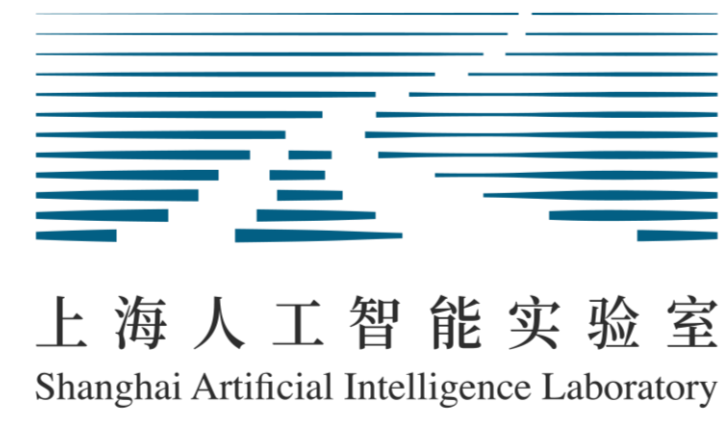


PalGAN: Image Colorization with Palette Generative Adversarial Networks



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Introduction

Key Challenges

- Multimodal ambiguity
- Color bleeding

Our method PalGAN

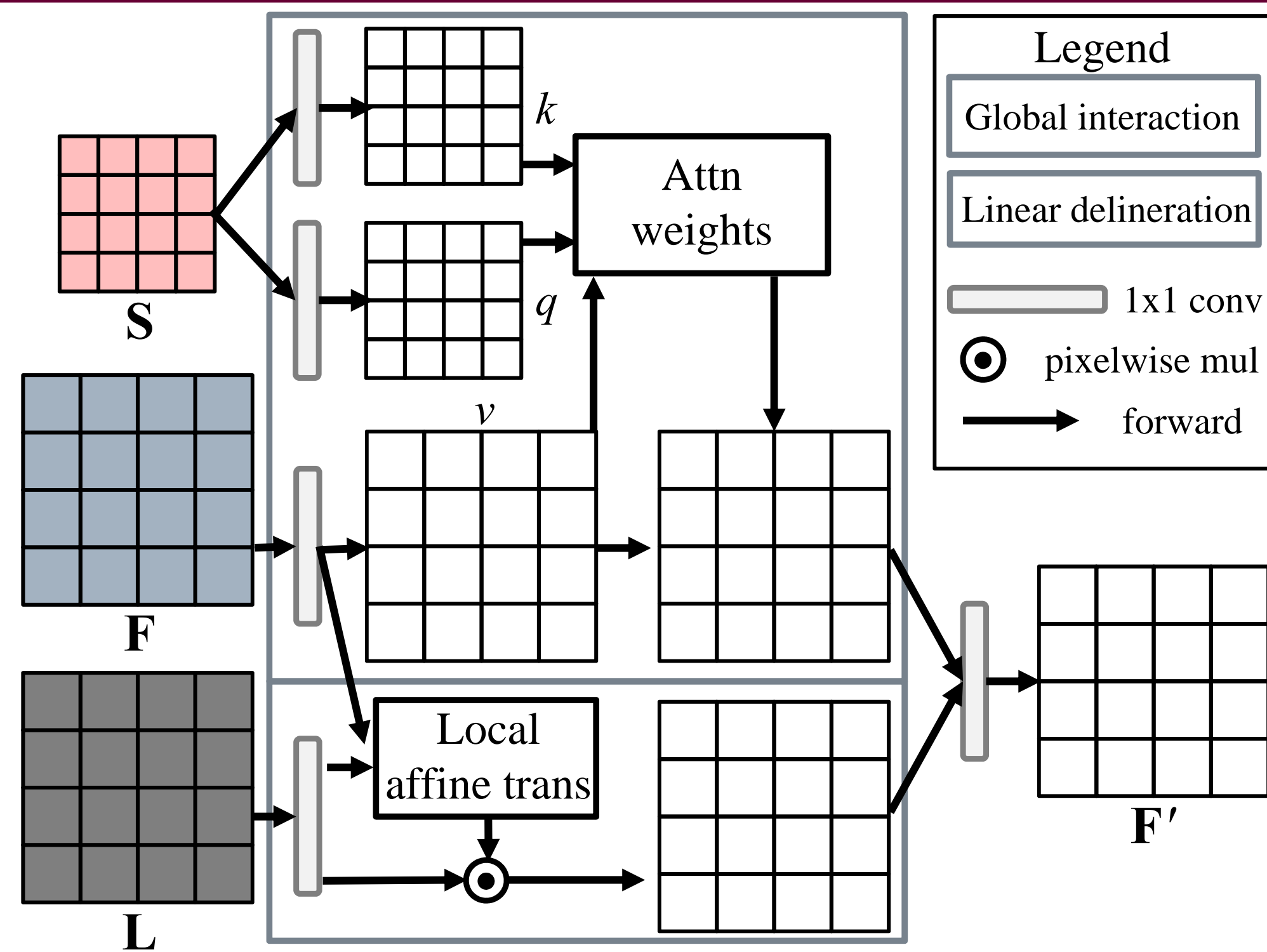
- Palette estimation
 - Predict a probabilistic palette from a gray image
 - Conduct color assignment based on the palette
- Chromatic attention (CA): study color affinities by considering semantic and intensity correlation.

The performance of PalGAN:

- It gets SOTA quantitative results in ImageNet / COCO.
- It gives diverse, contrastive, and edge-preserving appearances.



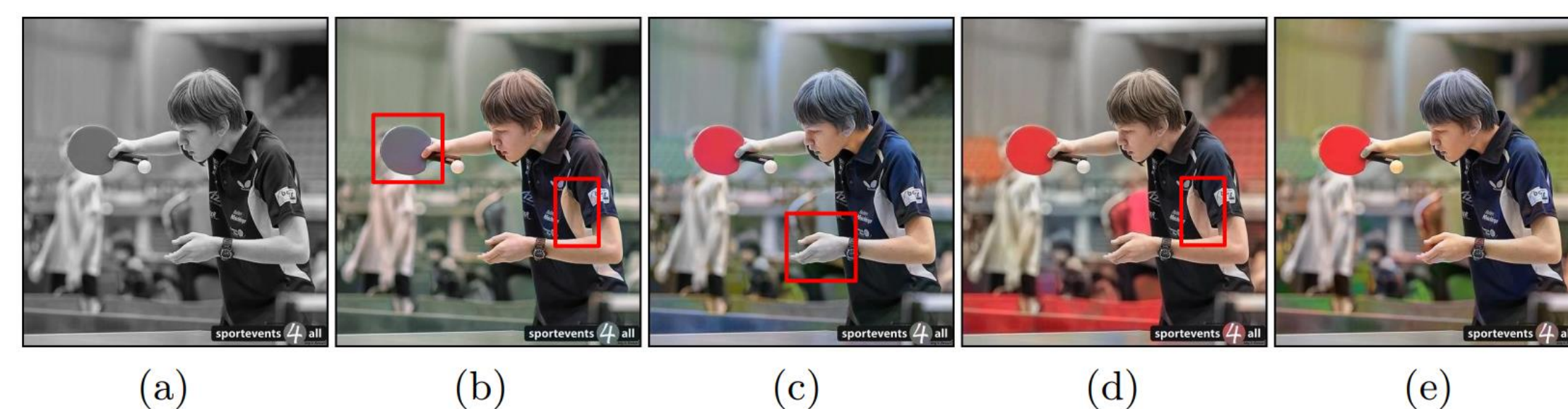
Framework



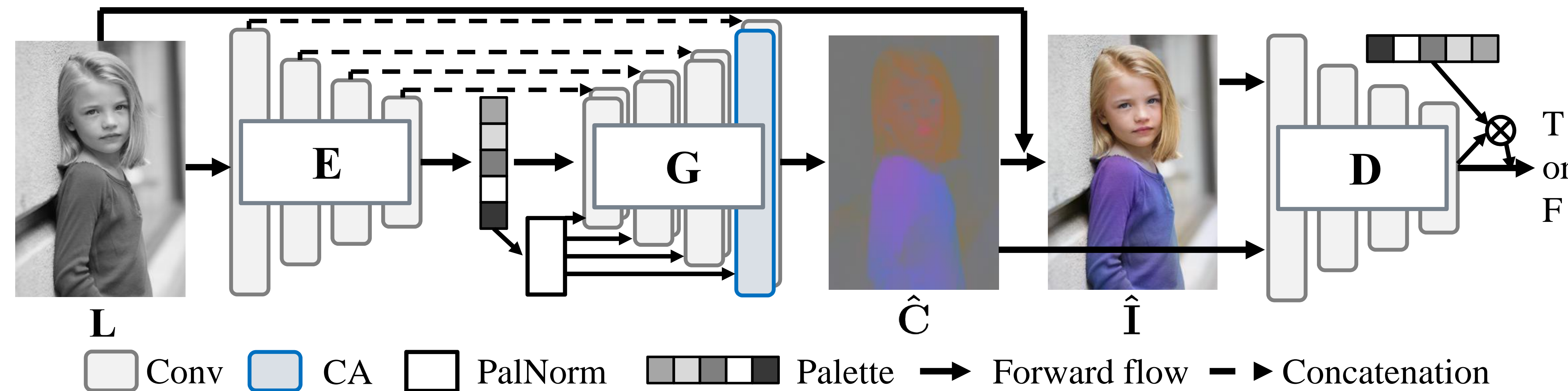
Motivation

- incorporates both semantic and low-level affinities into constructing color relations.

Components: global interaction & local delineation

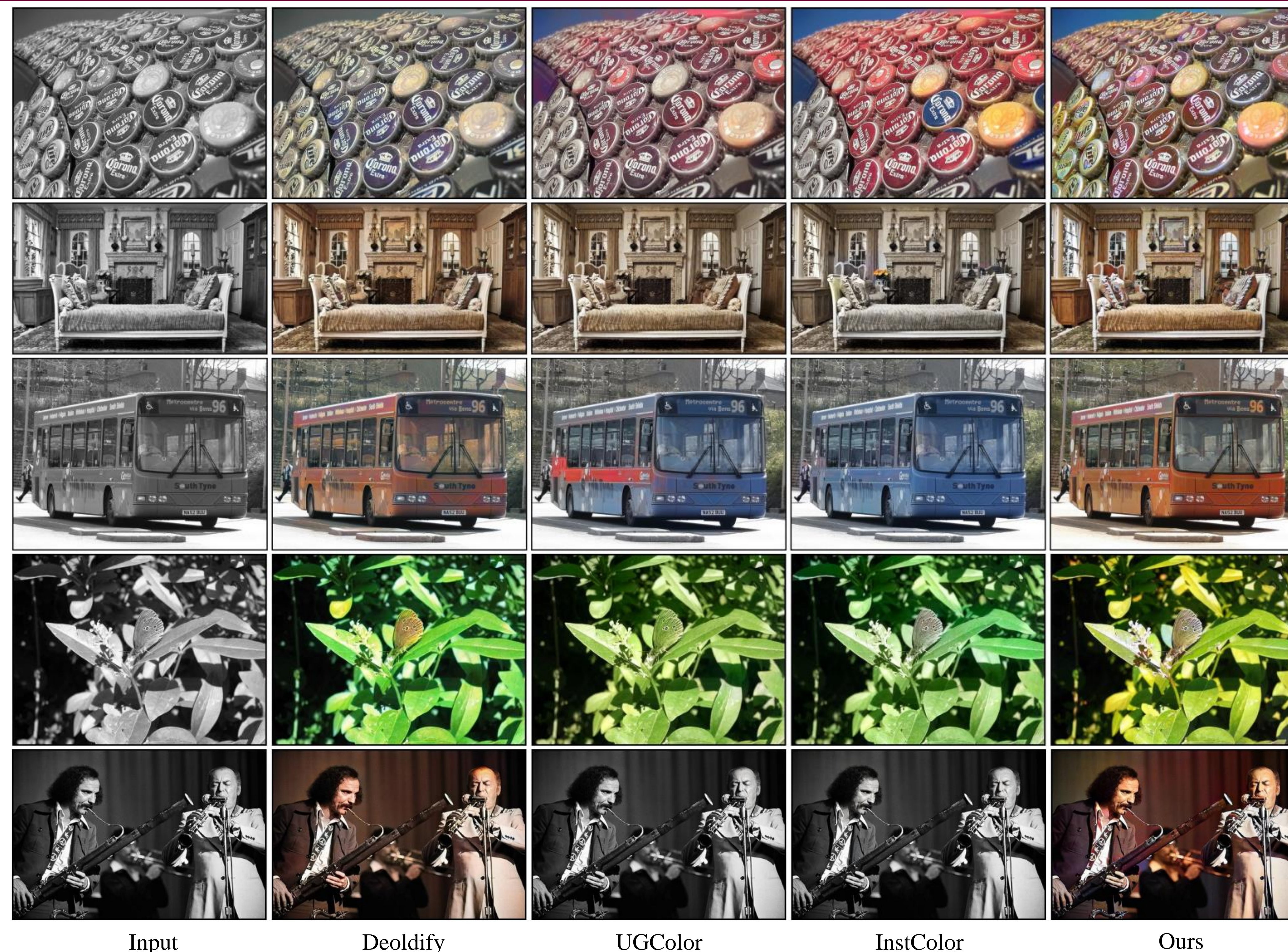


Framework



- Our PalGAN formulates colorization as a palette prediction and assignment problem.
- A palette generator E: estimates the global palette probabilities from the given gray image
- A palette assignment generator G: conducts color assignment task via conditional image generation.
 - Chromatic attention
- A color discriminator D: improves the result from adversarial training

Visual Quality

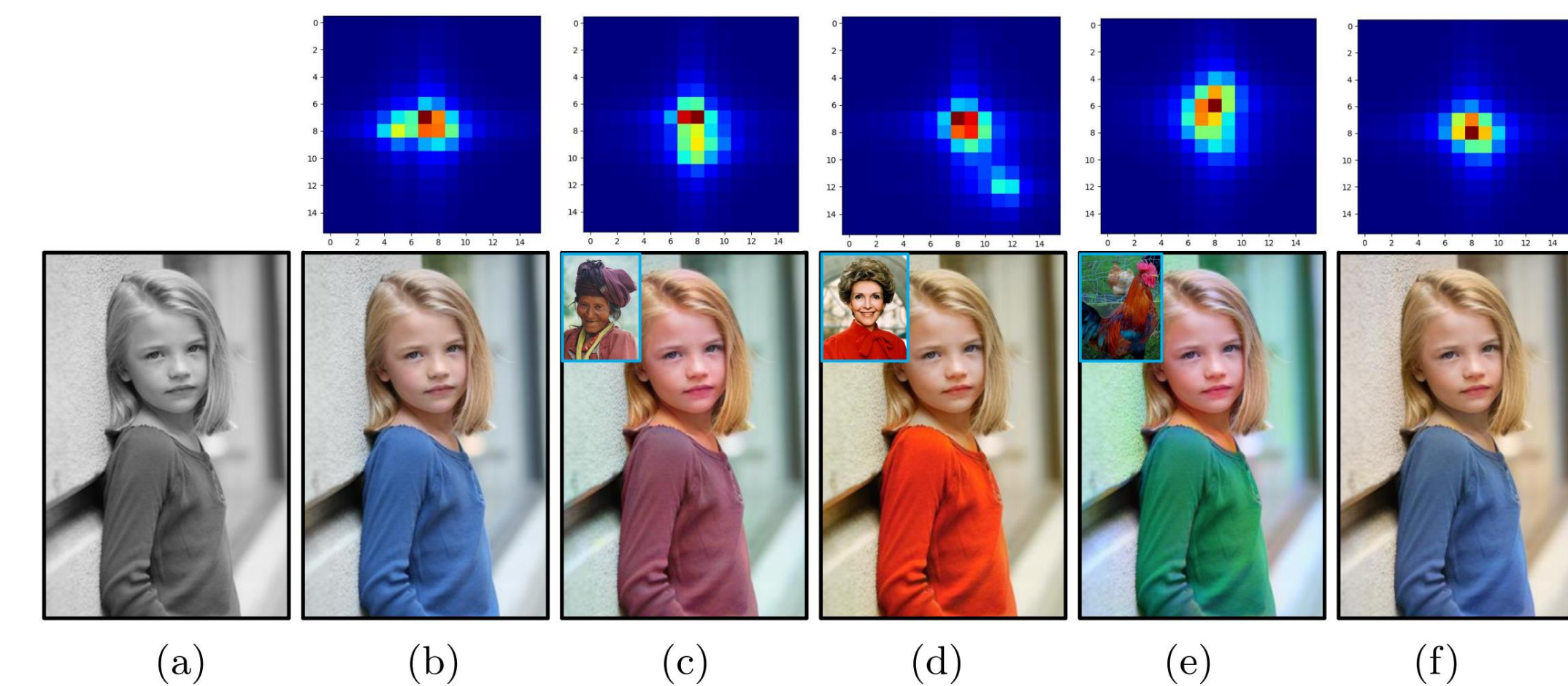


Experiment

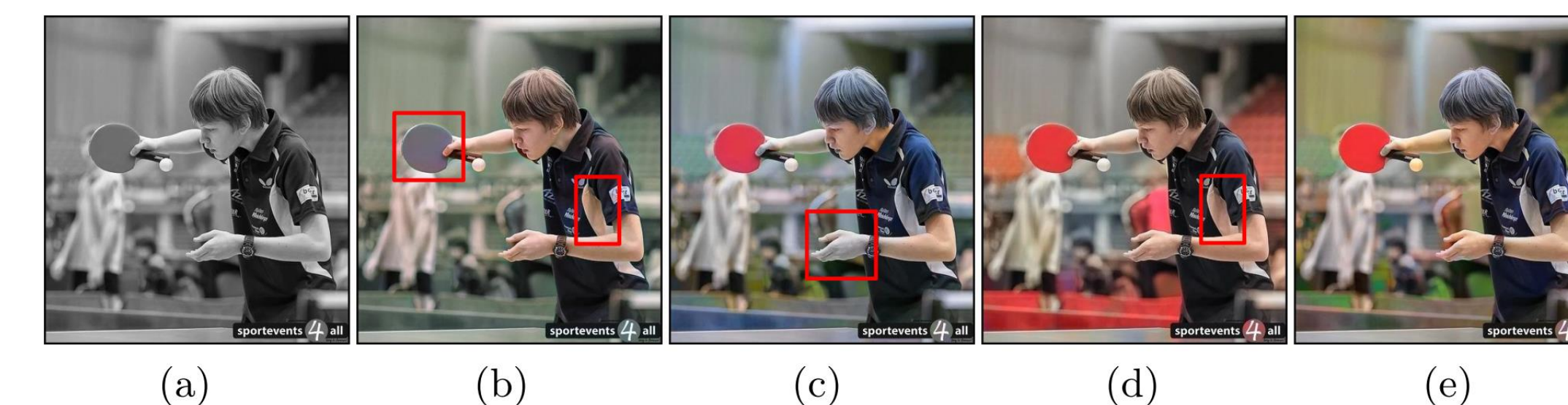
Quantitative results on the validation sets from different methods.

Method	ImageNet (cstest10k)				ImageNet (val50k)				COCO-Stuff			
	PSNR ↑	SSIM ↑	LPIPS ↓	FID ↓	PSNR ↑	SSIM ↑	LPIPS ↓	FID ↓	PSNR ↑	SSIM ↑	LPIPS ↓	FID ↓
CIColor [54]	22.30	0.902	0.221	12.20	22.26	0.902	0.221	9.39	21.84	0.895	0.234	22.32
UGColor [56]	24.26	0.918	0.174	7.49	24.26	0.919	0.173	4.60	24.34	0.924	0.165	14.74
Lei <i>et. al.</i> [26]	24.52	0.917	0.202	12.60	24.03	0.918	0.189	6.35	24.59	0.922	0.191	23.10
Deoldify [2]	23.54	0.914	0.187	5.78	22.97	0.911	0.185	3.87	23.98	0.939	0.161	12.75
ColTrans [23]	21.81	0.892	0.218	6.37	22.12	0.894	0.216	3.81	22.11	0.898	0.210	11.65
Ours ¹	24.19	0.917	0.161	4.60	24.25	0.917	0.161	2.78	24.56	0.924	0.148	7.70
Ours ²	24.66	0.920	0.170	5.24	24.54	0.920	0.168	3.62	24.72	0.944	0.156	8.93
InstColor* [38]	23.03	0.909	0.191	7.35	23.06	0.910	0.190	4.94	22.35	0.838	0.238	12.24
GPColor* [49]	21.66	0.871	0.230	5.46	21.81	0.880	0.230	3.62	N/A	N/A	N/A	N/A
Ours*	27.75	0.932	0.110	4.20	27.53	0.913	0.118	2.42	28.28	0.936	0.105	7.21

Visualizations of palettes (1st row, shown in jet colormap) and how they work on colorization (2nd row). (a) Input, (b) the ground truth, (c)-(e) reference-based colorization, (f) automatic colorization



Ablation studies of chromatic attention (CA). (a) input, (b) wo CA, (c) w Global, (d) w Local, (e) full CA. Please zoom in.



User study. Each entry gives the percentage of cases where colorization results are favored compared with GT.

Method	Ours	Coltrans	GPCol	InstCol	Deoldify	UGCol
Rate	47.20%	41.50%	39.25%	37.50%	41.13%	42.50%

Quantitative results on COCO-Stuff using different structures.

Structure	PSNR ↑	SSIM ↑	LPIPS ↓	FID ↓
AE	25.89	0.928	0.146	14.15
VAE	23.21	0.905	0.179	11.76
UGC w CA	24.52	0.923	0.162	11.38
PalGAN w rand ref	20.88	0.883	0.240	9.64
PalGAN w SA	22.68	0.892	0.175	9.02
PalGAN w PatchD [19,41]	23.07	0.895	0.183	8.44
PalGAN w BN	22.36	0.895	0.209	9.97
PalGAN w SPADE [35]	24.06	0.916	0.167	7.90
PalGAN wo $E(\hat{h})$	24.58	0.924	0.149	8.17
PalGAN	24.56	0.924	0.148	7.70

Quantitative results on COCO-Stuff by ablating chromatic attention.

G L	PSNR ↑	SSIM ↑	LPIPS ↓	FID ↓
× ×	21.93	0.902	0.203	9.90
× ✓	24.52	0.924	0.146	9.97
✓ ×	23.32	0.907	0.174	8.34
✓ ✓	24.56	0.924	0.148	7.70