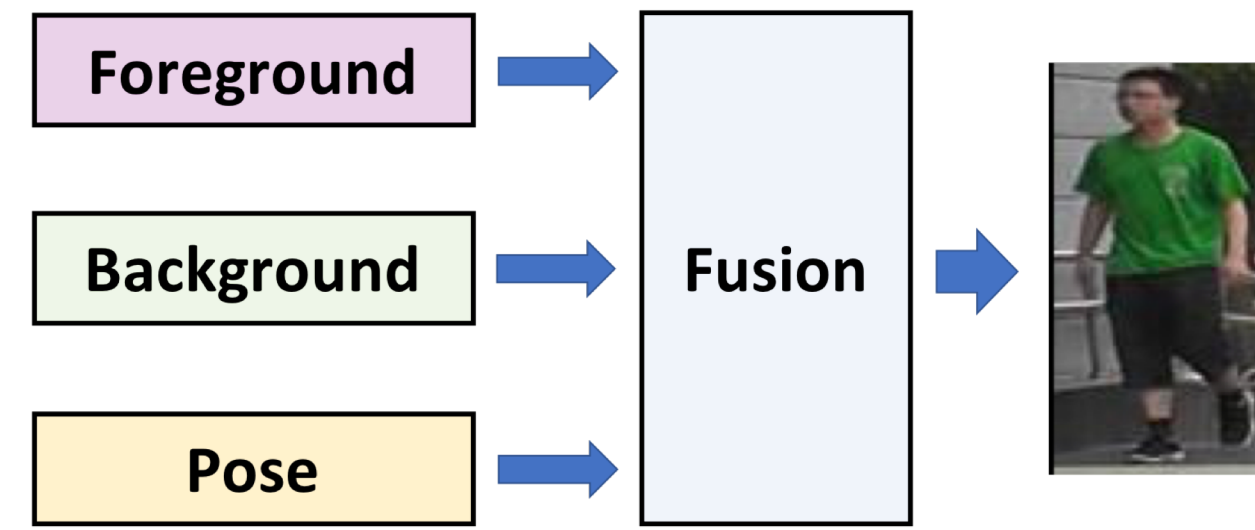


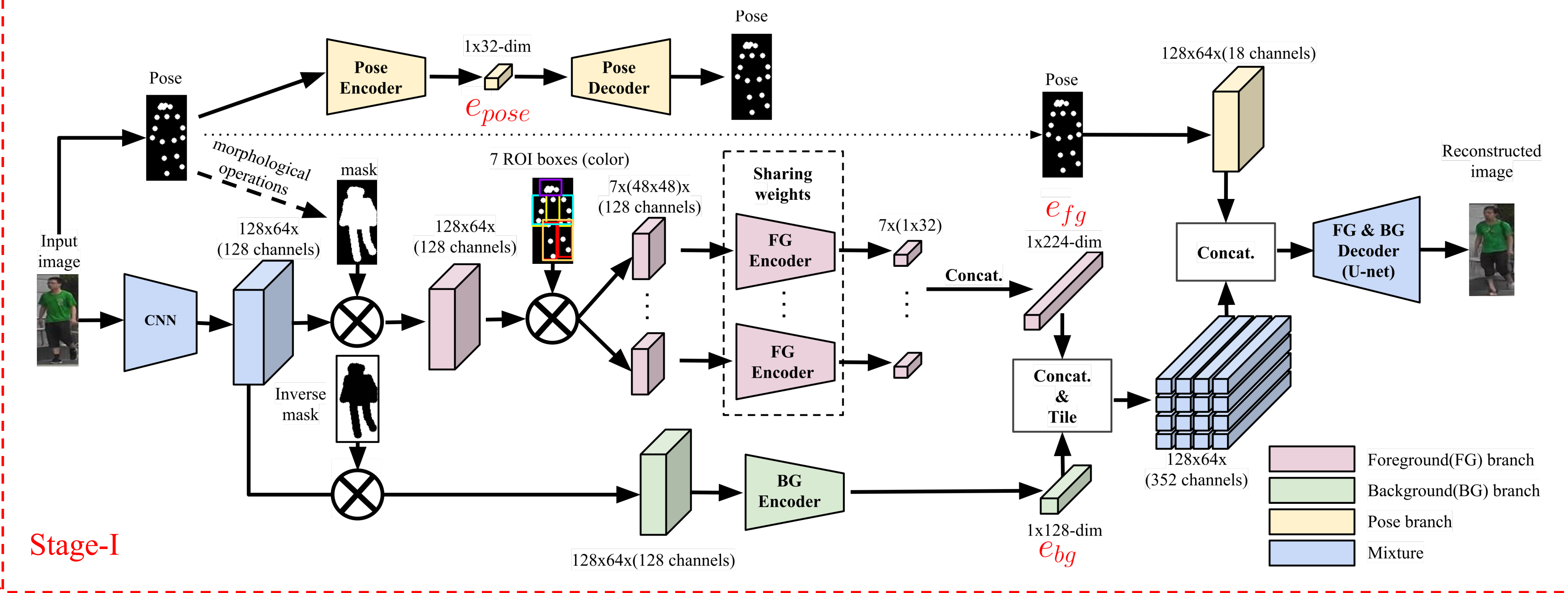
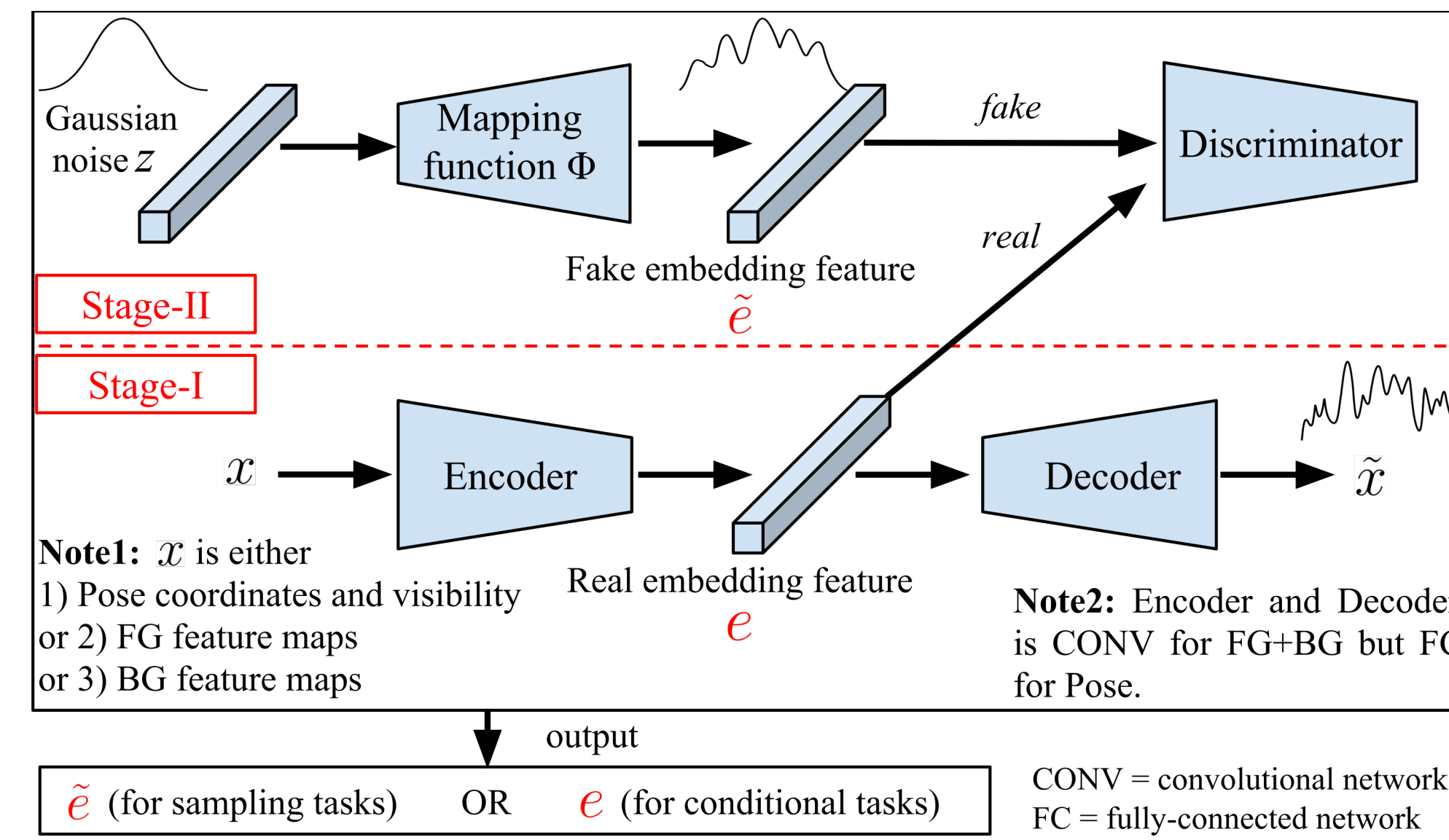


Introduction

- Motivation:** Learn image generation model for persons that explicitly represents foreground, background, and pose.
- Task:** Synthesize person images, while independently controlling foreground, background, and pose, in a *self-supervised* way.
- Key idea:** *Disentangle* person images into the aforementioned components, and then combine.
- Contributions:**
 - 1) A new task of generating person images by disentangling the input into weakly correlated factors.
 - 2) A two-stage framework to learn manipulatable embedding features.
 - 3) A technique to match the distribution of real and fake embedding features through adversarial training.
 - 4) An approach to generate image pairs for re-ID.

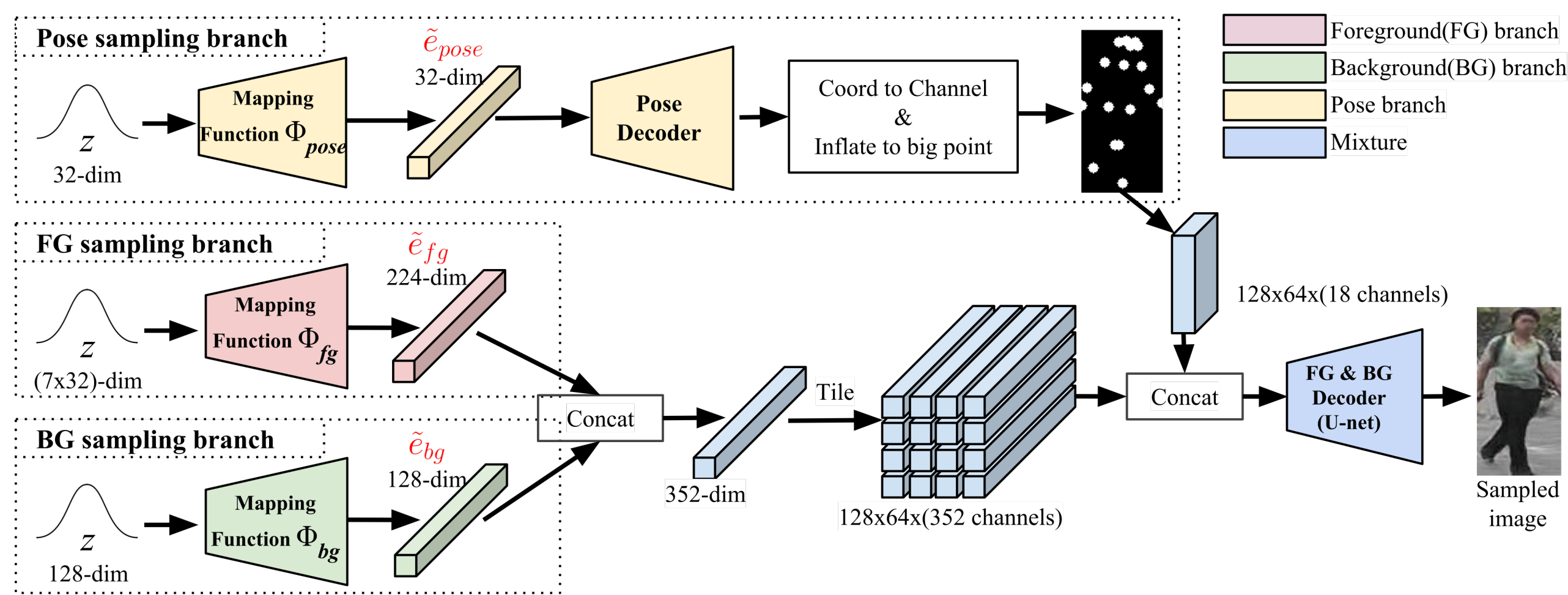


Two-stage learning framework



- Stage-I:** Learn the disentangled embeddings in a reconstruction pipeline.
- Stage-II:** Learn the mapping function from Gaussian space to embedding space for each factor.

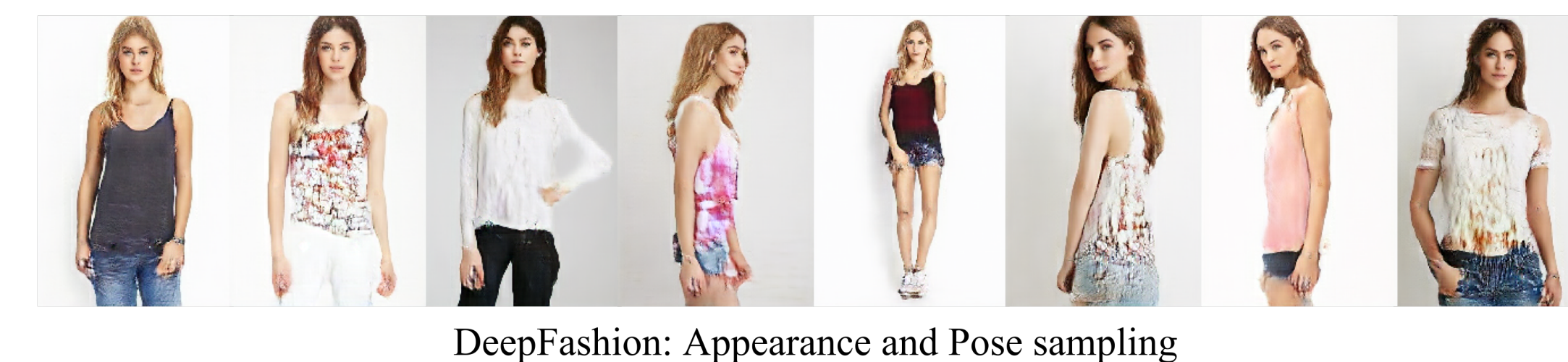
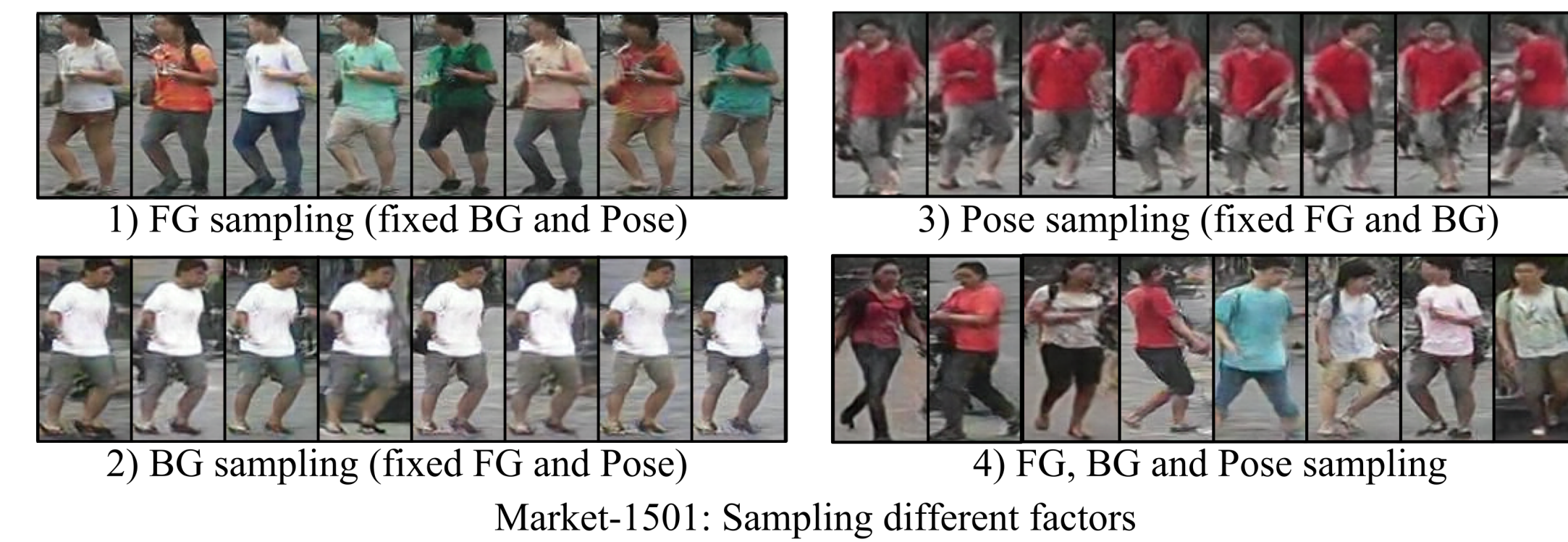
Sampling phase (testing)



- Sampling phase:** Sample foreground, background and pose from Gaussian noise to compose new person images.

Generation results

Sampling results



Person re-ID results

- 1) Our Virtual Market (VM) dataset.



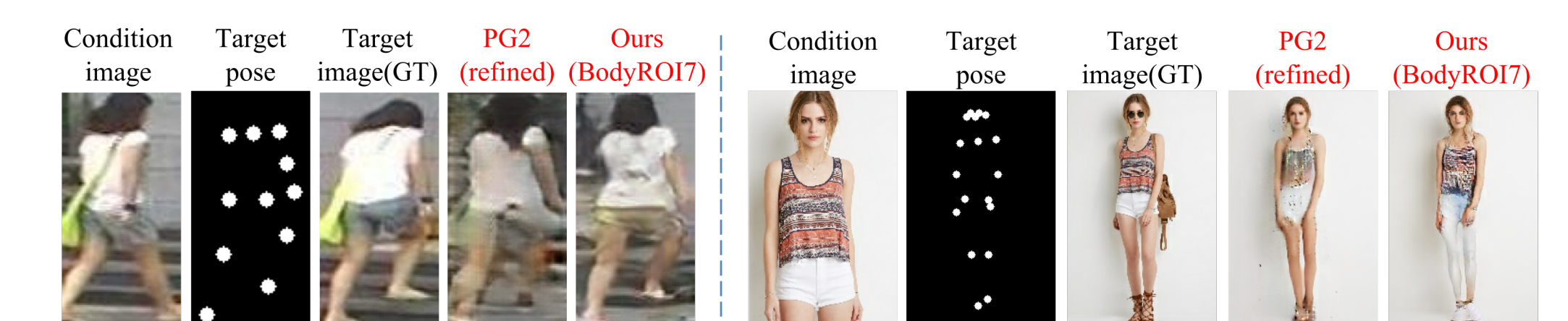
- 2) Person re-ID evaluation.

Model	Training data	Rank-1	mAP
Res50	CUHK03 (labeled)	0.300	0.115
Res50	Duke (labeled)	0.361	0.142
Res50	VM	0.338	0.134
Res50+PUL	VM+Market	0.369	0.156
Res50+PUL+KISSME	VM+Market	0.375	0.154

Inverse interpolation



Pose guided person image generation



Model	DeepFashion		Market-1501			
	SSIM	IS	SSIM	IS	Mask-SSIM	Mask-IS
PG2[1]	0.762	3.090	0.253	3.460	0.792	3.435
Ours	0.614	3.228	0.099	3.483	0.614	3.491

[1] Ma *et al.* Pose Guided Person Image Generation. NIPS'17.

[2] Fan *et al.* Unsupervised person reidentification: Clustering and fine-tuning. Arxiv' 17.