NoiseCollage: A Layout-Aware Text-to-Image Diffusion Model Based on Noise Cropping and Merging

Supplementary Material

This supplementary material shows additional results generated by NoiseCollage. Each of the following figures shows the layout condition L and the text conditions (S, s_*) , the generated image x_0 , and N individual object images cropped from x_0 by l_1, \ldots, l_N , from top to bottom. These cropped images not only show the detailed appearance of the individual objects but also show whether the objects are generated at their right place.

8. Total Inference step and Time efficiency

NoiseCollage (also Collage Diffusion [31]) requires O(NT)-times noise estimations, whereas Paint-with-words requires O(T)-times where N and T denote the number of objects in a layout and total denoising step, respectively. Therefore, in NoiseCollage, the total inference step becomes (N+1)*T including noise estimation for the whole image then the number of total inference step increases with the number of objects in a layout. In fact, NoiseCollage needs 25.7s to generate a single image with N=5, whereas Paint-with-words needs 7.42s on a single Nvidia A100GPU.

However, from an optimistic perspective, NoiseCollage can be refined through parallelization, as the O(N)-times noise estimations at each time step t are entirely independent. Consequently, it can be executed with O(T) computations.

9. Good and bad cases in the results of NoiseCollage

Figs. 8 and 9 show the images generated by NoiseCollage on the MD30 dataset. The former shows images with good scores, while the latter shows images with bad scores, according to the evaluation metric (multimodal similarity between the n-th object image and its text condition s_n) of Sec. 4.5. The good cases of Fig. 8 show the accurate correspondence between s_n and l_n , even the layout l_n is specified by a bounding box.

Even in the worst cases of Fig. 9, large objects (specified by l_1 and s_1) are generated at the right place with a correct appearance (except for the bicycle image). Except for the remote control that disappears from the generated image, small objects are generated in misaligned locations and still look good.

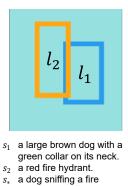
10. More Results of NoiseCollage with Control-Net

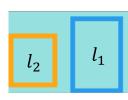
While we already showed several results of NoiseCollage with ControlNet [40] in Fig. 6, we show more results in Figs. 10, 11, and 12, which uses edge images, sketches, and pose skeletons as additional constraints, respectively. Like the results in Fig. 6, the additional results also show how the conditions for ControlNet guide the output images accurately. Note that bounding boxes specify the layout conditions, whereas polygons are used in Fig. 6. We can confirm that bounding boxes are also easy but appropriate layout conditions.

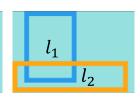
11. Results of more crowded layouts

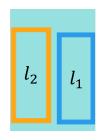
As already stated in the "Limitations" section, it is difficult for not only ours but also baselines to generate images under complex layouts with small objects or a large number of objects. This limitation may come from the fact that the common backbone, StableDiffusion [28], uses a low-resolution latent space of 64×64 .

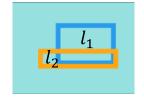
Fig. 13 shows the generated images with more crowded objects (N=7,9) by NoiseCollage with ControlNet. In these examples, the layout conditions are well reflected in the resulting images. However, if we want to put more objects, say, N=20, it is difficult to expect the accurate reflection of their conditions, as noted above.











- hydrant in a park.
- s_1 a black and white dog on s_1 a young boy in white and a leash sitting on the ground.
- s_2 a close up of a pink park bench.
- s_* a dog sitting on a leash next to a bench.
- black swim suit on a surfboard.
- s_2 a red surfboard on the water.
- s_* a man riding a surfboard in the ocean.
- the back of a zebra. s_1 the back of a zebra.
- two zebras standing next s_2 a white surfboard in the to each other in front of a fence.
- s_1 a muscular man on a surfboard.
 - water.
 - s_* A man surfing on a surfboard in the ocean.























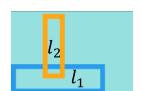




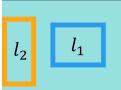




Figure 8. The best five cases by NoiseCollage on MD30. The lower part shows N individual object images cropped from x_0 by l_1, \ldots, l_N .



- s_1 a black horse with a saddle on its back.
- s_2 a standing young man. s_* a black and white photo of a man riding on the back of a horse.



man with helmet is riding

and a person is standing next to two parked bikes on a sidewalk.

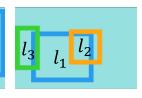
 s_* a man riding a motor bike

on the front.

on a motorcycle.

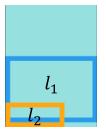


- s_1 a bicycle with a basket in a field of grass. s_2 a close up photo that a
 - s_2 a woman with green hair, s_3 a plant in a vase. sunglass, red rain boots. s_* a bench sitting in front of
 - s_* a woman standing next to a horse in a garden.



- s_1 a brown horse is standing s_1 a bench.
 - s_2 a plant in a vase.

 - a building.



- s_1 a couch with pillows on it in a living room.
- s_2 a remote control sitting on top of a round table.
- s_* a living room with a couch and a table.































Figure 9. The worst five cases by NoiseCollage on MD30.

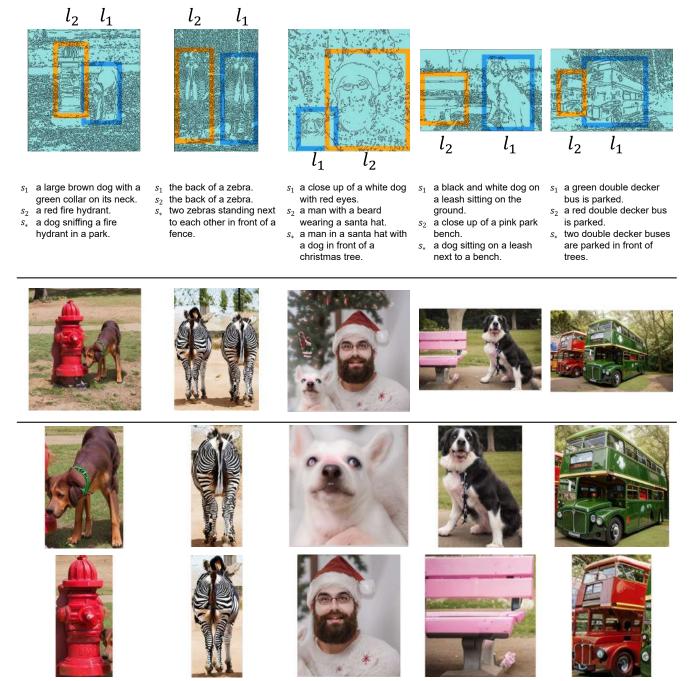


Figure 10. Images generated by NoiseCollage with ControlNet[40] of an edge image condition.

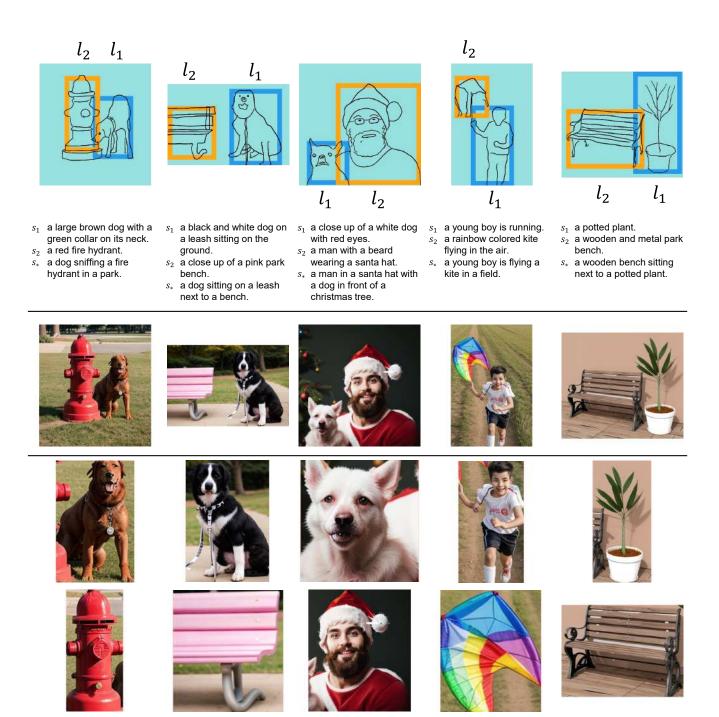
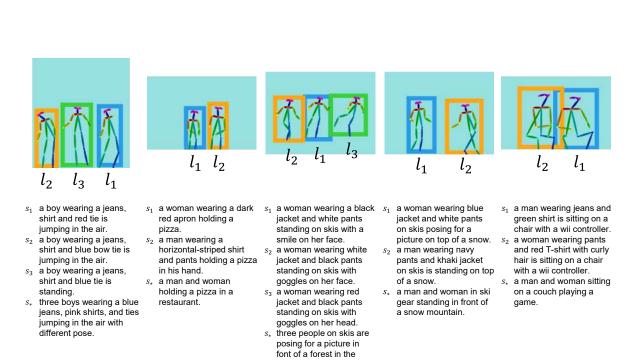


Figure 11. Images generated by NoiseCollage with ControlNet[40] of a sketch condition.



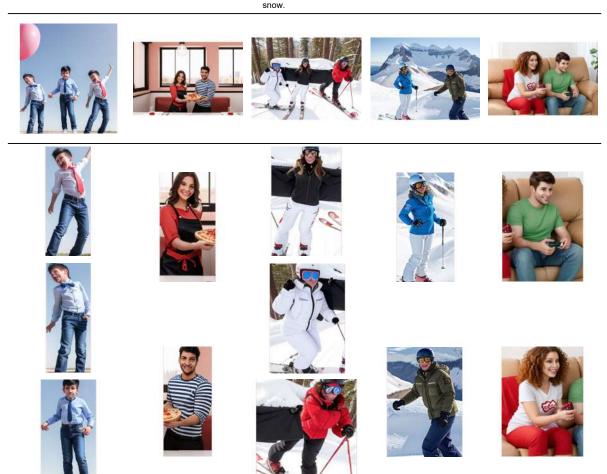
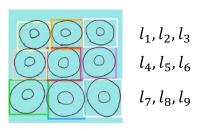
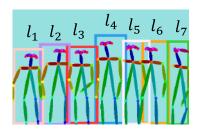


Figure 12. Images generated by NoiseCollage with ControlNet[40] of a pose skeleton condition.



 s_1 , s_2 , s_3 a purple donut s_4 , s_5 , s_6 a yellow donut s_7 , s_8 , s_9 a chocolate donut s_* a box of donuts with different color



 s_1 a woman in a black dress

 s_2 a man in a shirt and tie

 s_3 a woman in a white wedding dress

 s_4 a man in a navy suit

 s_5 a man in a shirt and tie

 s_6 a woman in a blue dress

 s_7 a young man in a shirt and jeans

 s_* a group of people posing for a picture in a room





Figure 13. Generated images with more complex layouts. Ours can control the layout of donuts and the clothes of each person.