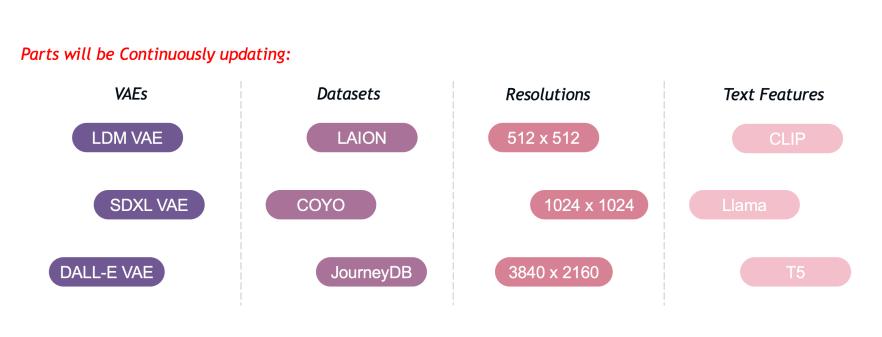


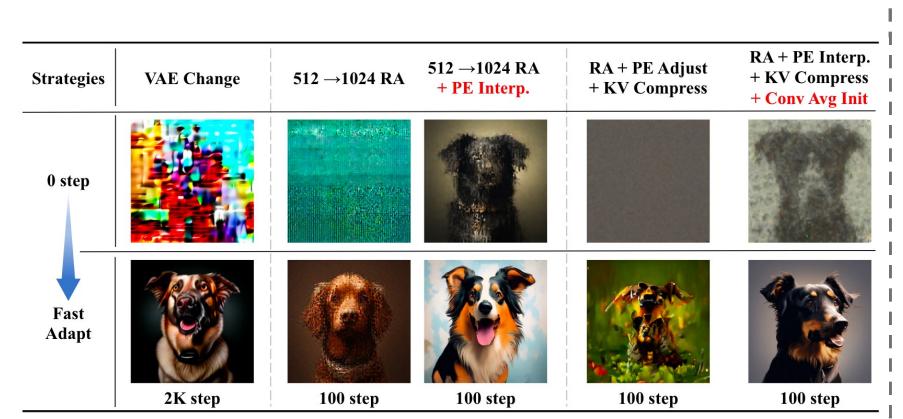


Problem Statement

Our Approach



(a) Multiple elements we may change in development

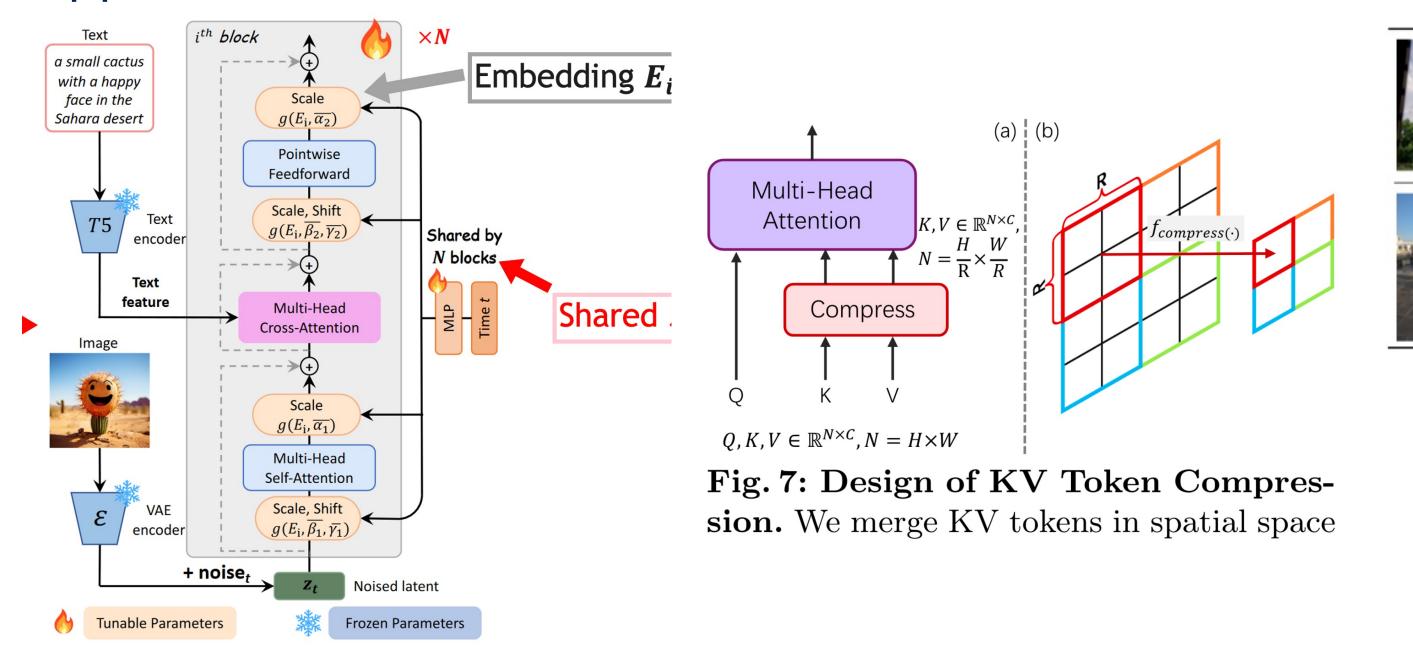


(b) Weak-to-Strong Training Strategy

PIXART- Σ is capable of directly generating images at 4K resolution. it evolves from the 'weaker' baseline to a 'stronger' model via incorporating higher quality data, a process we term "weak-to-strong training".

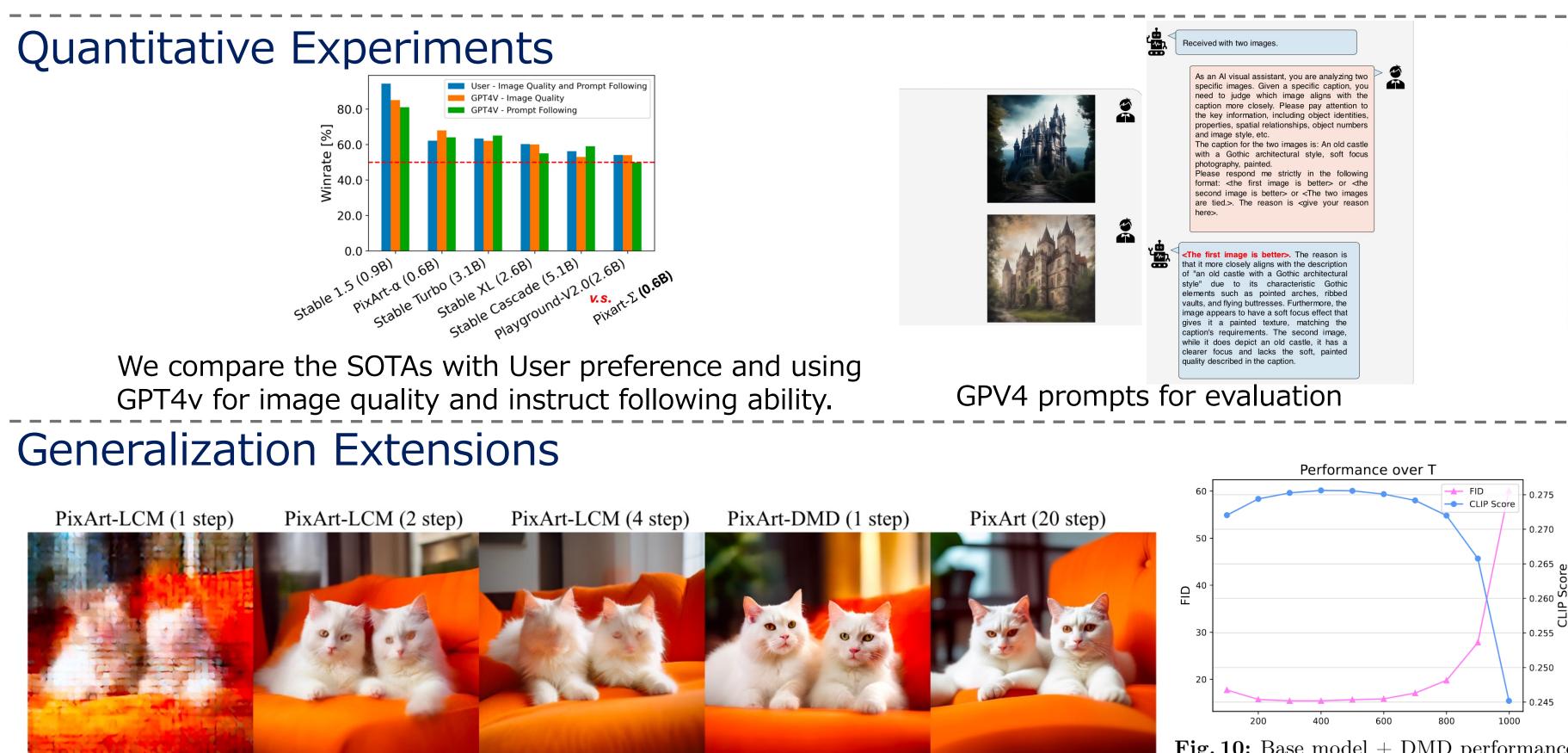
Contributions

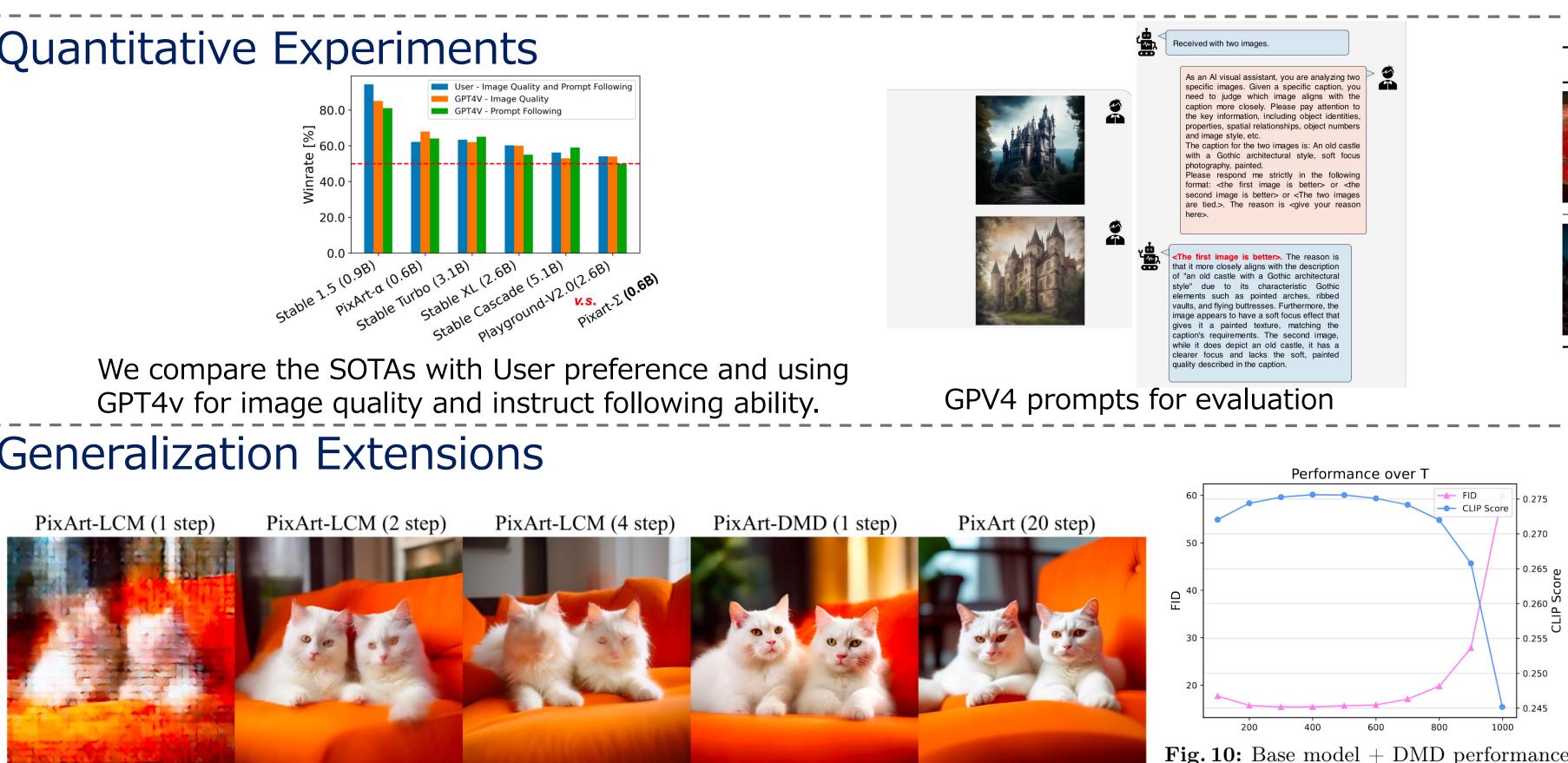
- We collect a high-quality dataset superior to that used in PixArt-a, focusing on **High-quality** images and Dense and accurate captions.
- We introduce an **Efficient Token Compression** to effectively reduce training and inference time by ~34% for high-resolution image generation.
- We propose Weak-to-Strong Training Strategy utilizing several fine-tuning techniques to rapidly adapt from a weak model to a strong model efficiently



Model architecture of PIXART-S. We merge KV tokens in spatial space in self-attention block to reduce the Comparative illustration of hallucinations: Contrasting differences in hallucination occurrences between computation complexity, specially for high-resolution image generation LLaVA and Share-Captioner, with red indicating hallucinations and green denoting correctness

over 7





Prompt: two white cats playing on top of the orange sofa, very comfortable Best quality.

PIXART-Σ: Weak-to-Strong Training of Diffusion Transformer for 4K Text-to-Image Generation

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Efficient T2I Transformer

PixArt-DMD 1 step generation

	LLaVA Hallucinations	The image features a large, ornate church with a tall, pointed roof and a large stained-glass window . The church is situated in a city, and it has a white and gray color scheme. The style of the church is Gothic, which is evident from the pointed roof and the intricate details of the stained-glass window . The presence of statues and a clock on the building further emphasizes the grand and historical nature of the structure.
	Correctness	The image captures the grandeur of a cathedral, painted in a pristine white, stands majestically against the backdrop of a clear blue sky. The entrance to the cathedral is adorned with a large rose window. Flanking the entrance are statues of saints, standing as silent guardians of the sacred space within. The perspective of the photo, taken from a low angle. The lush green trees in the background adding serenity to the scene.
	Hallucinations	The image features a woman and a man sitting on a brick walkway near a body of water, which could be a river or a lake. They are both wearing head coverings, and the woman is holding a handbag. The scene is set during the day, with the sun shining brightly, creating a warm and inviting atmosphere. The style of the image is a black and white photo, which adds a timeless and classic feel to the scene.
	Share-Captioner Correctness	The image captures a serene scene at a harbor. Two individuals are seated on a bench, their backs to the camera, engrossed in the view of the water. The water, a deep shade of blue, is dotted with boats of various sizes and colors, including a white boat with a green stripe and a red boat. The sky above is a light blue.

Table 1: Statistics of noun concepts for different datasets. VN: valid distinct nouns (appearing more than 10 times); **DN**: total distinct nouns; **Average**: average noun count per image; ACL: Average Caption length

Dataset	Volume	Caption	\mathbf{VN}/\mathbf{DN}	Total Noun	ACL	Average
Internal- α	14M	Raw	$187 \mathrm{K}/931 \mathrm{K}$	175M	25	$11.7/\mathrm{Img}$
Internal- α	14M	LLaVA	$28\mathrm{K}/215\mathrm{K}$	536M	98	$29.3/\mathrm{Img}$
Internal- α	14M	Share-Captioner	$51\mathrm{K}/420\mathrm{K}$	815M	184	$54.4/\mathrm{Img}$
Internal- Σ	33M	Raw	$294\mathrm{K}/1512\mathrm{K}$	$485\mathrm{M}$	35	$14.4/\mathrm{Img}$
Internal- Σ	33M	Share-Captioner	$77\mathrm{K}/714\mathrm{K}$	$1804 \mathrm{M}$	180	$53.6/\mathrm{Img}$
$4 ext{K-}\Sigma$	2.3M	Share-Captioner	$24\mathrm{K}/96\mathrm{K}$	115M	163	$49.5/\mathrm{Img}$

analyzing two caption, you	Image	Prompt	Image	Prompt	Models	#Params (]	B) FID \downarrow C	CLIP-Score
gns with the v attention to ect identities,	0	A red onelo citting on a		A photographic work	Stable 1.5	0.9	17.03	0.2748
An old castle	A red apple sitting on a wooden table, remote			capturing a polar bear walking through icy and snowy terrain.	Stable Turbo	3.1	10.91	0.2804
e, soft focus the following	and the second sec	control aerial photography.			Stable XL	2.6	7.38	0.2913
ter> or <the e two images your reason</the 					Stable Cascade	5.1	9.96	0.2839
	on ral hic ed he he he he he he he he he he	A serene beach with palm trees, turquoise water,		A bird known for its distinctive blue and orange plumage. The kingfisher is perched on a branch, its body angled slightly to the left as if poised	Playground-V2.0	2.6	8.68	0.2885
ason is scription itectural			turquoise water, hammock between		Playground-V2.5	2.6	7.64	0.2871
Gothic ribbed ore, the		and a hammock between two trees, star trail.			$PIXART-\alpha$	0.6	8.65	0.2787
fect that ng the image, t has a			to take flight at any moment.	PixArt - Σ	0.6	8.23	0.2797	

Table 4: Comparison of PIXART DMD performance compared to **PIXART** + **LCM.** These experiments are conducted on 512x512 resolution with a batch size of 1.

\mathbf{Method}	FID↓	$\mathrm{CLIP}\uparrow$	Speed↓
PIXART + LCM (1 step)	108.66	0.2247	0.11s
PIXART + LCM (2 step)	17.95	0.2736	0.16s
PIXART + LCM (4 step)	13.06	0.2797	0.26s
PIXART + DMD (1 step)	13.35	0.2788	0.11s
Teacher model (20 steps)	9.273	0.2863	1.44s

Training Details

Stage	Image Resolution	ה #Images T	Training Step	os Batch Size	Learning Rat	e GPU days
VAE adaption Better Text-Image align	$\begin{array}{ c c c c c } 256 \times 256 \\ 256 \times 256 \end{array}$	33M 33M	8K 80K	64×16 64×16	2×10^{-5} 2×10^{-5}	5 V100 50 V100
Higher aesthetics Higher aesthetics KV token compression	$ \begin{vmatrix} 512 \times 512 \\ 1024 \times 1024 \\ 1024 \times 1024 \end{vmatrix} $	18M 18M 18M	10K 5K 5K	$\begin{array}{c} 32 \times 32 \\ 12 \times 32 \\ 12 \times 16 \end{array}$	2×10^{-5} 1×10^{-5} 1×10^{-5}	30 V100 50 V100 20 V100
Higher aesthetics KV token compression Higher aesthetics KV token compression	$\begin{array}{c c} 2K \times 2K \\ 2K \times 2K \\ 4K \times 4K \\ 4K \times 4K \end{array}$	300K 300K 100K 100K	4K 4K 2K 2K	4×8 4×8 4×8 4×8	2×10^{-5} 2×10^{-5} 2×10^{-5} 2×10^{-5}	20 A800 14 A800 25 A800 20 A800

Timestep T need to be set to T=400 during training than T = 999 for better convergence



Appealing Generations



We create a 30K PixEval dataset for High quality FID and CLIP-Score assessment