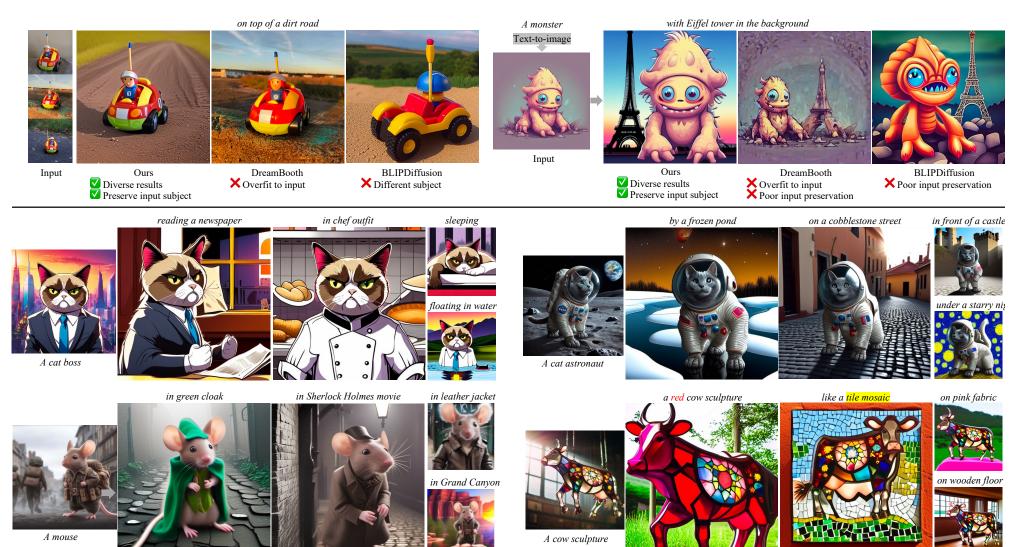




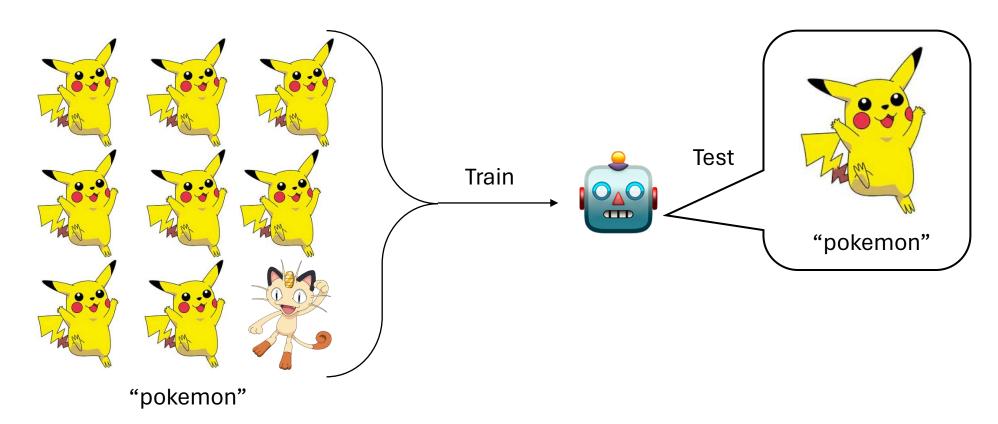
# JeDi: Joint-image Diffusion Models for Finetuning-free Personalized Text-to-image Generation

Yu Zeng, Vishal M. Patel, Haocheng Wang, Xun Huang, Ting-Chun Wang, Ming-Yu Liu, Yogesh Balaji



#### Limitation of learning-based generation

Challenges of rare concepts and novel concepts





## Limitation of learning-based generation







**SDXL** 



Pokemon meowth floating on top of the water



Dalle3 (ChatGPT)



**SDXL** 



Bojack horseman on the beach



## Represent new concepts

Use example images

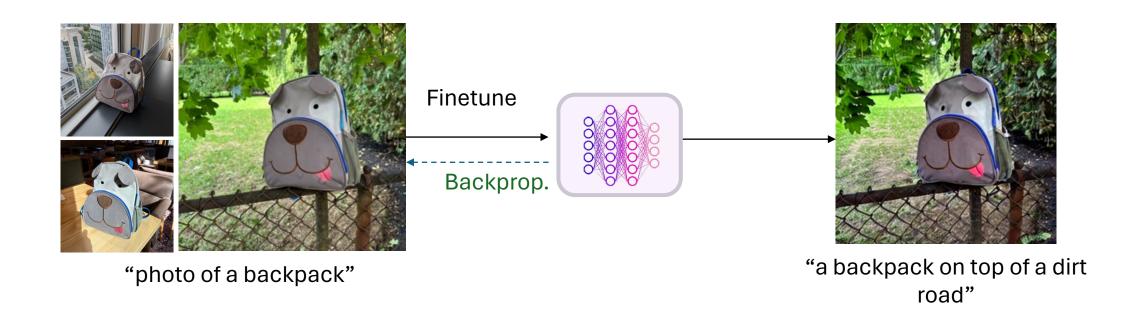






#### Adapt to new concepts

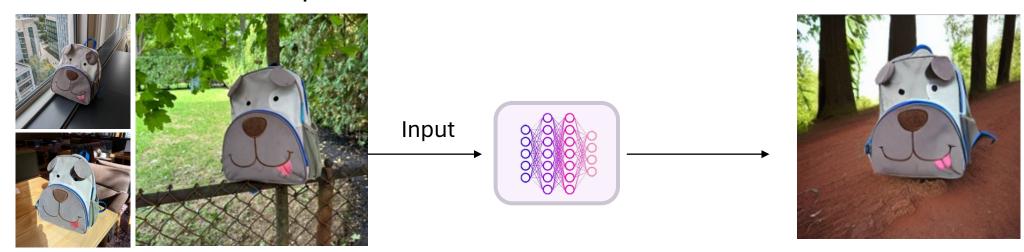
- Finetuning
  - Time- and resource- consuming
  - Overfitting to finetuning samples





### **Training-free fast adaptation**

- Finetuning
  - Time- and resource- consuming
  - Overfitting to finetuning samples
- Reference-based fast adaptation



"photo of a backpack"

"a backpack on top of a dirt road"

<u>Y. Zeng</u>, V. Patel, et al, "JeDi: Joint-image diffusion models for finetuning-free personalized text-to-image generation", CVPR, 2024.



<u>Y. Zeng</u>, Y. Balaji, T. Wang, X. Huang, M. Liu, "Neural networks to generate objects within different images," US Patent App. 18/518,430, Dec. 2023.

# **Training-free fast adaptation**

#### What's changing





Bojack horseman



New made-up monster



#### What's invariant





Same-subject relationship

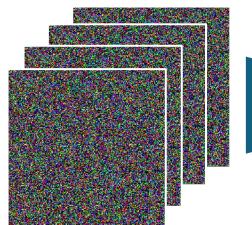


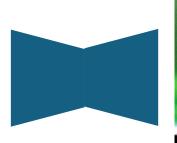
Individual

characters

## Joint-image diffusion

- Modeling the joint distribution of multiple images sharing the same concept
- Joint-image denoising diffusion





a red stuffed toy hanging in the window





a red stuffed toy on the ground



a red stuffed toy sitting on a window ledge

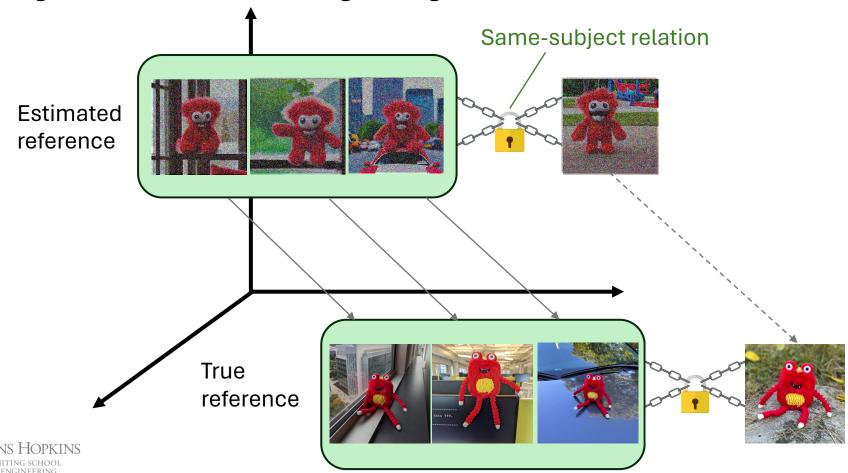


a red stuffed toy sitting on top of a car



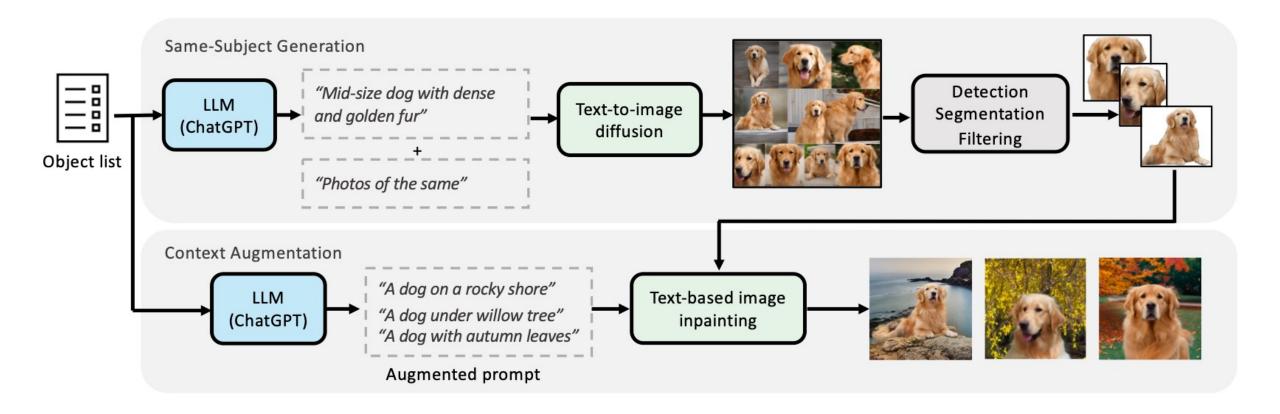
### Generation guided by reference

- Generate both the reference images and target image
- Using the true reference images as guidance



# Joint-image diffusion: training

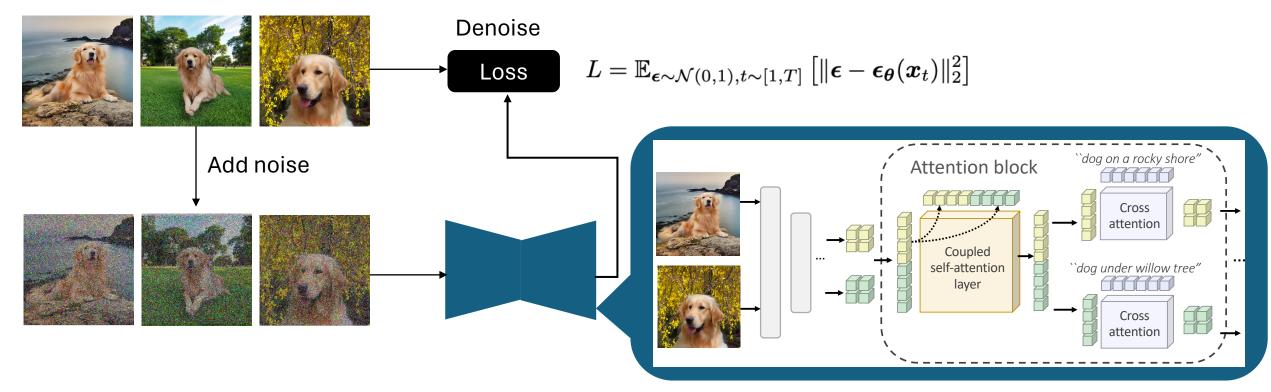
Training data: image sets





# Joint-image diffusion: training

- Modeling the joint distribution of an image set sharing the same concept
- Joint-image denoising diffusion

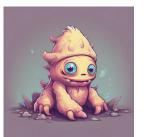


## **Training-free fast adaptation**

- Image synthesis and manipulation
  - Training-free fast adaptation

#### Unseen example images

Test-time ∏



OHNS HOPKINS
WHITING SCHOOL
of ENGINEERING









in jogging attire



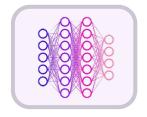
by a campfire in the woods

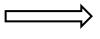


dressed in overalls



reading a newspaper





Pre-trained model

on top of a dirt road in a kitchen





#### Results

Rare concept generation
 Pokemon meowth



Pokemon meowth floating on top of the water







**SDXL** 

Ours

Bojack horseman



Bojack horseman on the beach







SDXL



Ours

#### Results

IOHNS HOPKINS

WHITING SCHOOL of ENGINEERING

Novel concepts generation







#### Underfitting

A backpack with a tree and autumn leaves in the background







A toy on top of a purple rug in a forest







Ours

Reference images

Finetuning (CD, DB)

14

#### Results

Novel concepts generation

Finetuning

Ours

Text alignment Concept preservation

Method	CLIP-T (†)	DINO (†)	MDINO (↑)
DreamBooth [27] Custom Diffusion [15]	0.2812 0.3015	0.6341 0.6343	0.7115 0.7109
JeDi (1 input)	0.3040	0.6190	0.7510
JeDi (3 inputs)	0.2932	0.6791	0.8037

13% improvement

.-----

2.82 second

Ours



15 minutes

Finetuning

320x faster

