

AnyDressing: Customizable Multi-Garment Virtual Dressing via Latent Diffusion Models

Xinghui Li, Qichao Sun, Pengze Zhang, Fulong Ye, Zhichao Liao, Wanquan Feng, Songtao Zhao, Qian He



Challenge

Multi-Garment Virtual Dressing:

- **Garment Fidelity:** Existing methods encounter confusion while preserving the intricate textures of each



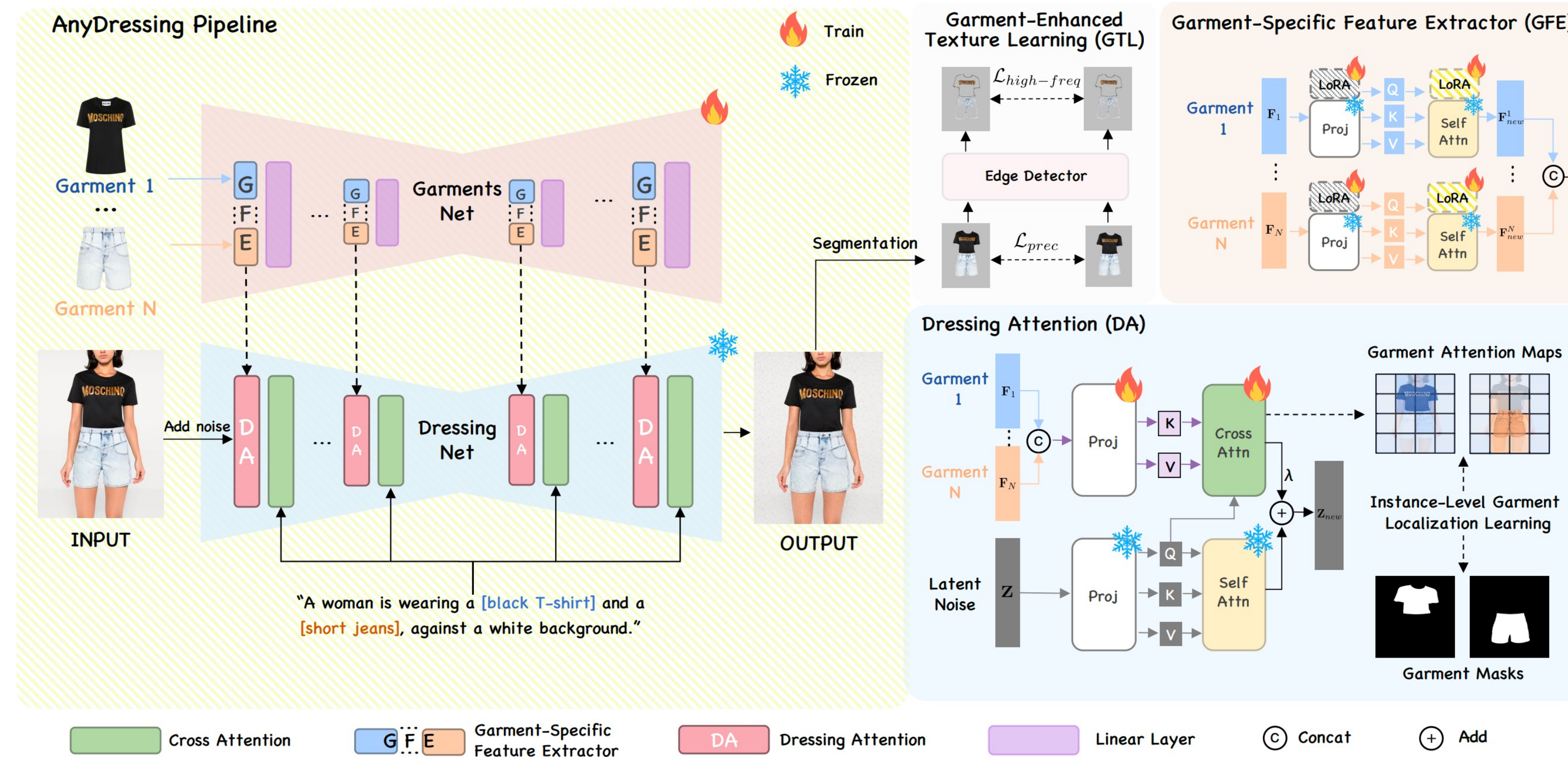
- **Text-Image Consistency:** Ensure the faithfulness to the prompts



- **Plugin Compatibility**



Method

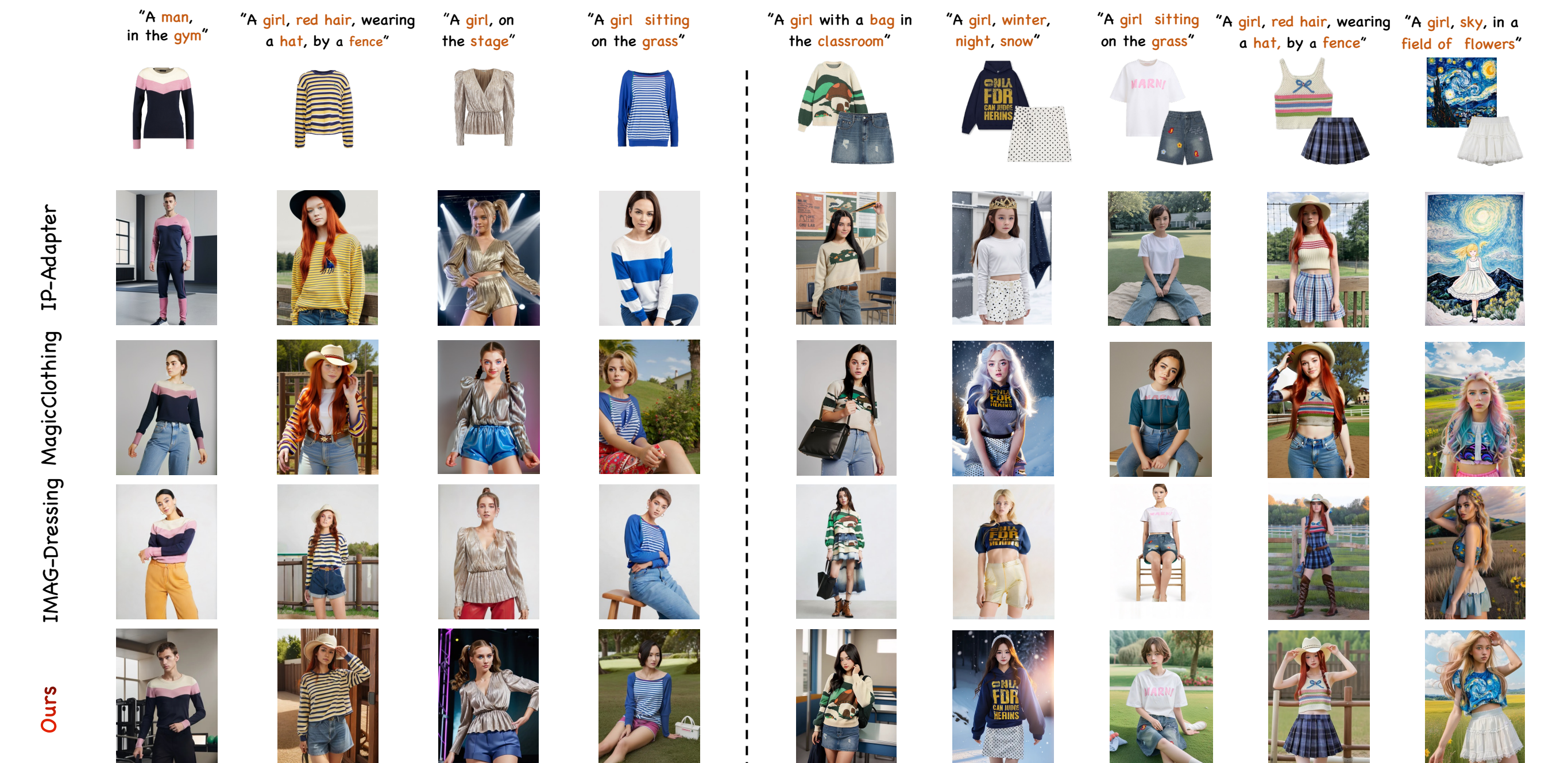


- **GarmentsNet:** Efficiently capture multi-garment textures in parallel by employing **Garment-Specific Feature Extractor**, which utilizes **parallelized self-attention Lora** layers within a shared U-Net architecture to individually encode garment textures
- **DressingNet:** Incorporating a **Dressing Attention (DA)** mechanism to seamlessly integrate multi-garment features into the denoising process; introducing a novel **Instance Level Garment Localization (IGL)** learning strategy to avoid influencing other irrelevant regions in the synthetic image
- **Garment-Enhanced Texture Learning:** Imposing constraints from **perceptual features** and **high-frequency information** to effectively enhance the fine-grained texture details in synthetic images
- **SOTA Performance:** Seamlessly integrate with any community control plugins and both quantitative and qualitative results demonstrate the superiority of our AnyDressing

Results

- **AnyDressing outperforms baselines in all metrics**

Method	Single Garment				Multiple Graments			
	VITON-HD [5]				Proprietary Dataset			
	CLIP-T	CLIP-I	CLIP-AS	DINO	CLIP-T	CLIP-I	CLIP-AS	DINO
IP-Adapter [54]	0.268	0.644	5.674	0.500	0.272	0.632	5.678	0.460
StableGarment [48]	0.285	0.583	5.781	0.522	0.281	0.587	5.648	0.510
MagicClothing [4]	0.288	0.640	5.703	0.363	0.298	0.619	5.784	0.340
IMAGDressing [42]	0.202	0.734	5.077	0.553	0.230	0.684	5.133	0.453
Ours	0.289	0.741	5.881	0.571	0.296	0.710	5.931	0.559



- **The effectiveness of each proposed module**

