



Navigating Text-To-Image Customization: From LyCORIS Fine-Tuning to Model Evaluation

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Q & A

Q: What does LyCORIS stands for?

A: LyCORIS is the abbreviation of *"Lora beYond Conventional methods, Other Rank adaptation Implementations for Stable diffusion"*

Q: Where does the name LyCORIS comes from?

A: LyCORIS refers to the flower Lycoris radiata, aka red spider lily, or "higanbana". In particular, the name does **NOT** come from the anime Lycoris Recoil.

Q: Why does someone claim LyCORIS leads to better results whereas the others claim the opposite?

A: There can be many reasons. To name a few, they may use different algorithms from LyCORIS, train on very different datasets, or run with different hyperparameters / optimization algorithms ...

Related Projects

- Automatic111/stable-diffusion-webui
- kohya-ss/sd-scripts
- cloneofsimo/lora

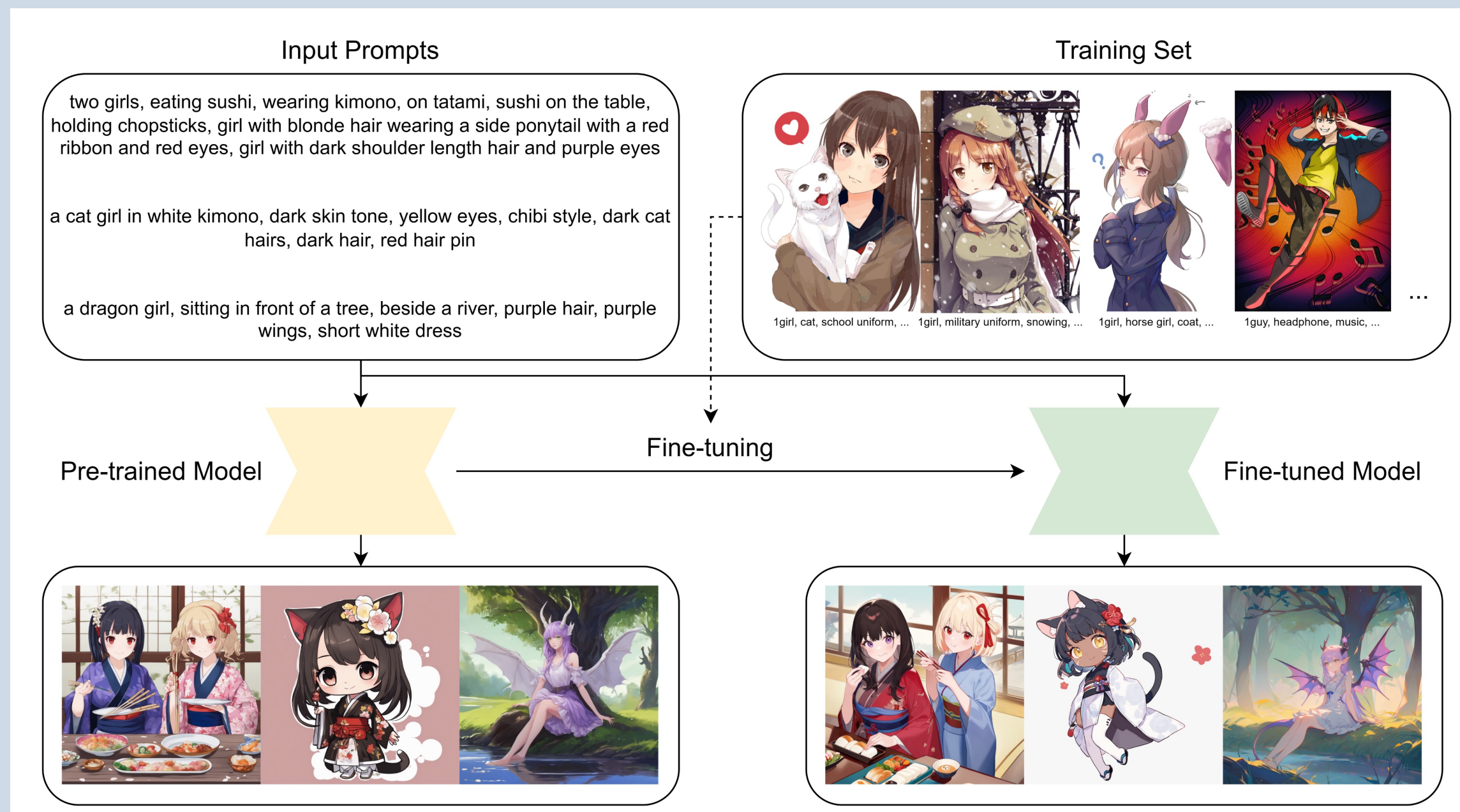
Algorithm References

- (LoRA) LoRA: Low-Rank Adaptation of Large Language Models. Hu et al., ICLR 2022.
- (IA³) Few-shot parameter-efficient fine-tuning is better and cheaper than in-context learning. Liu et al, NeurIPS 2022.
- (GLoRA) One-for-all: Generalized lora for parameter-efficient fine-tuning. Chavan et al, 2023.
- (DoRA) Dora: Weight-decomposed low-rank adaptation. Liu et al, 2024.
- (OFT) Controlling text-to-image diffusion by orthogonal finetuning. Qiu et al., NeurIPS 2023.
- (BOFT) Parameter-efficient orthogonal finetuning via butterfly factorization. Liu et al., ICLR 2024.

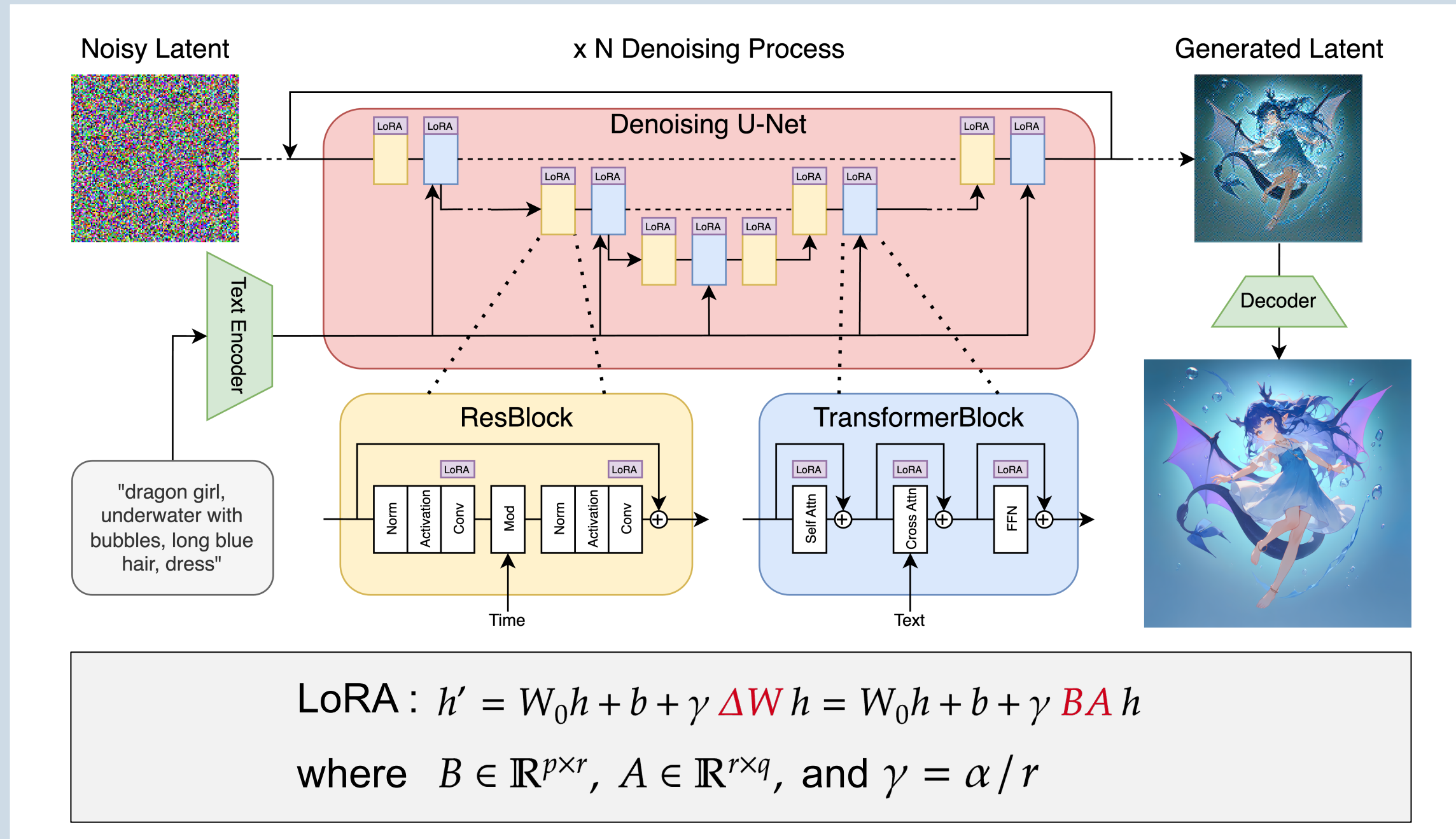
Image Credits

- 米望ゆめ / bogdan / Animenbo / ke-ta
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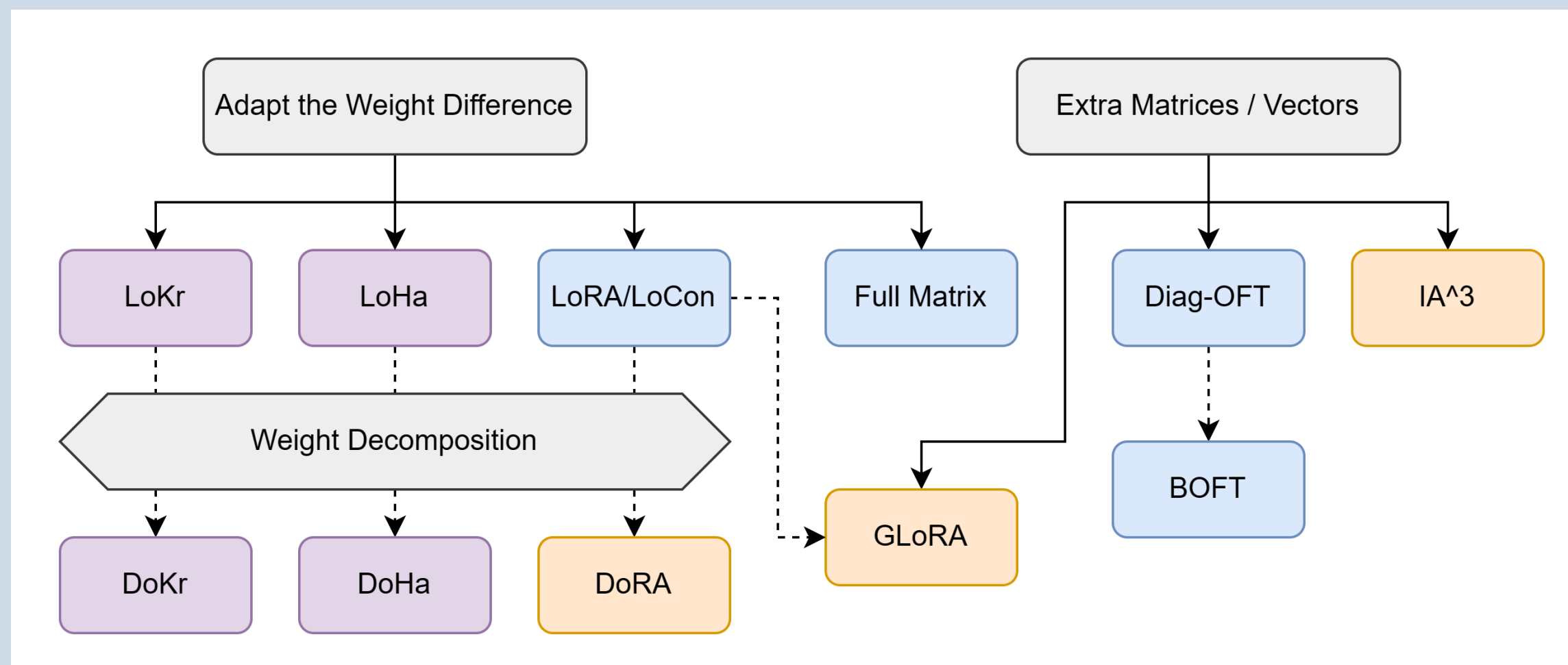
Customization of Text-To-Image Models



Stable Diffusion and LoRA Fine-Tuning



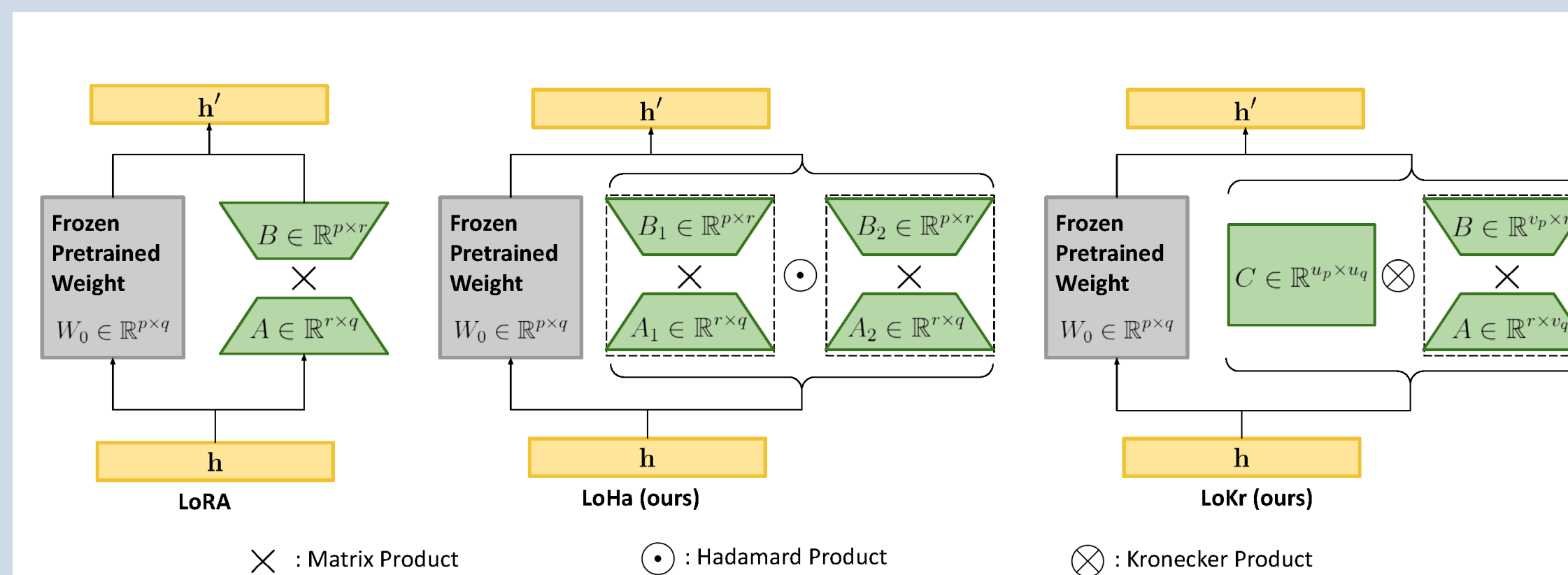
Algorithms Implemented in LyCORIS



TL;DR

- LyCORIS is an open-source library that promotes easy adaption of various fine-tuning algorithms for Stable Diffusion
- To enable fair comparison, we propose an extensive framework for model evaluation and conduct it for a subset of algorithms in LyCORIS

LoCon LoHa LoKr



$$\text{LoHa: } h' = W_0h + b + \gamma \Delta W h = W_0h + b + \gamma [(B_1 A_1) \odot (B_2 A_2)] h$$

$$\text{LoKr: } h' = W_0h + b + \gamma \Delta W h = W_0h + b + \gamma [C \otimes (BA)] h$$

Theorem (Effect of γ)

Assume that forward pass is modified to

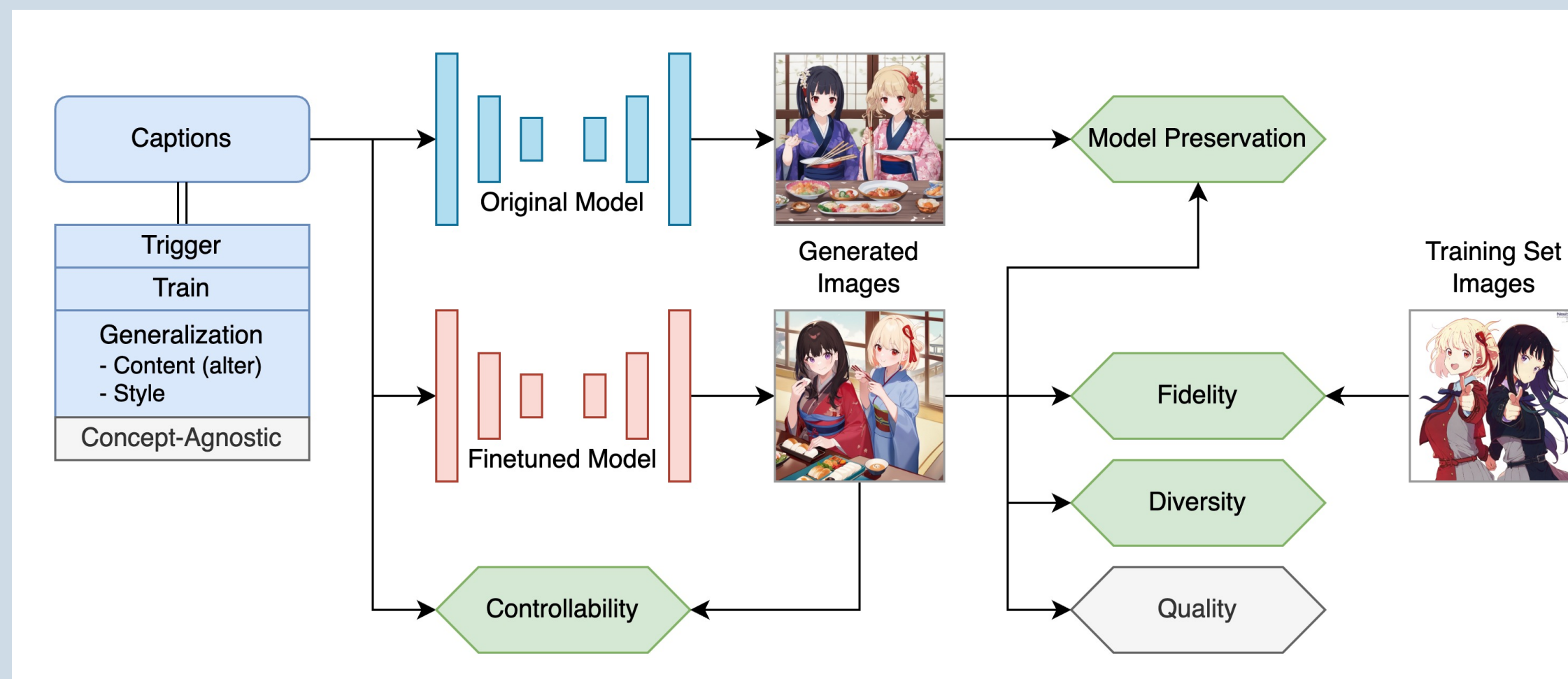
$$h' = W_0h + b + \gamma \Delta W h = W_0h + b + \gamma \mathcal{T}(A_1, \dots, A_m) h$$

where \mathcal{T} is a homogeneous operator

Mathematically, it is equivalent to training without γ and

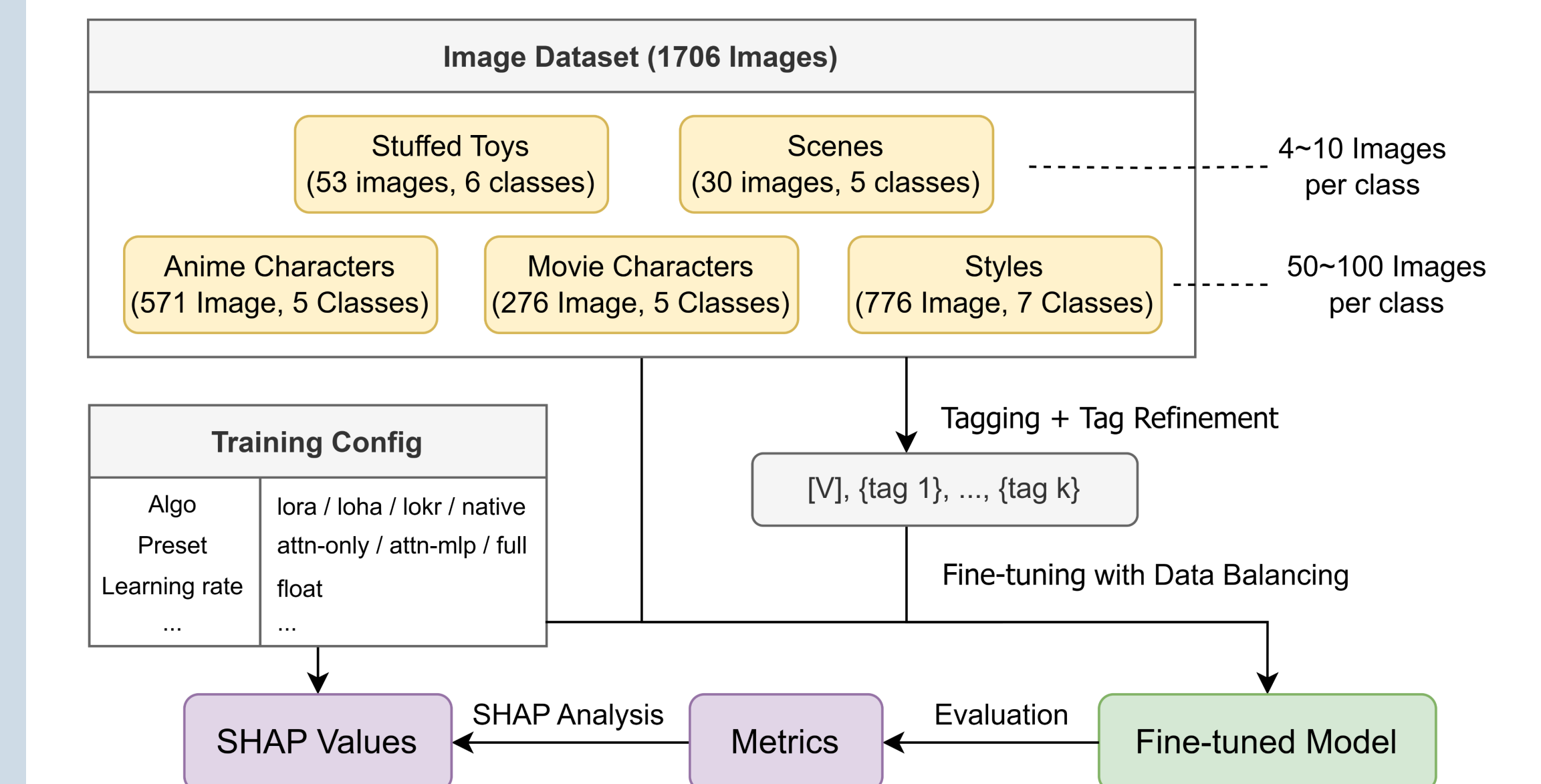
- scaling the initialization parameters by $\gamma^{1/m}$
- scaling the learning rate by $\gamma^{c/m}$, where $c=2$ for SGD and $c=1$ for Adam, etc

Evaluation Criteria



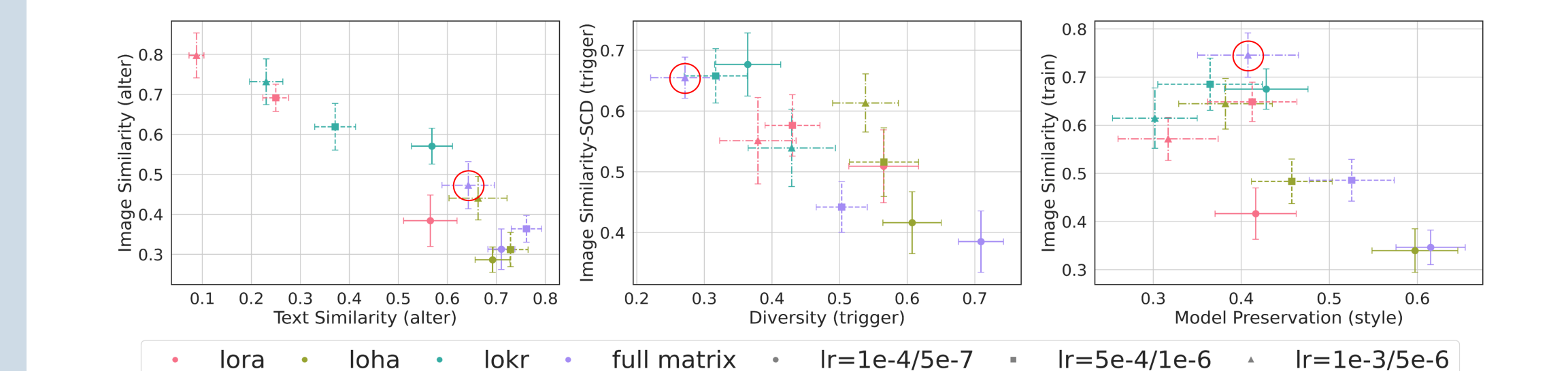
Experiments

Setup

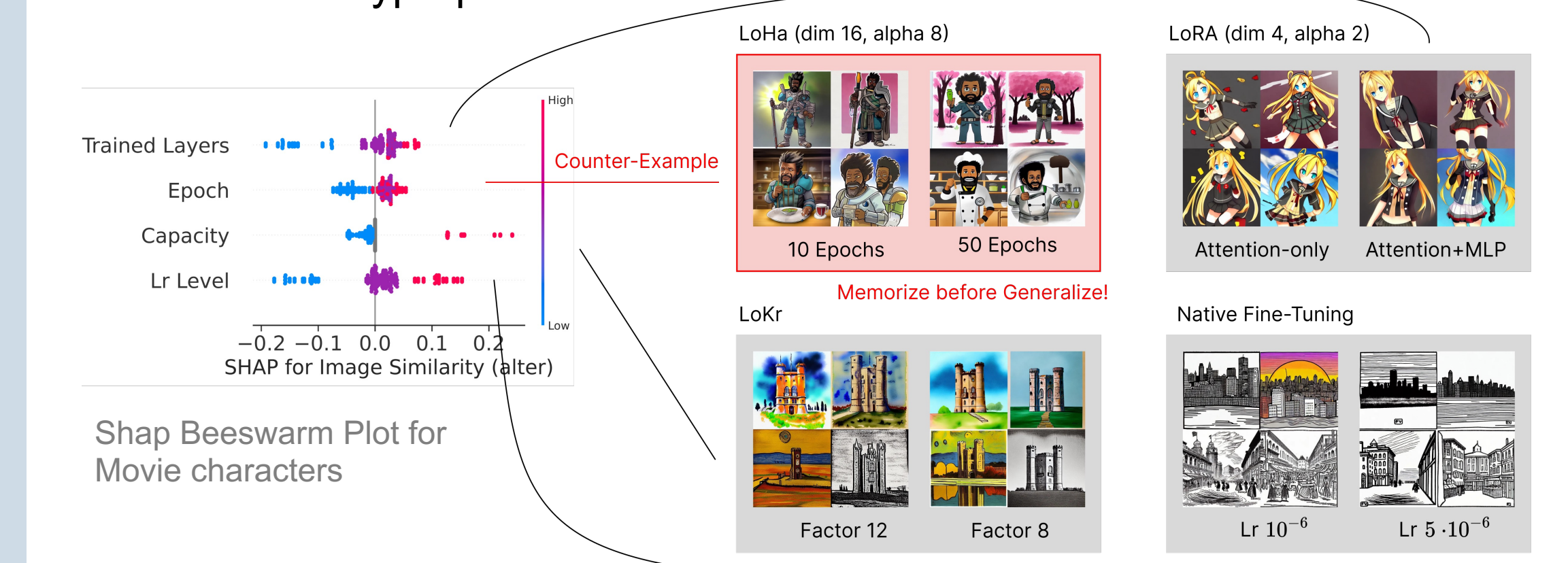


Results

- High-Level Observations
 - Image Fidelity is mostly negative correlated with other metrics
 - Choice of prompt can affect significantly the evaluation



Influence of Hyperparameters



Tentative Ranking

	LoRA	LoHa	LoKr	Native Lr 5e-6	Native Lr 1e-6
Fidelity	★★★	★★	★★★★	★★★★★	★
Controllability	★★	★★★★	★	★★★	★★★★★
Diversity	★★★	★★★★	★★	★	★★★★★