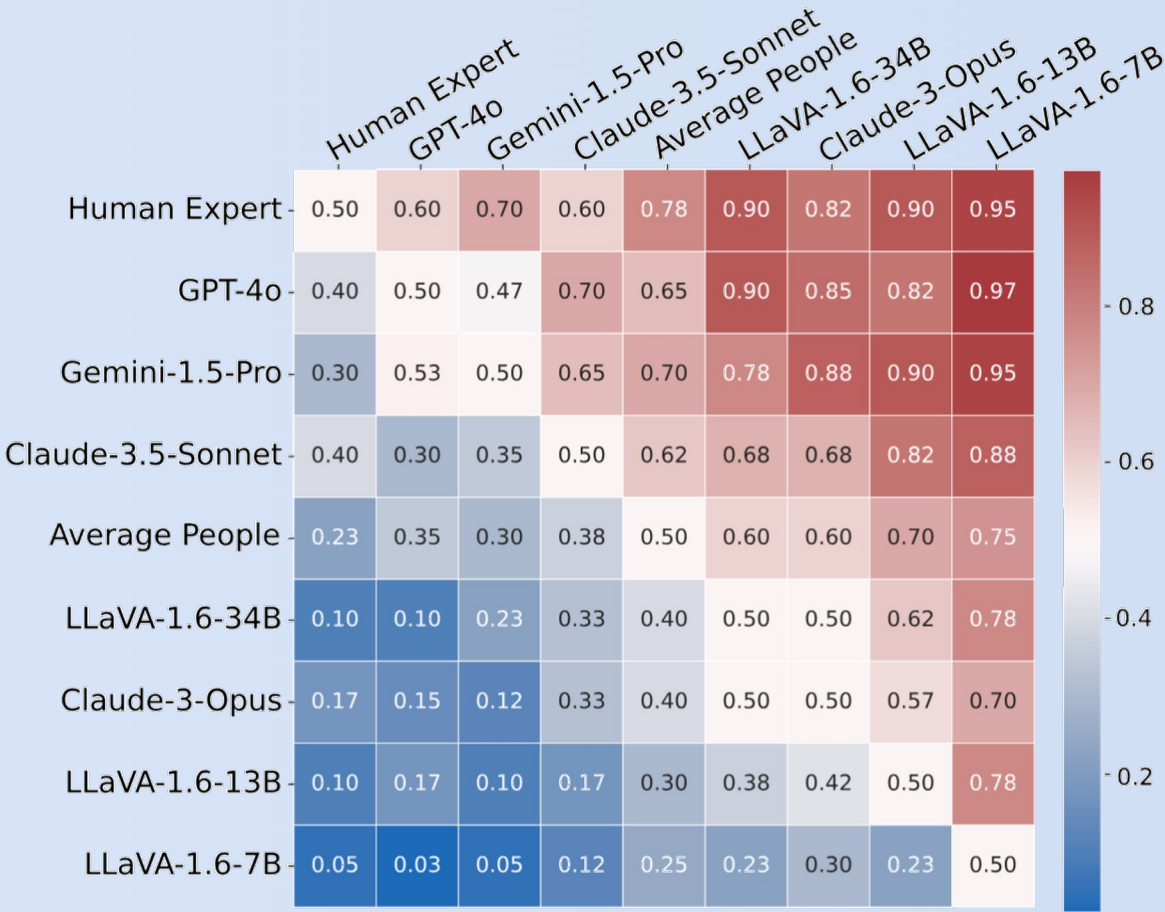


Pair-wise
comparison

Implication
from Model II

Do VLMs Understand Combinational Creativity?










Model	Identification		Explanation	Implication
	P↑	R↑	P↑	WR↑
Human Expert	-	-	-	78.3
Average People	53.42	70.33	69.89	51.0
GPT-4o [36]	75.67	85.00	74.19	73.5
GPT-4V [34]	60.83	75.00	63.44	71.9
Gemini-1.5-Pro [41]	73.67	81.33	54.34	71.7
Claude-3.5-Sonnet [3]	60.08	74.83	74.19	62.9
Claude-3-Opus [2]	63.17	72.50	65.59	39.2
LLaVA-1.6-34B [28]	64.67	72.17	62.37	40.6
LLaVA-1.6-13B [28]	60.33	67.33	40.86	34.3
LLaVA-1.6-7B [28]	50.33	57.83	48.39	20.8
LLaVA-1.5-7B [29]	49.62	63.00	43.01	20.1
MiniCPM [22]	64.40	72.33	50.54	41.7
Qwen-VL-Chat [4]	55.50	62.50	65.59	41.9



State-of-the-art models **have achieved human-level understanding** in combinational creativity.

Human experts still surpass models in the realm of combinational creativity.

Generation Task

(b) Generation Task			
Human Expert	Identification + Implication	Identification + Explanation + Implication	Three Levels of Combinational Creativity
			Identification Heart + Trash bag Explanation [Shape, Texture] Implication Pollution is detrimental to health
			Identification Pistol + Megaphone Explanation [Functionality, Shape] Implication Speech is powerful
			Identification Paper money + Mask Explanation [Shape] Implication Wealth can buy silence

RQ: Can the explicit integration of the combination process enhance models' ability to generate more creative products?

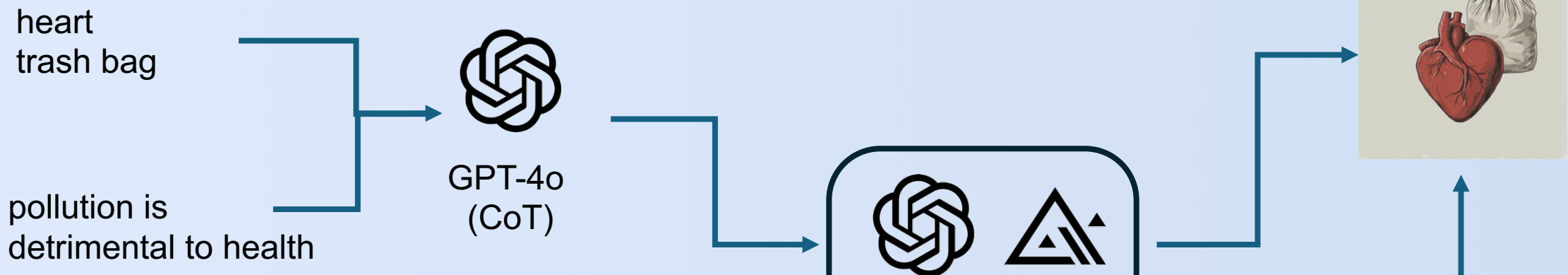
Human Expert

Identification + Implication
(Chain of Thought)

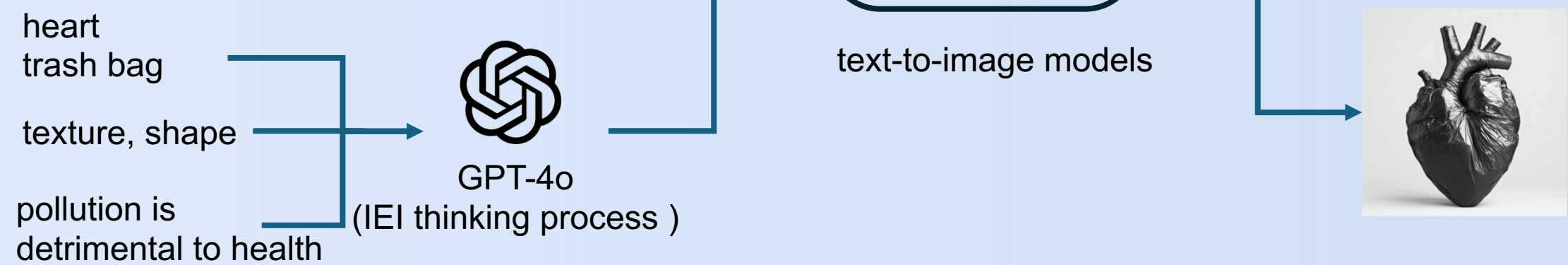
Identification + Explanation + Implication
(Conceptual Blending)

Generation Task

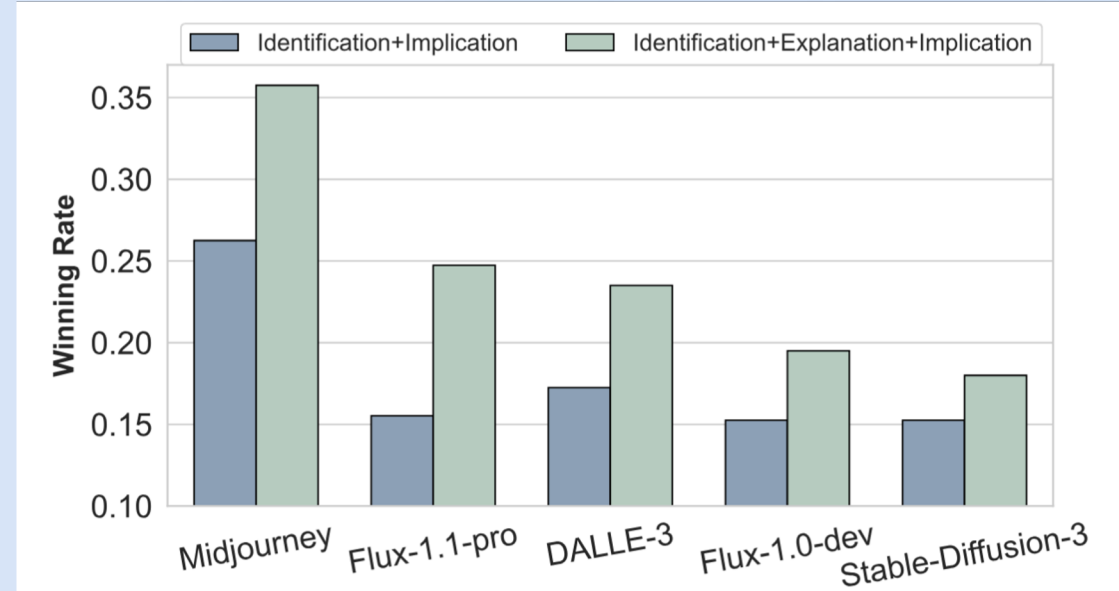
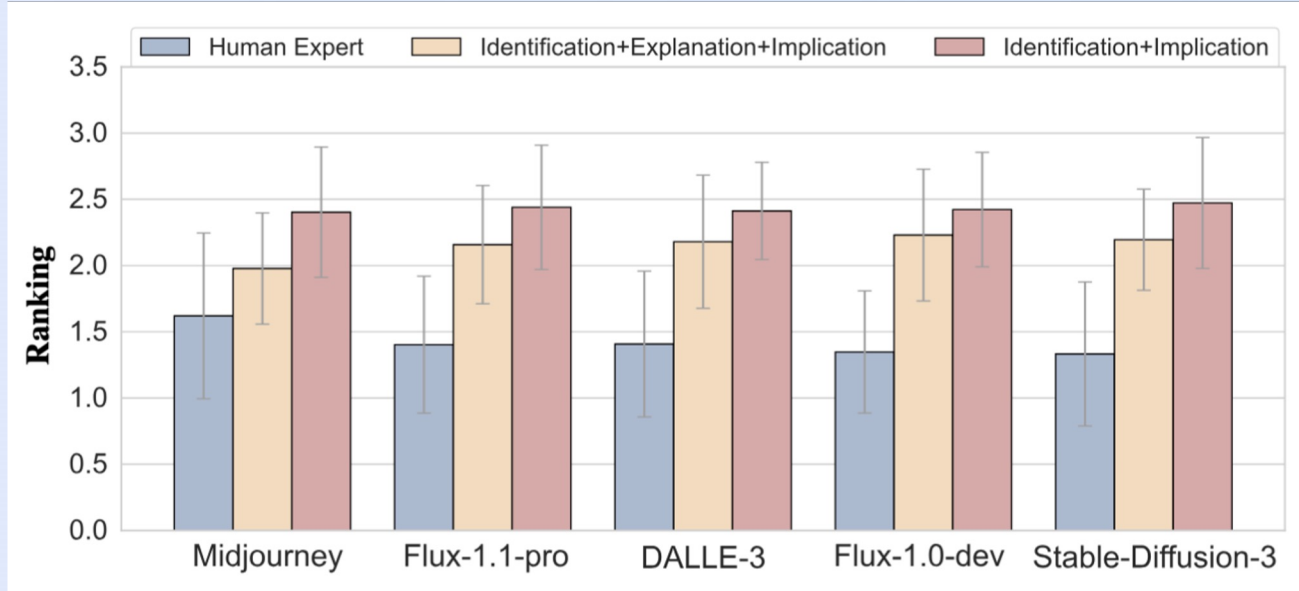
Identification + Implication (Chain of Thought)



Identification + Explanation + Implication (Conceptual Blending)



Results of Generation Experiment



The explanation level of combinational creativity can be leveraged to **enhance creativity without making the prompts significantly longer**.

Text-to-image models are currently the bottleneck in generating visual combinational creativity.

II (M = 487.95)
IEI (M = 514.53)
T-test stat: 0.84 p-value: 0.40

Text: 90% (36/40)
Text-to-image: 27.8% (10/36)

Take-away Message

- State-of-the-art models **have achieved human-level understanding** in combinational creativity, but still lag behind human experts.
- The explanation level of combinational creativity can be leveraged to **enhance creativity without making the prompts significantly longer**.
- **Text-to-image models** are currently the bottleneck in generating visual combinational creativity.

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