

The Power of Context: How Multimodality Improves Image Super-Resolution

CVPR 2025

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Vishal M. Patel², Peyman Milanfar¹, Mauricio Delbracio¹

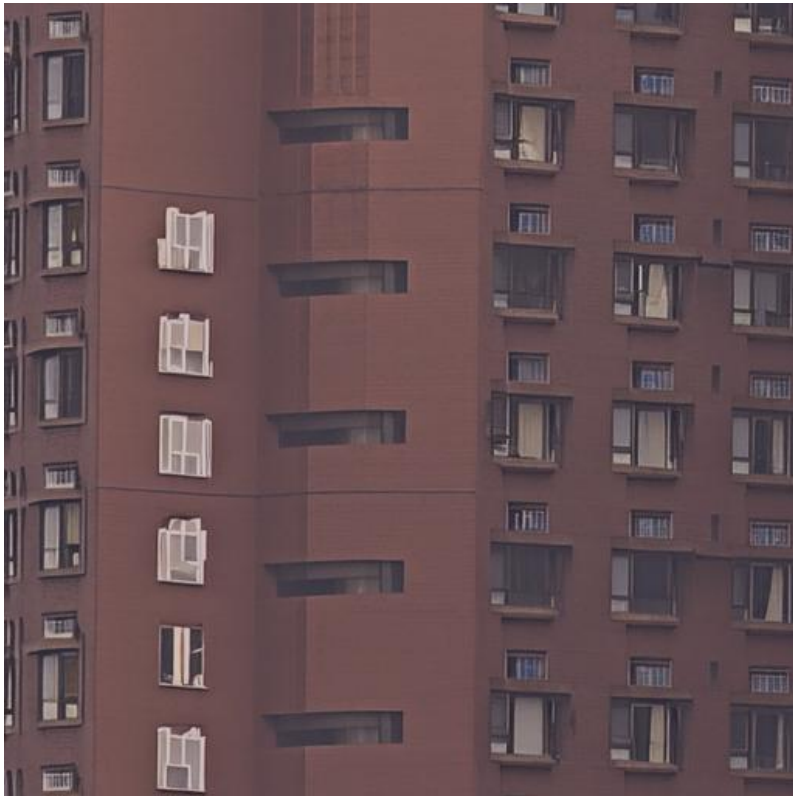
¹ Google, ² Johns Hopkins University

Project Page: <https://mmsr.kfmei.com/>

Presented by Zejia Fan
2025.10.12

Background

- Super resolution with diffusion needs semantic guidance



DIT4SR



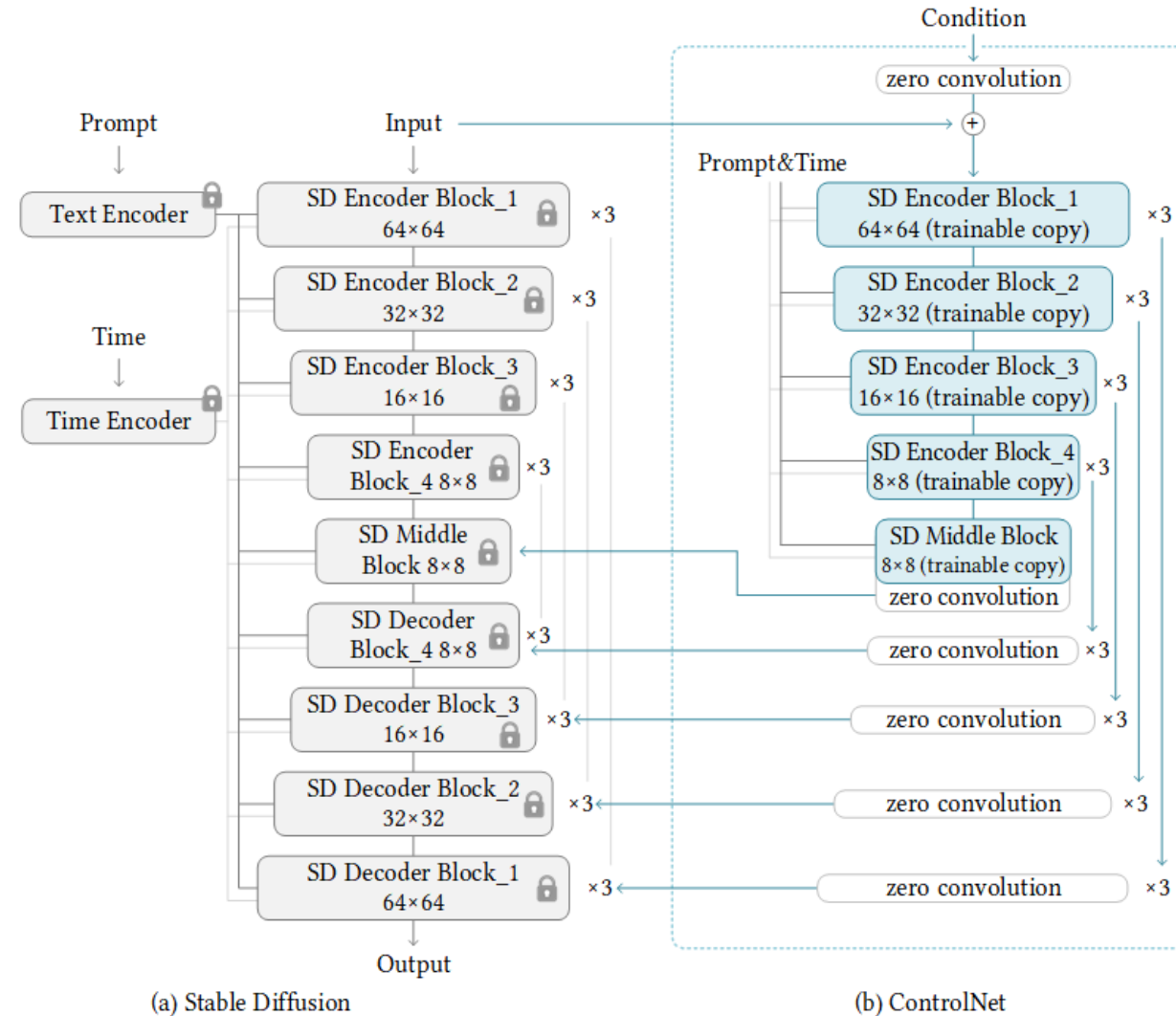
HR



LR

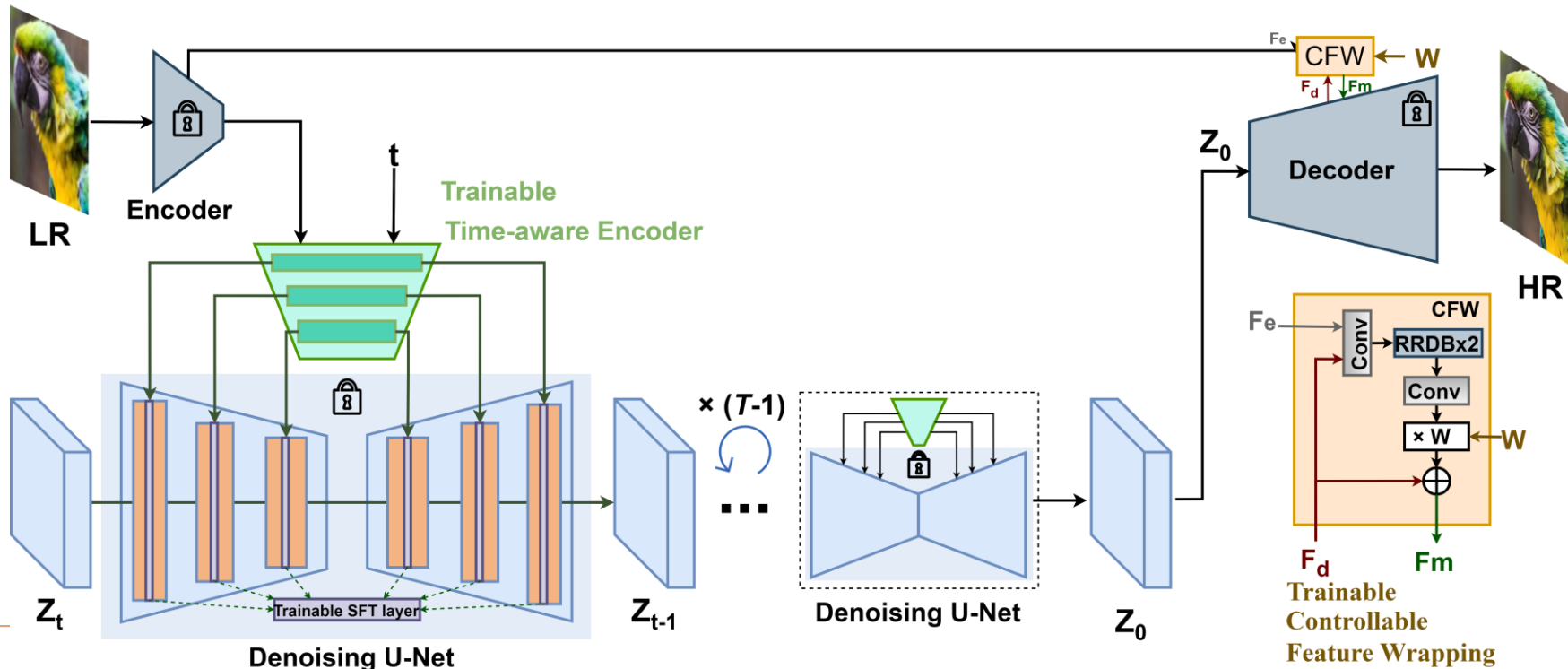
Background

- SR with diffusion
- Add LR as condition
 - SD-up sampler with additional LR channel added directly
- ControlNet



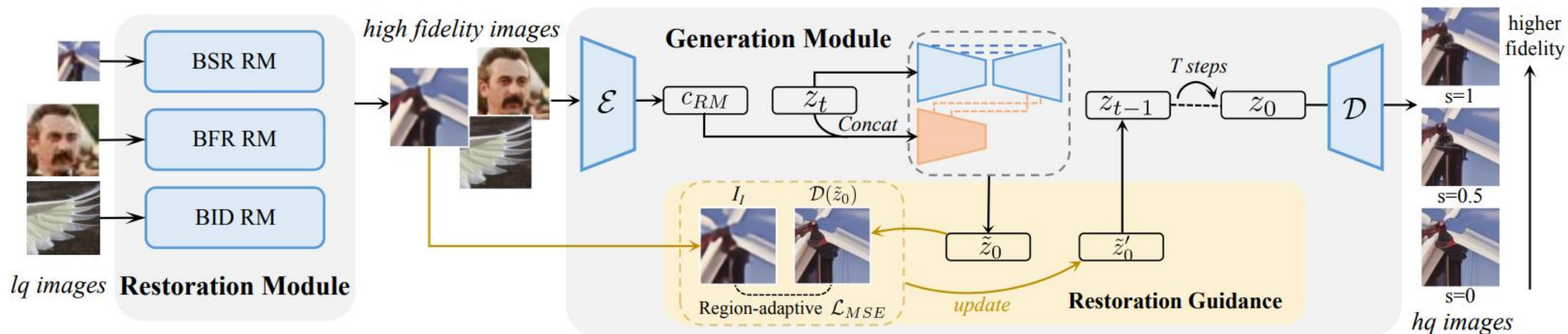
Background

- StableSR
 - IJCV 2024



Background

- DiffBIR
 - ECCV 2024



Background

- SUPIR
 - CVPR 2024

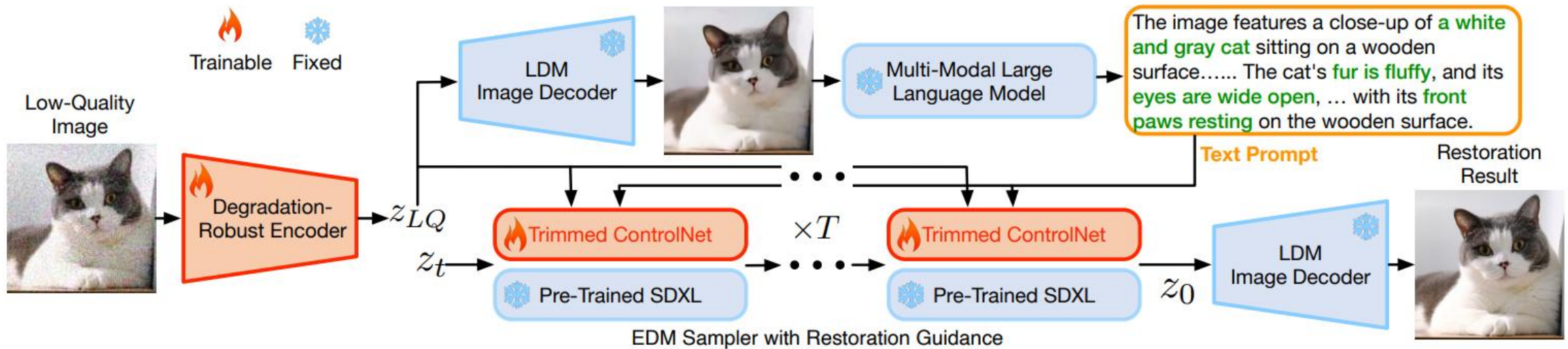
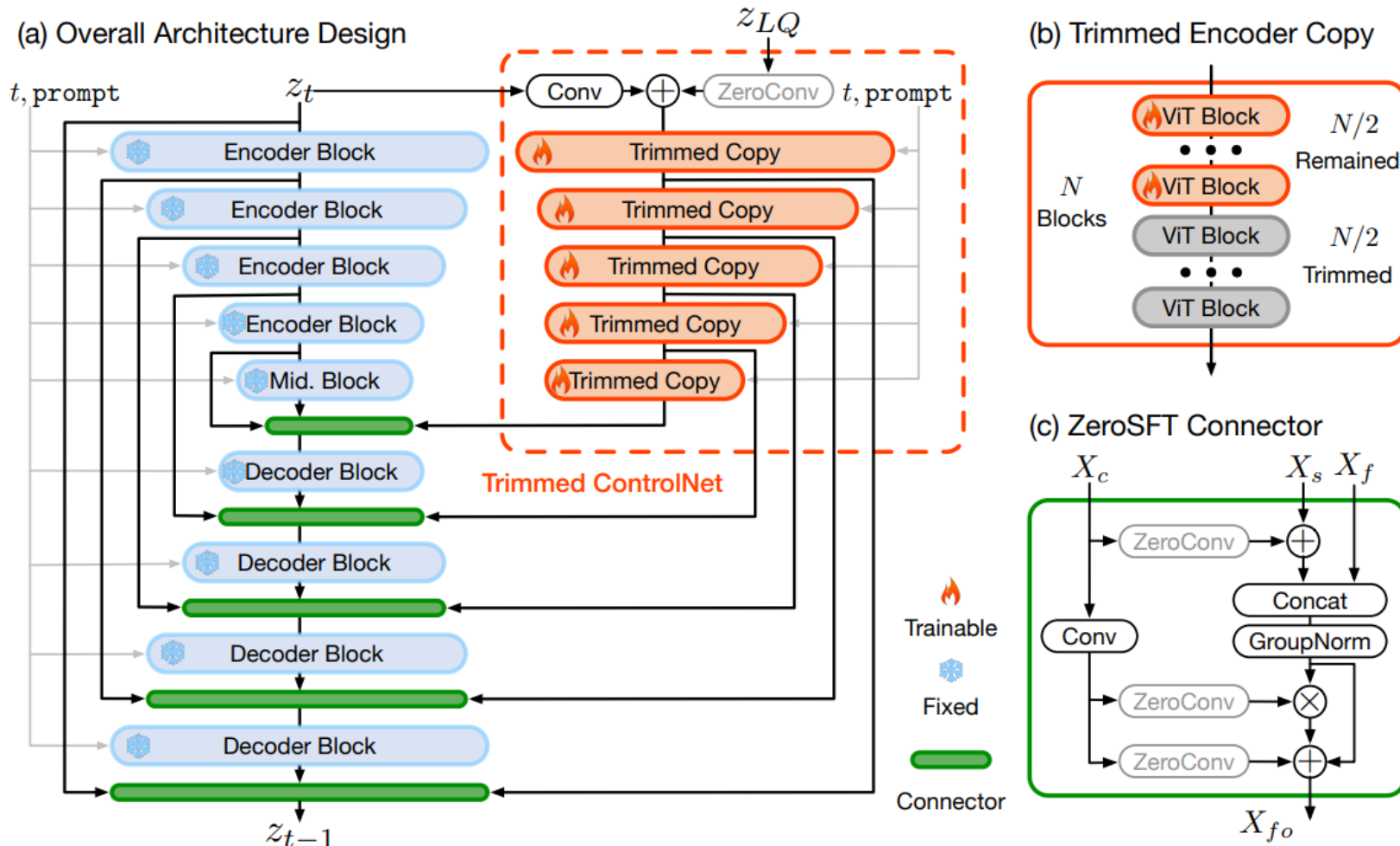


Figure 2. This figure briefly shows the workflow of the proposed SUPIR model.

Background

- SUPIR
 - CVPR 2024

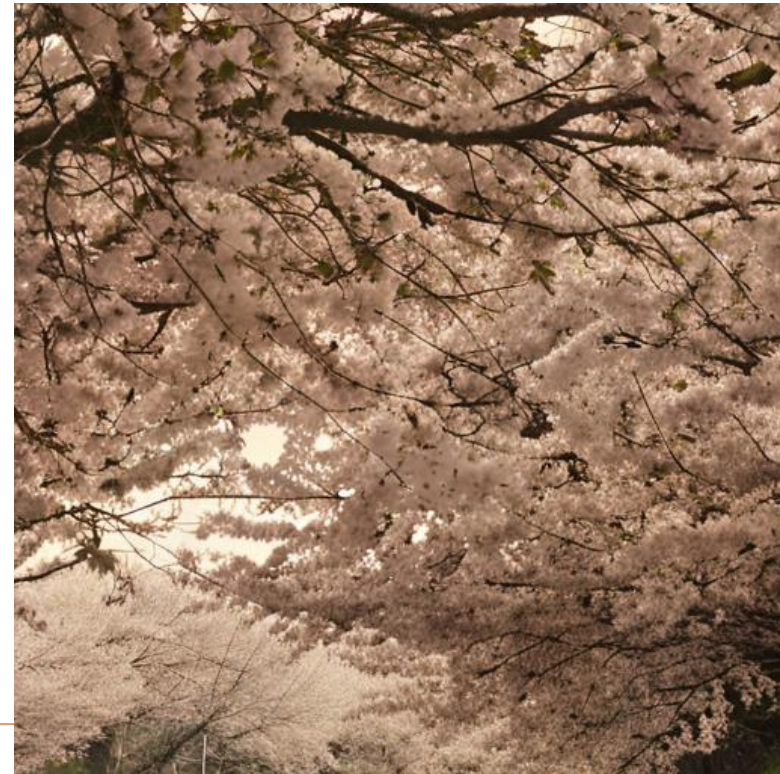


Background

- SUPIR
 - CVPR 2204



LR



result

Background

- SUPIR
 - CVPR 2204



LR



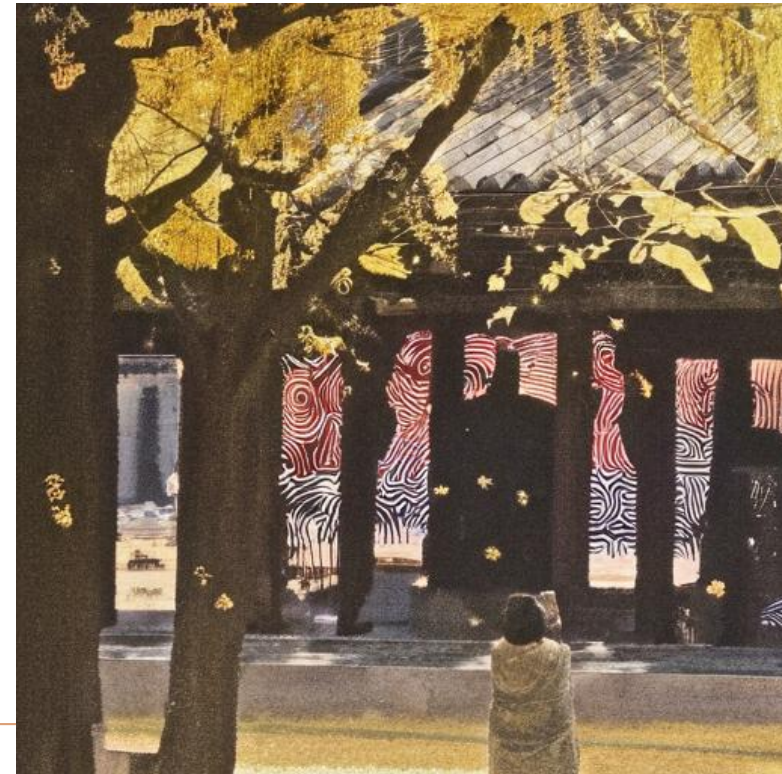
result

Background

- SUPIR
 - CVPR 2204



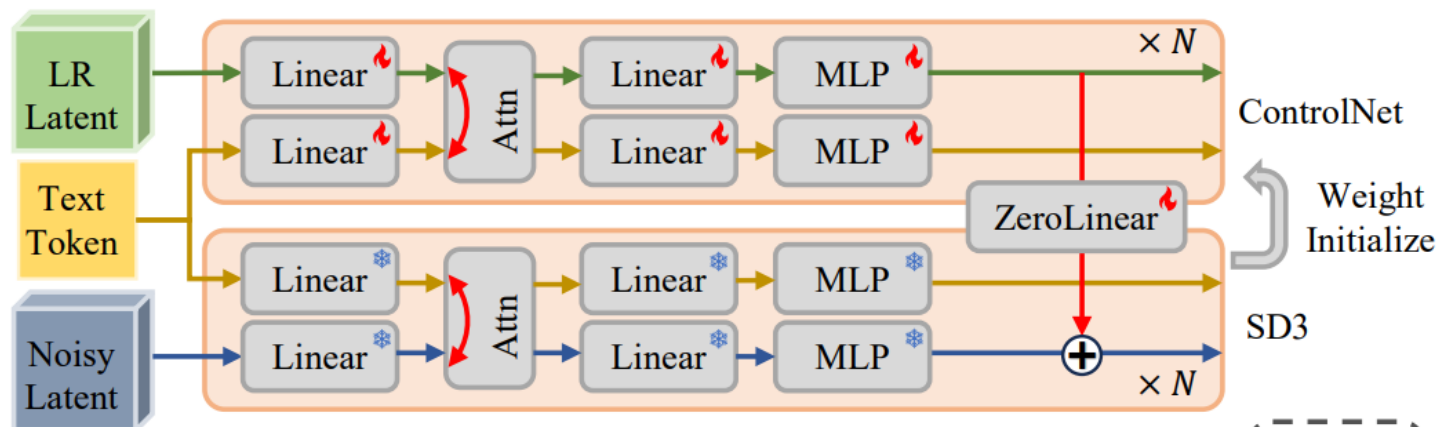
LR



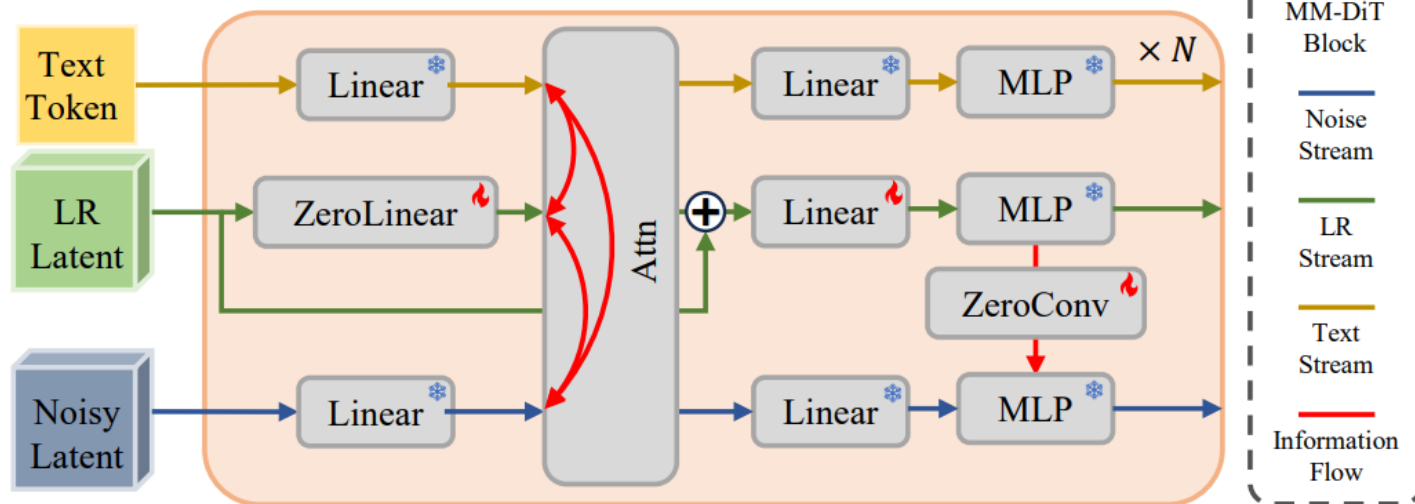
result

Background

- DiT4SR
 - ICCV 2025



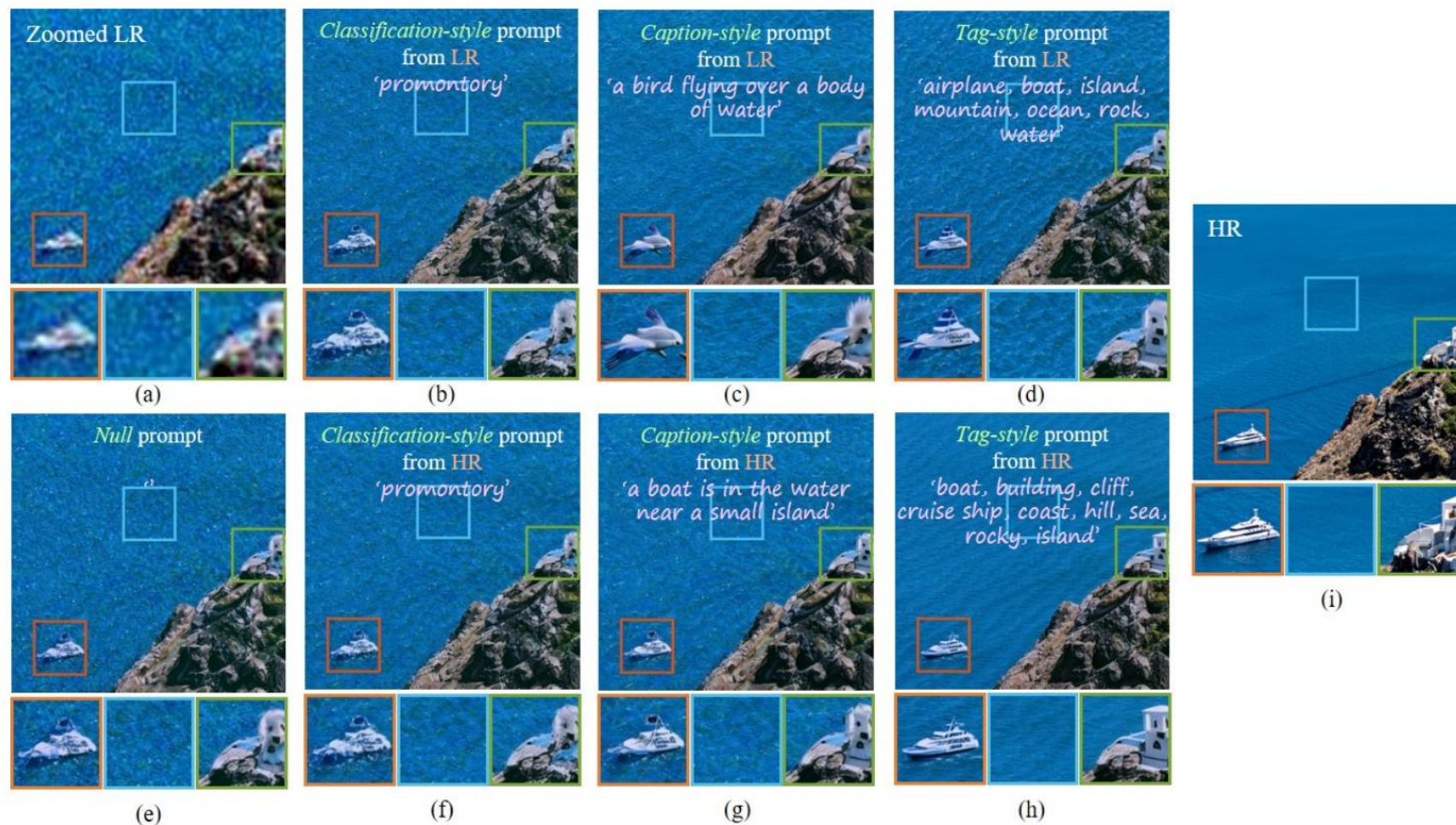
(a) SD3-ControlNet [68]



(b) DiT4SR

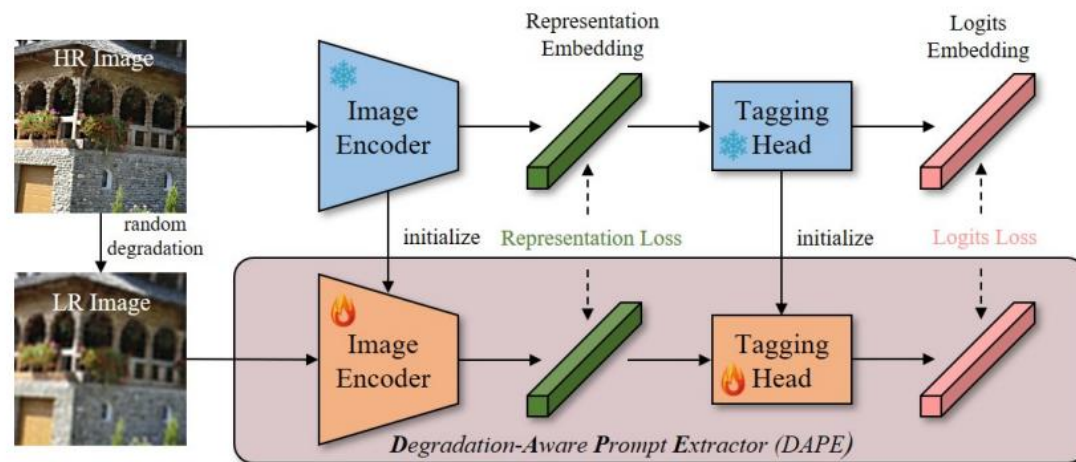
Background

- SeeSR
 - CVPR 2024

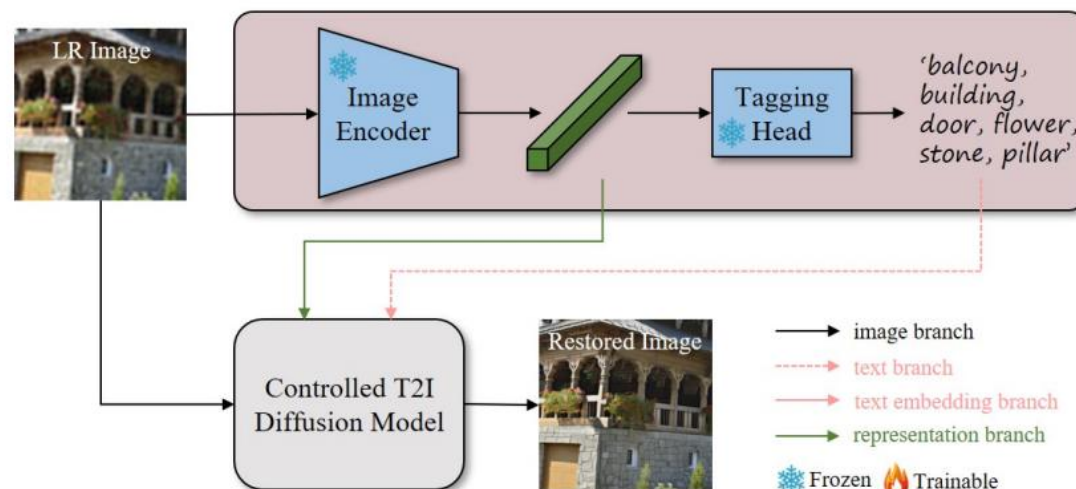


Background

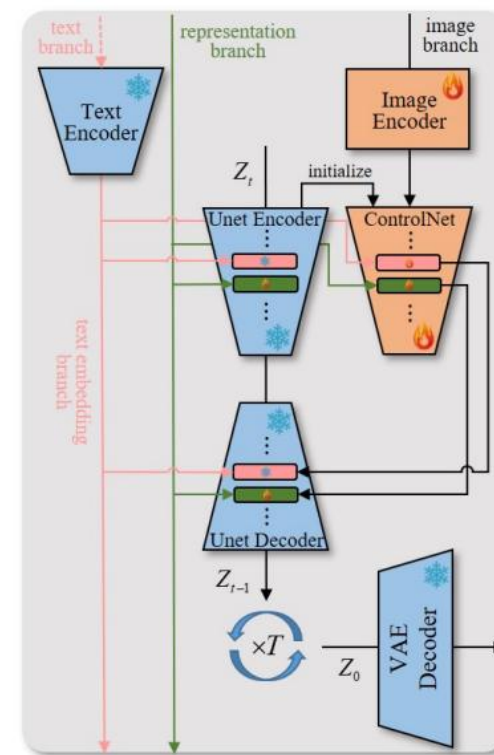
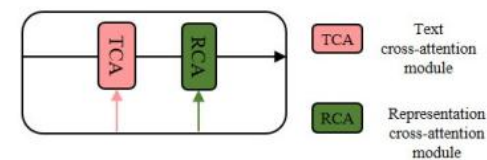
- SeeSR
 - CVPR 2024



(a) Degradation-aware prompt extractor



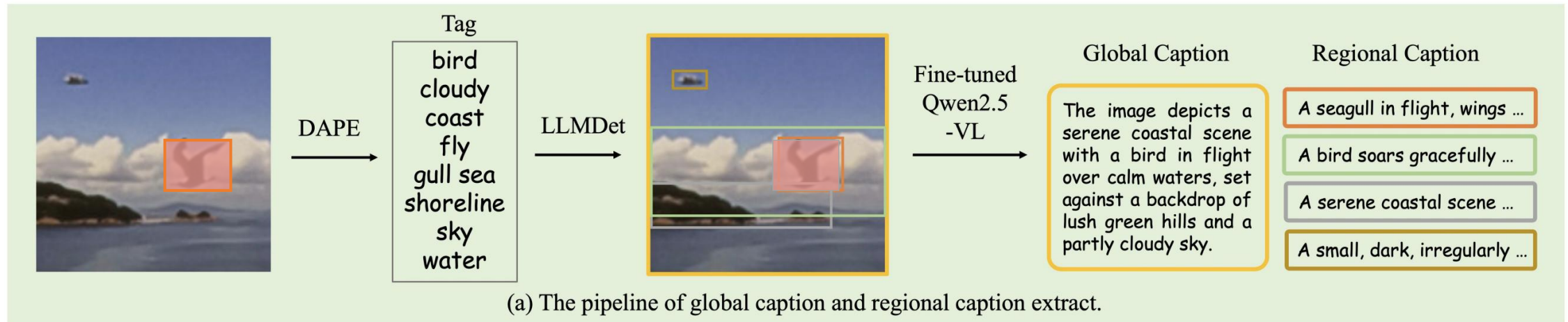
(b) Real-ISR with DAPE



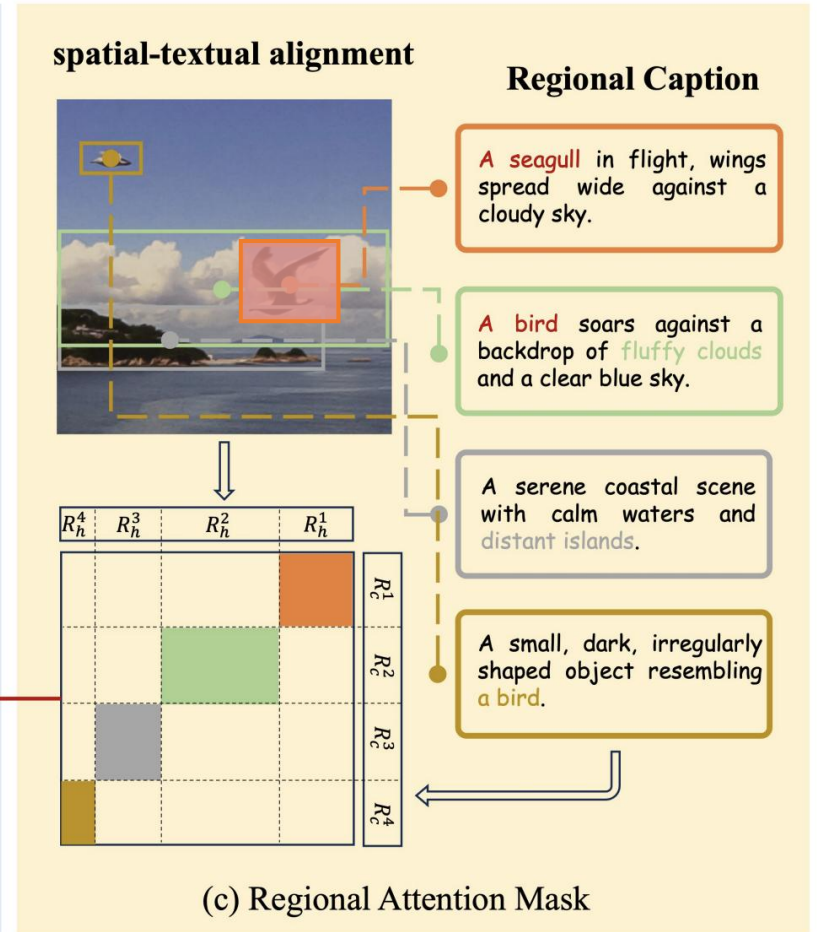
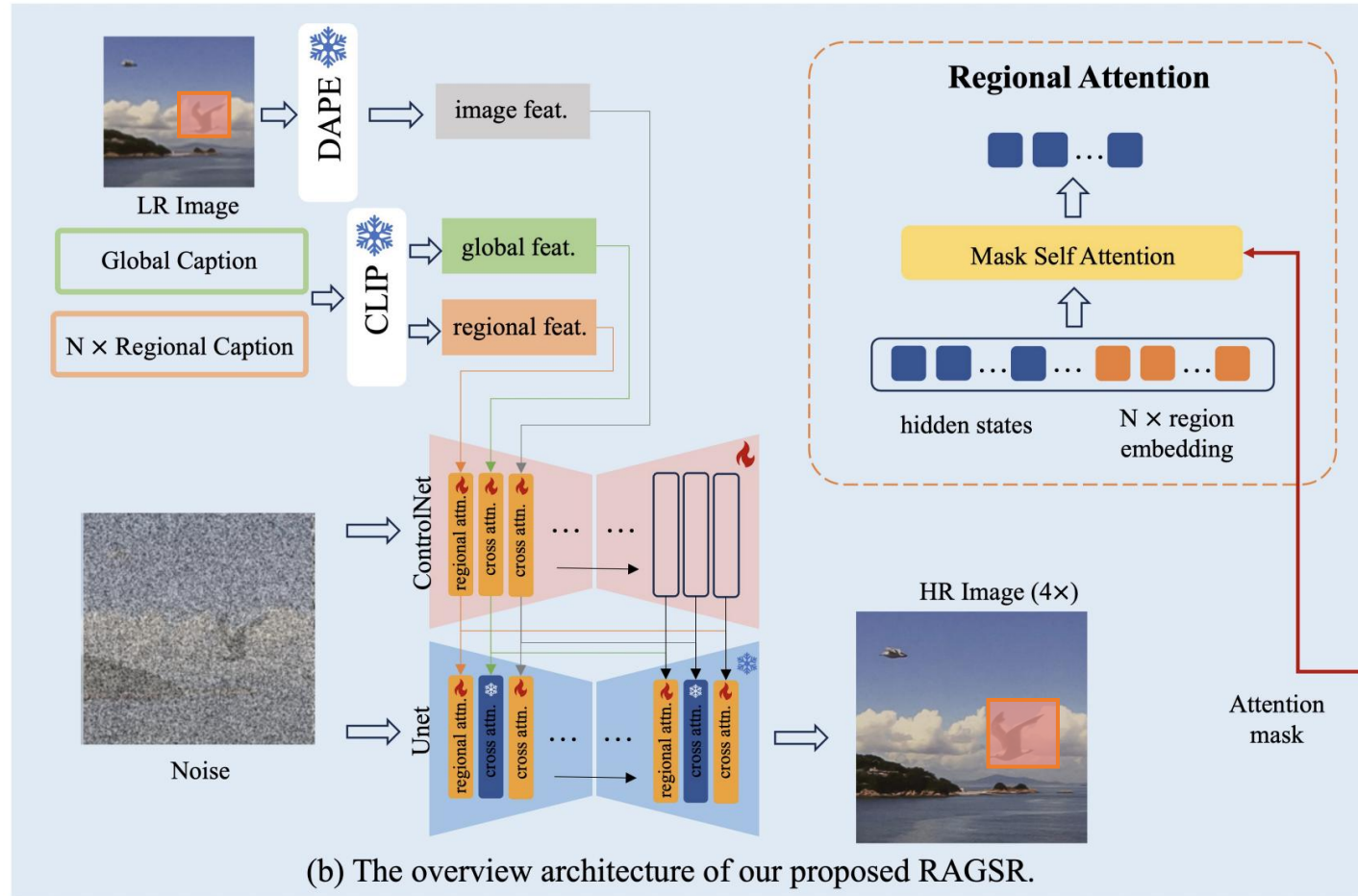
(c) Controlled T2I diffusion model

Background

- RAGSR
 - arXiv 2025

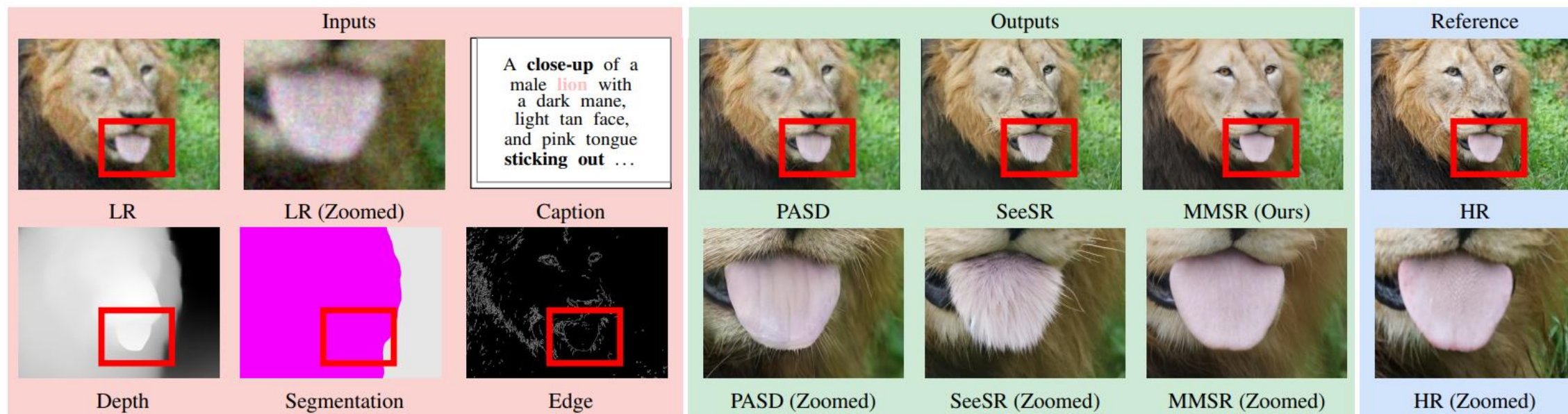


Background



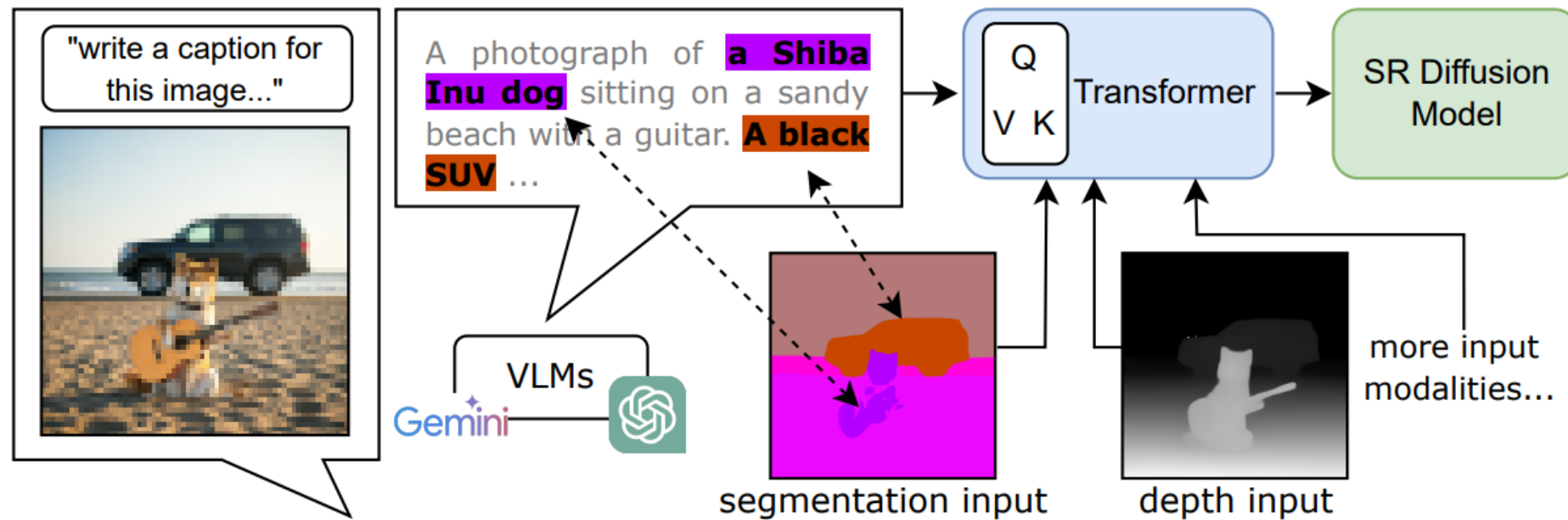
MMSR

- The Power of Context
 - CVPR 2025



MMSR

- Language models struggle to accurately represent spatial information



MMSR

- **Problem**

- **Spatial Ambiguity:** Text-based approaches struggle to accurately represent spatial relationships and details.
- **Hallucinations:** Generative models can introduce false details that weren't present in the original image.
- **Limited Controllability:** Users have minimal control over the super-resolution process.

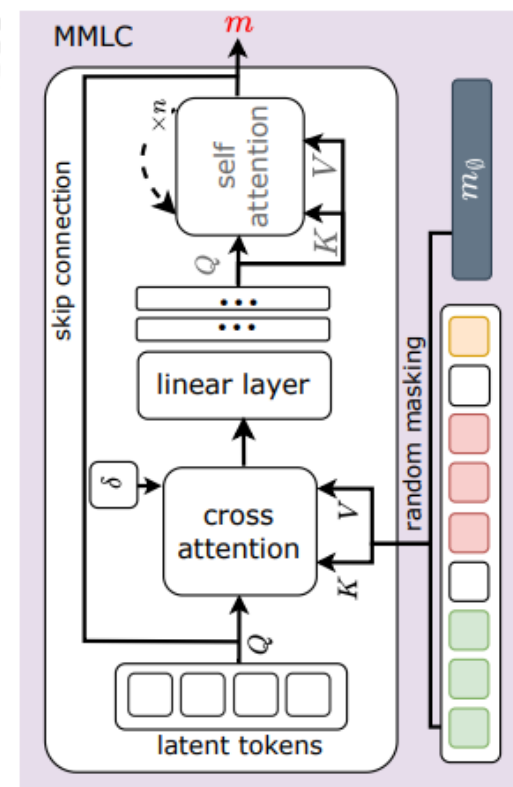
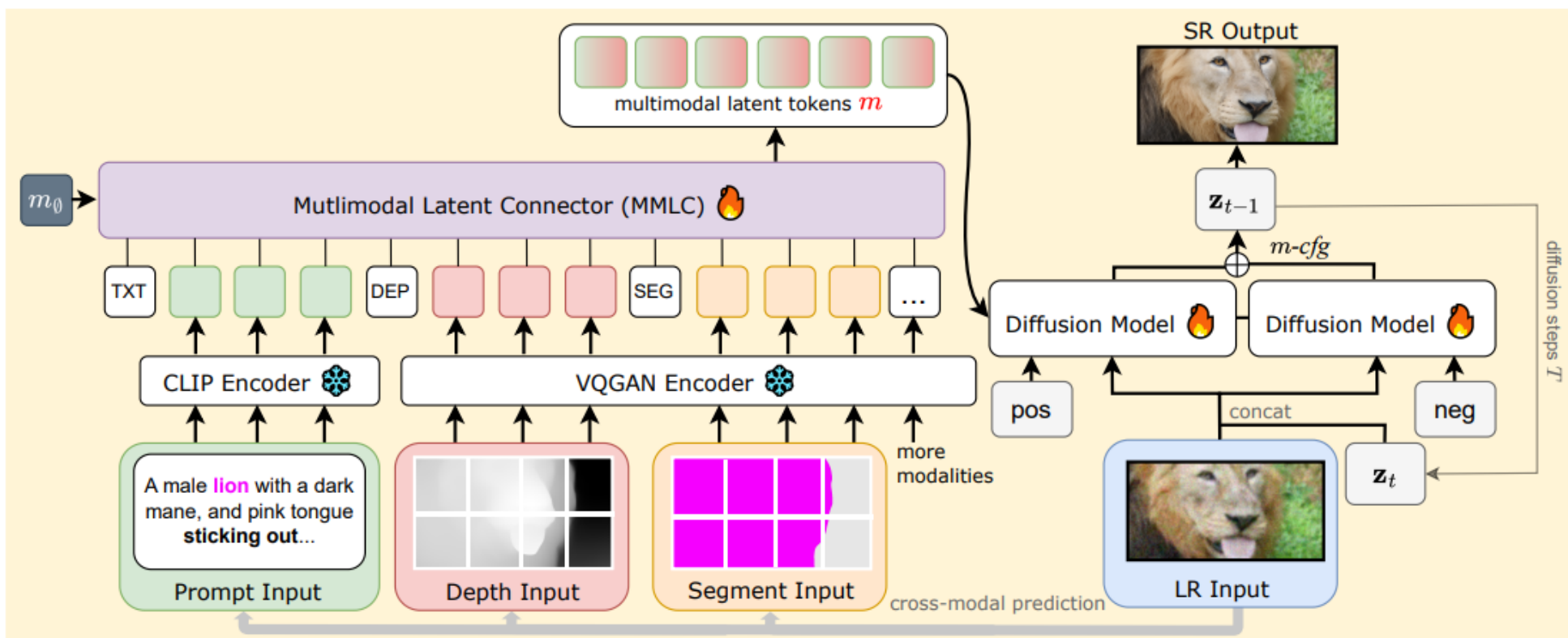
MMSR

- **Guidance types**

- **Text captions:** Provide semantic understanding of the image content
- **Depth maps:** Offer spatial information about object positioning and scene geometry
- **Semantic segmentation:** Supply object boundaries and class information
- **Edge maps:** Provide structural information about object contours

MMSR

- Framework



MMSR

- Discrete tokens for encoding each modality rather than continuous features

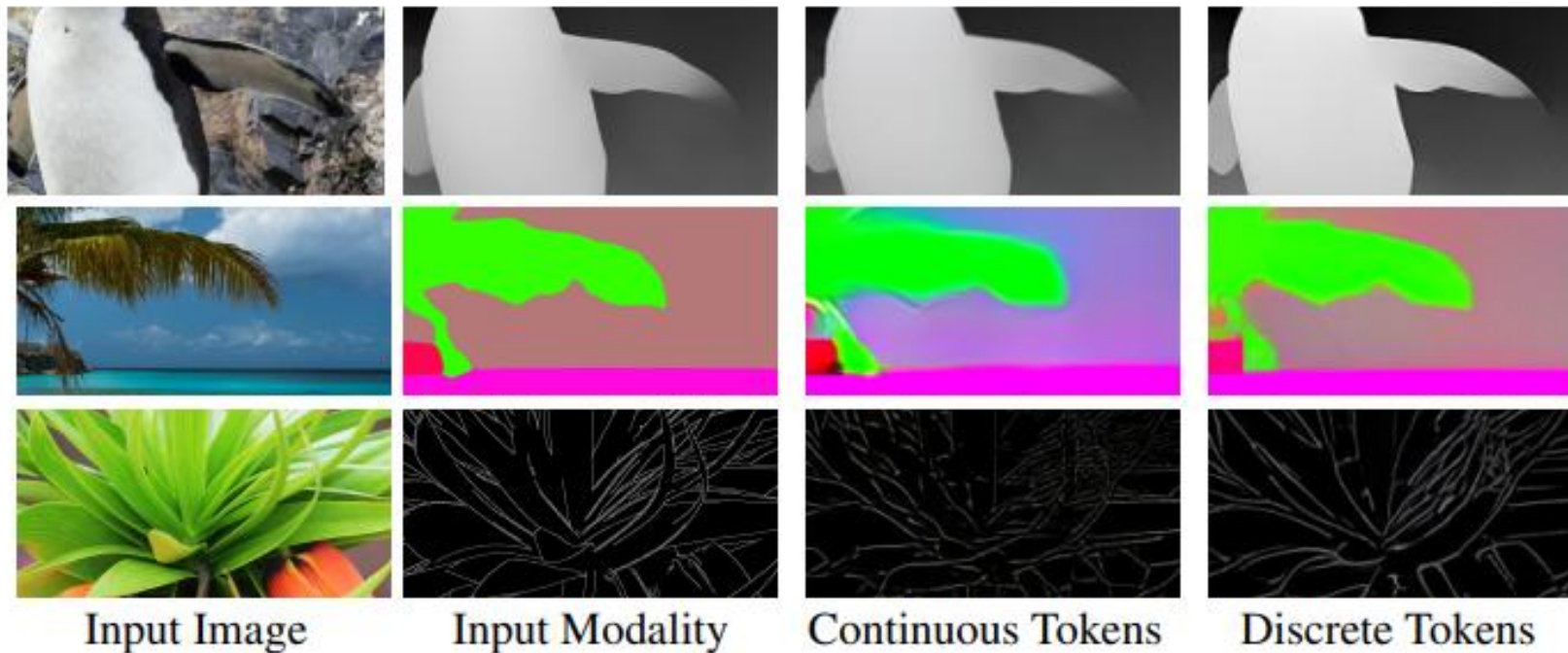


Figure 4. Using discrete multimodal tokens leads to superior reconstruction of modalities compared to continuous tokens.

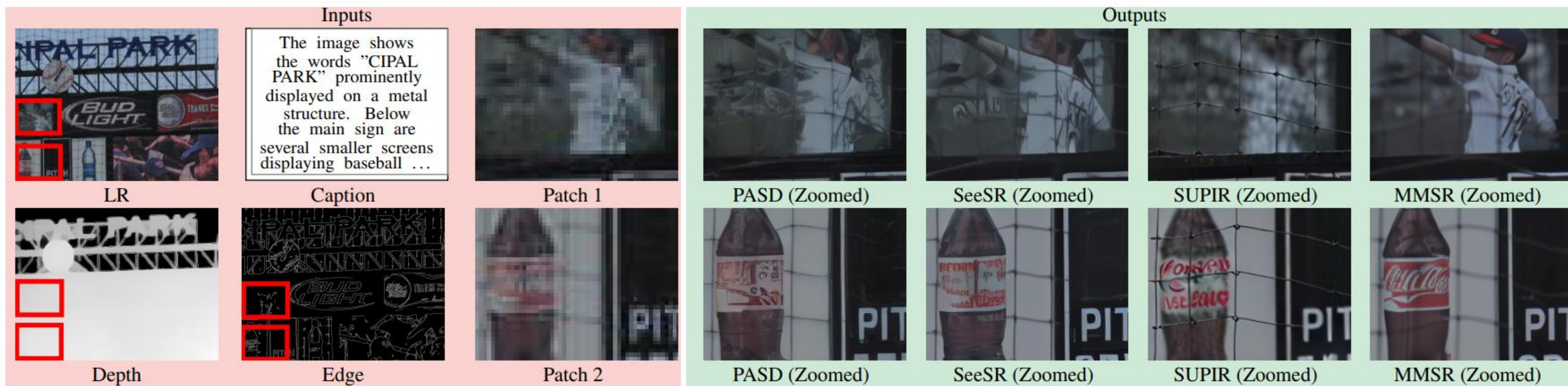
MMSR

- By scaling the temperature during sampling, achieve precise control over the super-resolution process

$$\text{Attention}(Q, K, V) = \text{softmax} \left(\frac{QK^T}{\delta} \right) V.$$

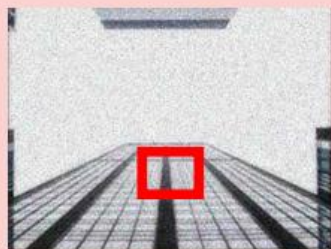
MMSR

- Comparison Result



MMSR

Inputs



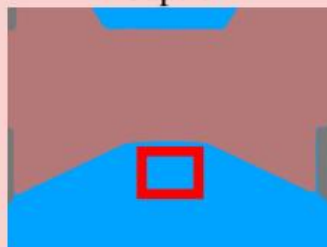
LR

Two skyscrapers converge towards the center, creating a symmetrical composition. Perspective is from ground level ...

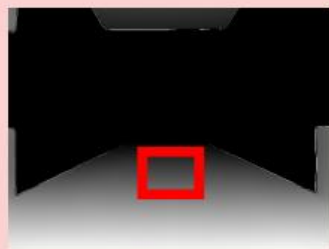
Caption



LR (Zoomed)



Segmentation



Depth

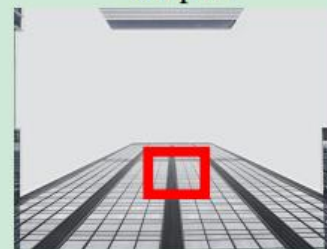


Edge

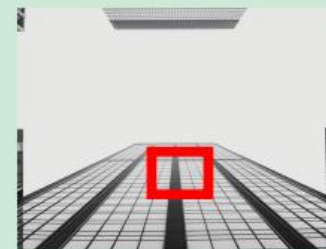
Outputs



PASD



SeeSR



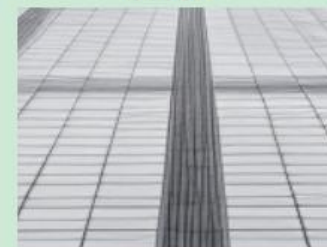
MMSR (Ours)



HR



PASD (Zoomed)



SeeSR (Zoomed)



MMSR (Zoomed)



HR (Zoomed)

A person wearing a blue hooded jacket and white pants carries a white surfboard down a grassy hill towards a dark sandy beach ...

Caption



LR



Depth



PASD



SeeSR



MMSR (Ours)



HR



MMSR

- Comparison Result

Methods	<i>DIV2K-Val-3k</i> 512×512							
	PSNR	SSIM	LPIPS ↓	DISTS ↓	NIQE ↓	FID ↓	MUSIQ	CLIPQA
BSRGAN	21.87	0.5539	0.4136	0.2737	4.7615	64.28	59.11	0.5183
R-ESRGAN	21.94	0.5736	0.3868	0.2601	4.9209	53.46	58.64	0.5424
LDL	21.52	0.5690	0.3995	0.2688	5.0249	58.94	57.90	0.5313
DASR	21.72	0.5536	0.4266	0.2688	4.8596	67.22	54.22	0.5241
FeMASR	20.85	0.5163	0.3973	0.2428	4.5726	53.70	58.10	0.5597
LDM	21.26	0.5239	0.4154	0.2500	6.4667	41.93	56.52	0.5695
StableSR	20.84	0.4887	0.4055	0.2542	4.6551	36.57	62.95	0.6486
ResShift	21.75	0.5422	0.4284	0.2606	6.9731	55.77	58.23	0.5948
PASD	20.77	0.4958	0.4410	0.2538	4.8328	40.77	66.85	0.6799
DiffBIR	20.94	0.4938	0.4270	0.2471	4.7211	40.42	65.23	0.6664
SeeSR	21.19	0.5386	0.3843	0.2257	4.9275	31.93	68.33	0.6946
MMSR	21.74	0.5693	0.3707	0.2071	4.2532	29.35	70.06	0.7164

Methods	<i>DIV2K-Val-100</i> 1024×1024							
	PSNR	SSIM	LPIPS ↓	DISTS ↓	NIQE ↓	MANIQA	MUSIQ	CLIPQA
R-ESRGAN	21.77	0.5813	0.3624	0.1990	3.6573	0.4046	47.54	0.5358
StableSR	20.07	0.3947	0.5097	0.2427	3.6260	0.4113	65.39	0.6938
PASD	21.27	0.5369	0.3473	0.1753	3.6321	0.4708	69.69	0.6914
SUPIR	20.65	0.5350	0.3849	0.1814	3.6458	0.4051	65.88	0.5697
SeeSR	21.31	0.5578	0.3273	0.1620	4.0215	0.5439	69.79	0.6941
MMSR	21.87	0.5565	0.2810	0.1492	3.4243	0.4885	72.31	0.7294

MMSR

- Comparison Result

Methods	<i>RealSR 512 × 512</i>							
	PSNR	SSIM	LPIPS ↓	LIQE	NIMA	MANIQA	MUSIQ	CLIPQA
R-ESRGAN	25.69	0.7616	0.2727	3.3574	4.6548	0.5487	60.18	0.4449
StableSR	24.70	0.7085	0.3018	3.6106	4.8150	0.6221	65.78	0.6178
PASD	24.29	0.6630	0.3435	3.5749	4.8554	0.6493	68.69	0.6590
SUPIR	22.97	0.6298	0.3750	3.5682	4.5757	0.5745	61.49	0.6434
SeeSR	<u>25.18</u>	<u>0.7216</u>	0.3009	<u>4.1360</u>	<u>4.9193</u>	<u>0.6442</u>	<u>69.77</u>	<u>0.6612</u>
MMSR	24.83	0.7003	<u>0.2952</u>	4.3468	5.1094	0.6578	71.33	0.6717

Methods	<i>DrealSR 512 × 512</i>							
	PSNR	SSIM	LPIPS ↓	LIQE	NIMA	MANIQA	MUSIQ	CLIPQA
R-ESRGAN	28.64	0.8053	0.2847	2.9255	4.3258	0.4907	54.18	0.4422
StableSR	26.71	0.7224	0.3284	3.2425	4.4861	0.5594	58.51	0.6357
PASD	27.00	0.7084	0.3931	3.5908	4.6618	0.5850	64.81	0.6773
SUPIR	24.61	0.6123	0.4294	3.4710	4.3815	0.5381	57.32	0.6758
SeeSR	26.75	0.7405	<u>0.3174</u>	<u>4.1270</u>	<u>4.6942</u>	<u>0.6052</u>	<u>65.09</u>	<u>0.6908</u>
MMSR	<u>27.28</u>	<u>0.7456</u>	0.3249	4.5023	5.0558	0.6301	68.93	0.6999

MMSR

- Ablation study

	MUSIQ	NIQE ↓	DISTS ↓	LPIPS ↓	Throughput
<i>w/o. MMLC</i>	<u>69.69</u>	<u>3.4845</u>	<u>0.1781</u>	<u>0.3929</u>	3.48 img/s
<i>w. MMLC</i>	72.31	3.4243	0.1492	0.2810	3.32 img/s

MMSR

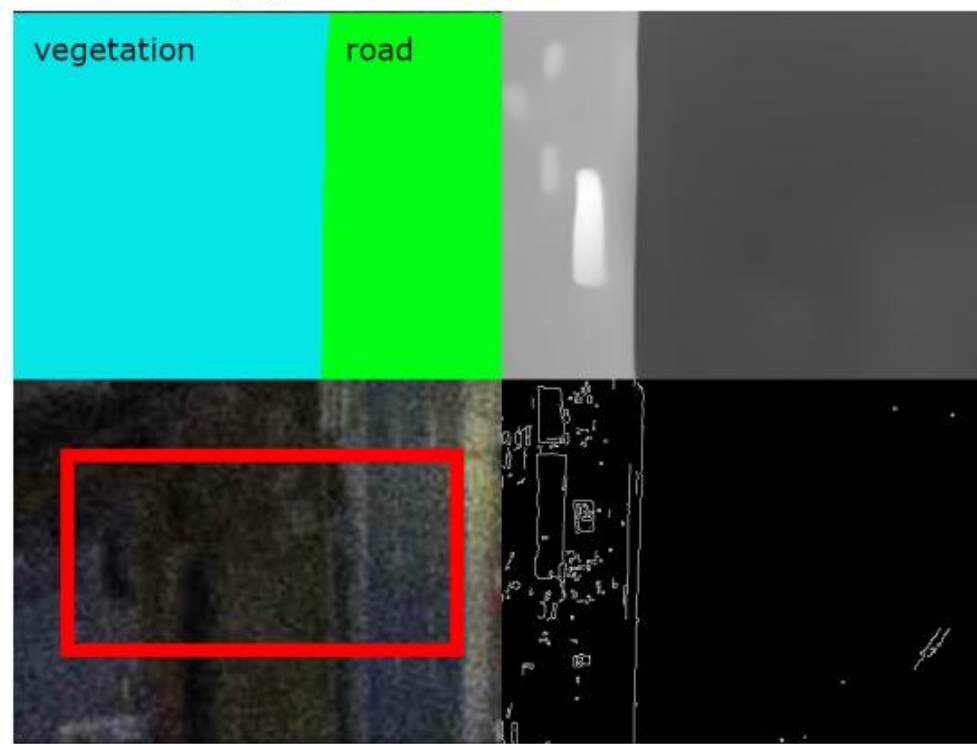
- Ablation study

guidance	2	10	14
cfg	0.3239	<u>0.4491</u>	<u>0.5064</u>
$m_{\emptyset}-cfg$	<u>0.2815</u>	0.4803	0.5493
$m-cfg$	0.2810	0.3471	0.3772

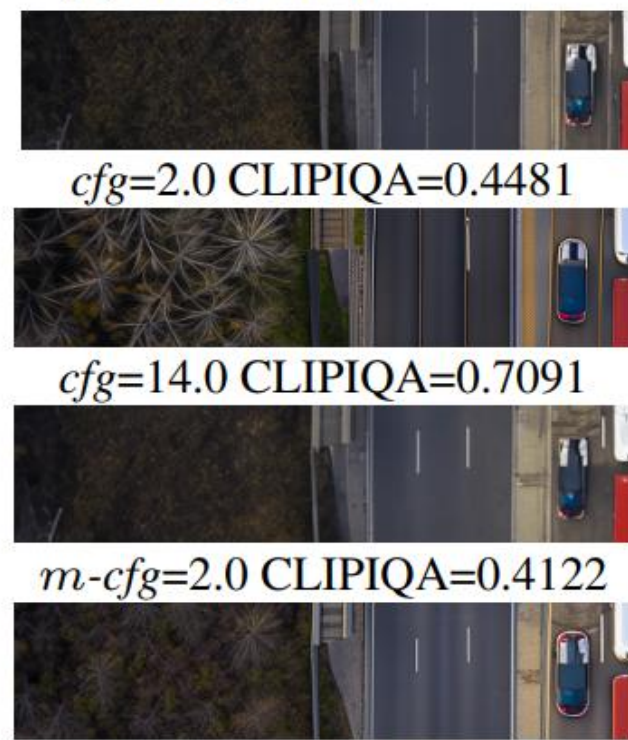
(a) LPIPS score

guidance	2	10	14
cfg	<u>3.577</u>	<u>4.6179</u>	<u>5.1886</u>
$m_{\emptyset}-cfg$	3.6261	5.1081	5.9175
$m-cfg$	3.4679	3.7419	3.9815

(b) NIQE score



Segment, Depth, LR, and Edge Input



$m-cfg=14.0$ CLIPQA=0.7352

(c) Visual comparisons

MMSR

- Ablation study



LR Input



Depth Input



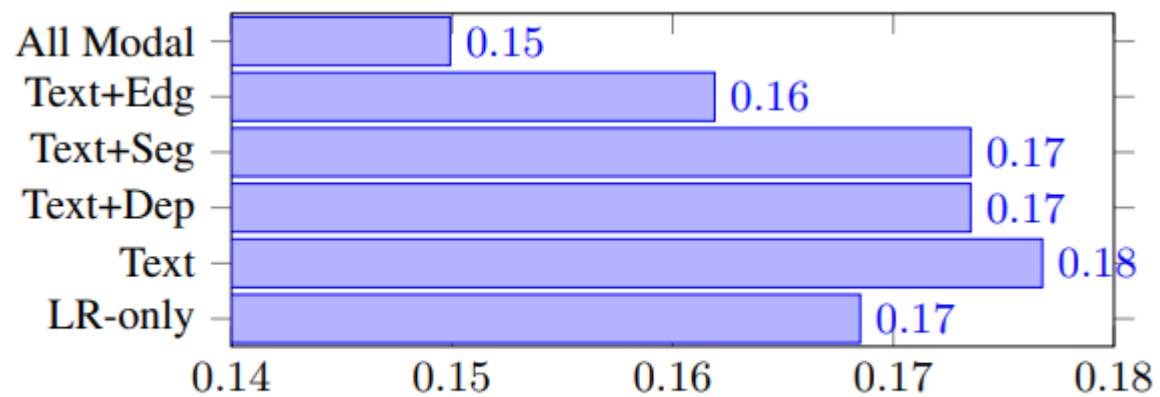
w/o. MMLC Result



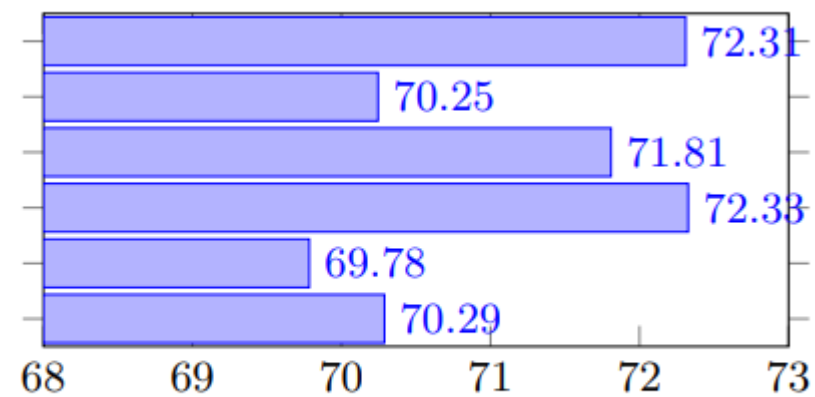
w. MMLC Result

MMSR

- Ablation study



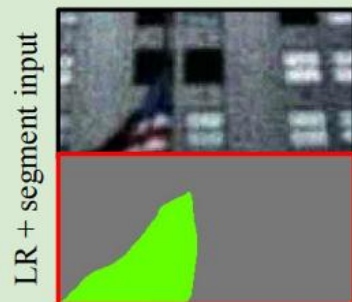
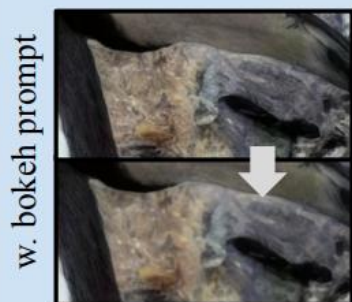
(a) DISTs score (← is better)



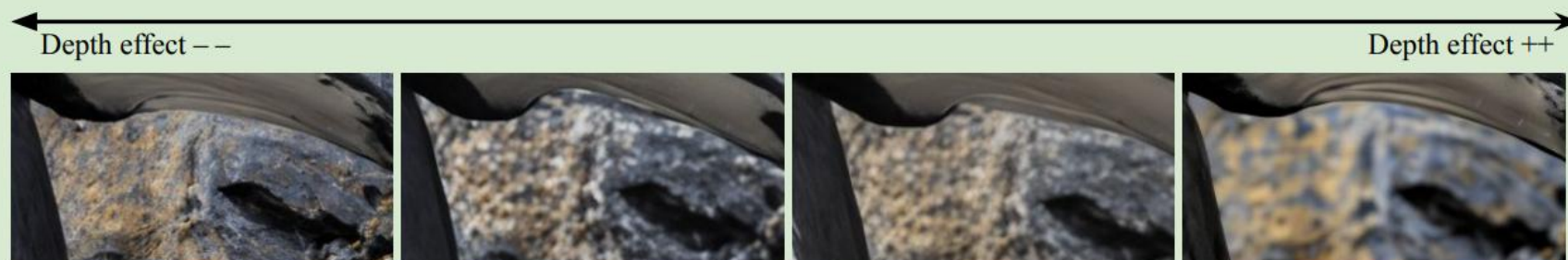
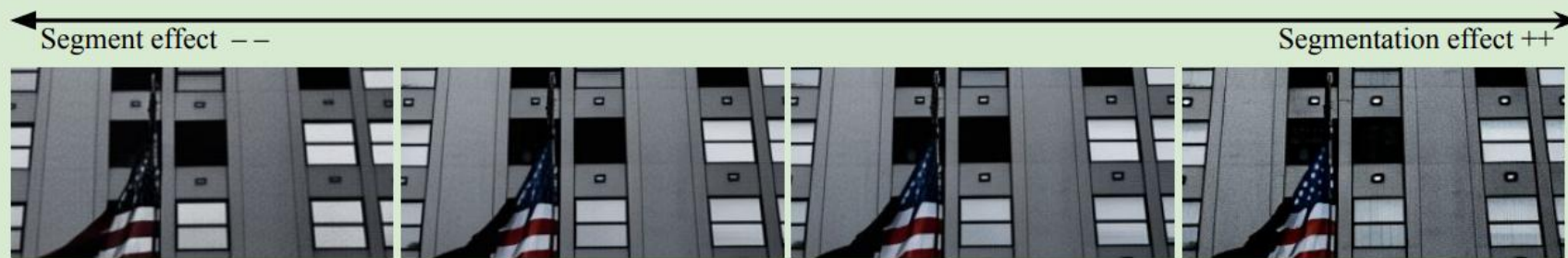
(b) MUSIQ score (→ is better)

MMSR

(a). Results of PASD



(b). Results of MMSR under different modality-specific controls



Conclusion

- A novel diffusion-based framework seamlessly integrates diverse modalities
- a learnable multimodal token and modality-specific contribution controls provide fine-grained control over the SR process

Thanks for your listening!