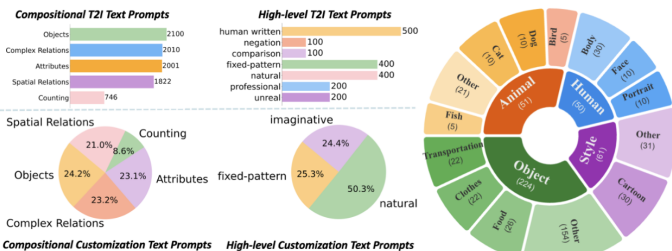


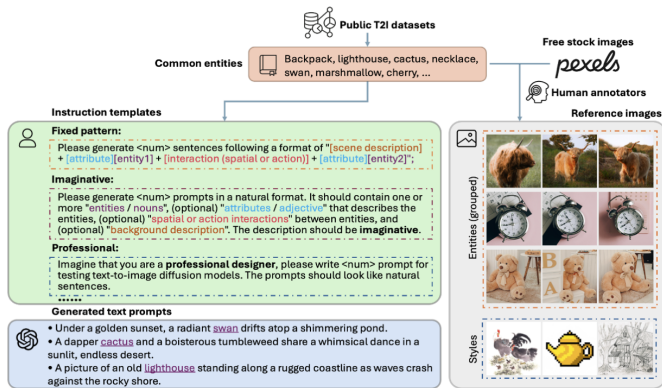
Contribution

Unified task coverage and multi-modal input. We collect over 380 groups (animal, object, human, and style) comprising 1,750 multiview object-centric images enabling rigorous reference-based generation. We also construct 4,850 richly annotated prompts across compositionality (attributes, relations, objects, and numeracy), style (fixed pattern, professional, natural, human-written), realism (imaginative) and common sense (comparisons, negations).

Three-level evaluation suite. We propose a multilevel scoring framework for comprehensive evaluation. (1) Low-level metrics assess visual artifacts and identity preservation of objects; (2) At mid-level, we propose the **Aspect Matching Score (AMS)** : a novel VQA-based metric that captures fine-grained semantic alignment, showing strong correlation with human perception; (3) high-level metrics measure aesthetics and human preferences.



Data Curation



Overview of MMIG-Bench



Performance on Text-to-Image Generation Tasks

Method	Low Level		Mid Level		High Level		
	CLIP-T ↑	PALAVST ↓	AMS ↑	Human ↑	Aesthetic ↑	HPSv2 ↑	PickScore ↑
Diffusion Models							
SDXL [41]	33.529	14.340	79.08	72.29	6.337	0.277	0.120
Photon-v1 [40]	33.296	2.947	77.12	69.49	6.391	0.284	0.088
Lumina-2 [42]	33.281	15.531	84.11	73.18	6.048	0.287	0.116
HunyuanDiT-v1.2 [31]	33.701	8.024	83.61	74.89	6.379	0.300	0.144
Pixart-Sigma-xl2 [2]	33.682	9.283	83.18	76.65	6.409	0.304	0.165
Flux.1-dev [25]	33.017	2.171	84.44	76.44	6.433	0.307	0.210
SD 3.5-large [6]	33.873	6.359	85.33	77.04	6.318	0.294	0.157
HiDream-II-Full [50]	33.876	1.522	89.65	83.18	6.457	0.321	0.450
Autoregressive Models							
JanusFlow [33]	31.498	365.663	70.25	75.69	5.221	0.209	0.031
Janus-Pro-7B [3]	33.358	31.954	85.35	80.36	6.038	0.275	0.129
API-based Models							
Gemini-2.0-Flash [11]	32.433	11.053	85.35	81.98	6.102	0.275	0.110
GPT-4o [35]	32.380	3.497	82.57	81.02	6.719	0.279	0.263



Visit our GitHub repo for code and data
<https://github.com/hanghuacs/MMIG-Bench>



Performance on Customization Tasks

Method	Low Level			Mid Level			High Level		
	CLIP-T ↑	CLIP-I ↑	DINOv2 ↑	CUTE ↑	PALAVST ↓	BLIPQA ↑	AMS ↑	Aesthetic ↑	HPSv2 ↑
Diffusion Models									
BLIP Diffusion [29]	26.137	80.286	26.232	69.681	56.780	0.247	41.59	5.830	0.213
DreamBooth [35]	24.227	88.758	38.961	79.780	43.535	0.108	28.00	5.368	0.179
Emu2 [48]	28.410	79.026	31.831	71.132	10.461	0.378	53.13	5.639	0.243
IP-Adapter-XL [40]	28.577	85.387	34.177	74.995	8.531	0.290	51.10	5.840	0.233
MS Diffusion [34]	31.446	77.827	23.600	71.306	4.748	0.496	71.40	5.979	0.271
API-based Models									
GPT-4o [34]	33.527	75.152	25.174	64.776	1.973	0.672	98.90	6.368	0.289

Aspect Matching Score (AMS)

Method	BLIPVQA ↑	VQ2 ↑	DSG ↑	AMS ↑	Human ↑
Diffusion Models					
SDXL	0.433	69.07	87.63	79.08	72.29
Photon-v1	0.440	66.84	86.26	77.12	69.49
Lumina-2	0.517	72.51	90.12	84.11	73.18
HunyuanDiT-v1.2	0.513	73.13	89.77	83.61	74.89
Pixart-Sigma-xl2	0.521	71.51	89.69	83.18	76.65
Flux.1-dev	0.511	71.41	83.33	84.44	76.44
SD 3.5-large	0.525	73.28	91.41	85.33	77.04
HiDream-II-Full	0.572	75.09	92.43	89.65	83.18
Autoregressive Models					
JanusFlow [33]	0.390	57.24	85.43	70.25	75.69
Janus-Pro [3]	0.530	67.41	92.15	85.35	80.36
API-based Models					
Gemini-2.0-Flash	0.495	72.01	92.93	85.40	81.98
GPT-4o	0.497	70.34	89.99	82.57	81.02

Case Study

