

# Conditional Generation for Inverse Problems and Class/Text-Based Conditioning

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## A. Posterior Samplers for Inverse Problems

for each of the following tasks, I got results for three images from each of the two datasets, for each of the four inverse problems. I will only be including one image from each dataset for each task in the report, but the full results are included in the zip file submission. Please refer to the appendix for the images corresponding to each method and dataset. I will report the PSNR, SSIM, and LPIPS for the images shown in the report, but the full results across all three images are also included in the zip file submission.

I used my lab's server to generate the images for this task, which has an NVIDIA A6000, with 1TB ram and 128 CPU cores.

1) A: Predict  $\hat{x}_0$  and sample ddim: I implemented the DDIM from HW3, the results are shown in Figure 1. The images generated look qualitatively similar to those from HW3, and the model for each dataset creates images from that distribution.

2) B: Implement and compare posterior samplers: I found that an ILVR weight of 0.8 worked well during initial tests, so I used that value for all tasks. For this task I included the results from CelebA-HQ and ImageNet datasets in Figures 2 and 3.

The CelebA model had an average performance of:

- Time: 50s
- SRx4:

PSNR: 30.97

SSIM: 0.880

LPIPS: 0.0729

- SRx8:

PSNR: 26.44

SSIM: 0.7464

LPIPS: 0.1243

- 80% random inpainting:

PSNR: 20.86

SSIM: 0.5541

LPIPS: 0.4891

- 128x128 box inpainting:

PSNR: 20.33

SSIM: 0.8244

LPIPS: 0.1231

The ImageNet model had an average performance of:

- Time: 176s

- SRx4:

PSNR: 25.33

SSIM: 0.735

LPIPS: 0.2653

- SRx8:

PSNR: 22.10

SSIM: 0.5647

LPIPS: 0.2642

- 80% random inpainting:

PSNR: 15.95

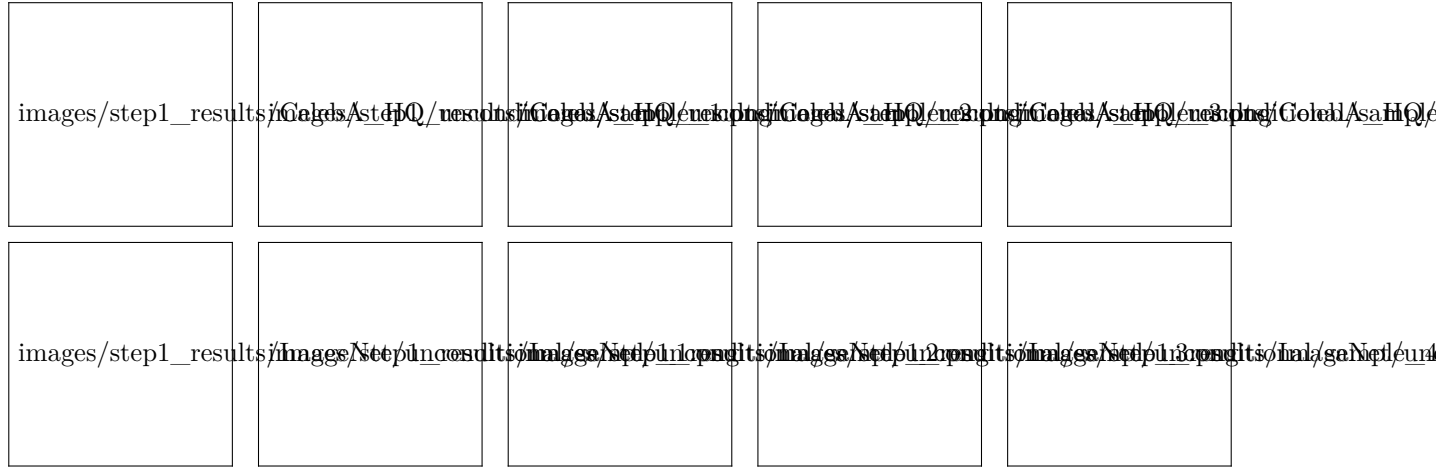


Fig. 1. Unconditional samples from CelebA-HQ (top row) and ImageNet (bottom row) pretrained ADM models using DDIM sampling with 1000 steps.

SSIM: 0.1889	SSIM:0.8047
LPIPS: 0.9146	LPIPS: 0.1856
- 128x128 box inpainting:	- Time: 178s
PSNR: 17.23	- SRx4:
SSIM: 0.7826	PSNR: 12.75
LPIPS: 0.2218	SSIM: 0.3567
3) C: Manifold Constrained Gradient (MCG):	LPIPS: 0.6224
MCG produced	- SRx8:
- Time: 55s	PSNR: 14.46
- SRx4:	SSIM: 0.3693
PSNR: 19.68	LPIPS: 0.4716
SSIM: 0.8047	- 80% random inpainting:
LPIPS: 0.1846	PSNR: 23.98
- SRx8:	SSIM: 0.7838
PSNR: 25.68	LPIPS: 0.2242
SSIM: 0.7404	- 128x128 box inpainting:
LPIPS: 0.1335	PSNR: 14.11
- 80% random inpainting:	SSIM: 0.7472
PSNR: 33.40	LPIPS: 0.3034
SSIM: 0.9260	4) D: Denoising Diffusion Null-Space Model
LPIPS: 0.0359	(DDNM):
- 128x128 box inpainting:	- Time: 3s
PSNR: 19.68	- SRx4:

PSNR: 18.25	PSNR:
SSIM: 0.2746	SSIM:
LPIPS: 0.8742	LPIPS:
- SRx8:	- SRx8:
PSNR: 17.02	PSNR:
SSIM: 0.2033	SSIM:
LPIPS: 1.0188	LPIPS:
- 80% random inpainting:	- 80% random inpainting:
PSNR: 10.87	PSNR:
SSIM: 0.0629	SSIM:
LPIPS: 14.827	LPIPS:
- 128x128 box inpainting:	- 128x128 box inpainting:
PSNR: 15.83	PSNR:
SSIM: 0.7219	SSIM:
LPIPS: 0.5116	LPIPS:
- Time: 9s	- Time:
- SRx4:	- SRx4:
PSNR: 13.33	PSNR:
SSIM: 0.3296	SSIM:
LPIPS: 0.8221	LPIPS:
- SRx8:	- SRx8:
PSNR: 12.76	PSNR:
SSIM: 0.2855	SSIM:
LPIPS: 0.9485	LPIPS:
- 80% random inpainting:	- 80% random inpainting:
PSNR: 8.03	PSNR:
SSIM: 0.0324	SSIM:
LPIPS: 1.4114	LPIPS:
- 128x128 box inpainting:	- 128x128 box inpainting:
PSNR: 13.24	PSNR:
SSIM: 0.7202	SSIM:
LPIPS: 0.5390	LPIPS:

5) E: Noisy measurements and Diffusion Posterior Sampling (DPS):

- Time:
- SRx4:

6) F: Diffusion Posterior Sampling (DPS) for noisy measurements:

- Time: 55s
- SRx4:

PSNR: 22.94

SSIM: 0.7031

LPIPS: 0.2730

- SRx8:

PSNR: 26.43

SSIM: 0.7592

LPIPS: 0.1143

- 80% random inpainting:

PSNR: 19.02

SSIM: 0.7389

LPIPS: 0.3173

- 128x128 box inpainting:

PSNR: 11.92

SSIM: 0.3863

LPIPS: 0.5397

- Time: 177s

- SRx4:

PSNR: 13.22

SSIM: 0.3813

LPIPS: 0.4515

- SRx8:

PSNR: 16.65

SSIM: 0.4041

LPIPS: 0.3862

- 80% random inpainting:

PSNR: 19.34

SSIM: 0.5619

LPIPS: 0.4727

- 128x128 box inpainting:

PSNR: 13.69

SSIM: 0.5530

LPIPS: 0.4243

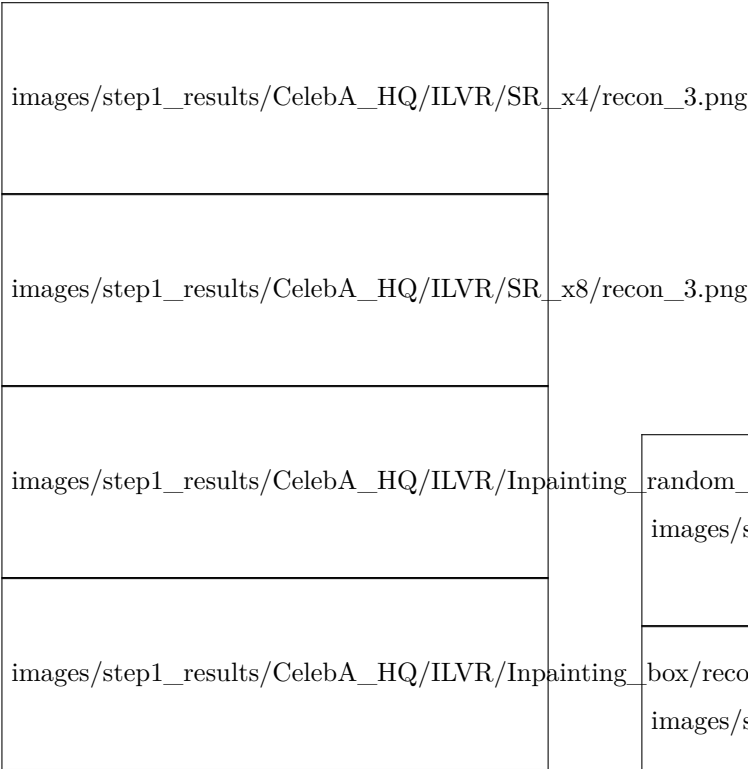


Fig. 2. ILVR Reconstructions on CelebA-HQ for SRx4, SRx8, 80% random inpainting, and 128x128 box inpainting.

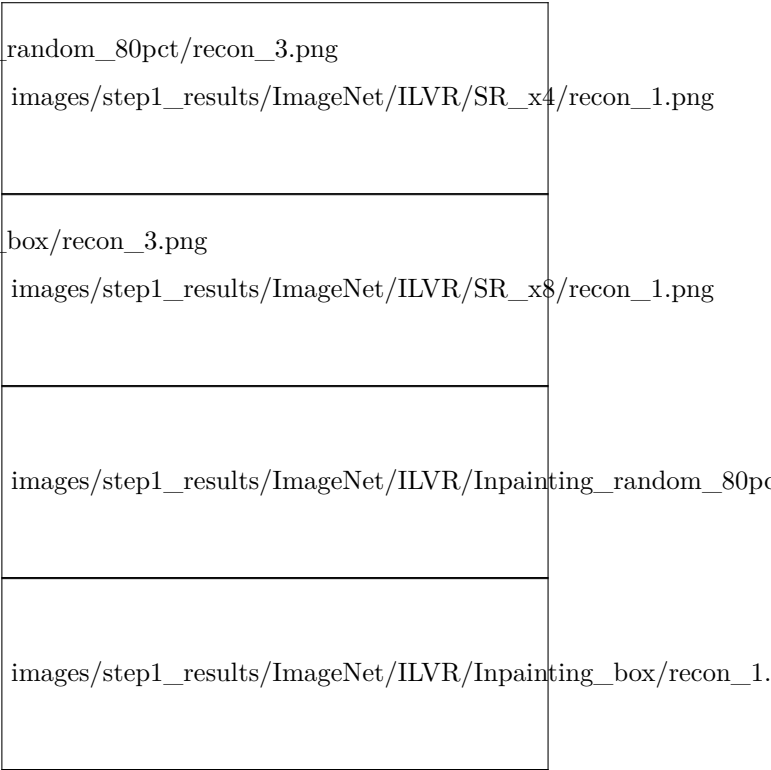


Fig. 3. ILVR Reconstructions on ImageNet for SRx4, SRx8, 80% random inpainting, and 128x128 box inpainting.



Fig. 4. MCG Reconstructions on CelebA-HQ for SRx4, SRx8, 80% random inpainting, and 128x128 box inpainting.

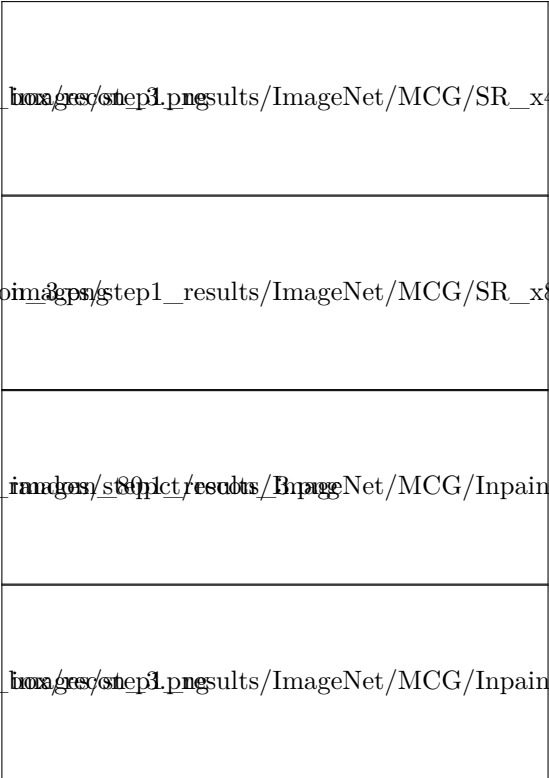


Fig. 5. MCG Reconstructions on ImageNet for SRx4, SRx8, 80% random inpainting, and 128x128 box inpainting.

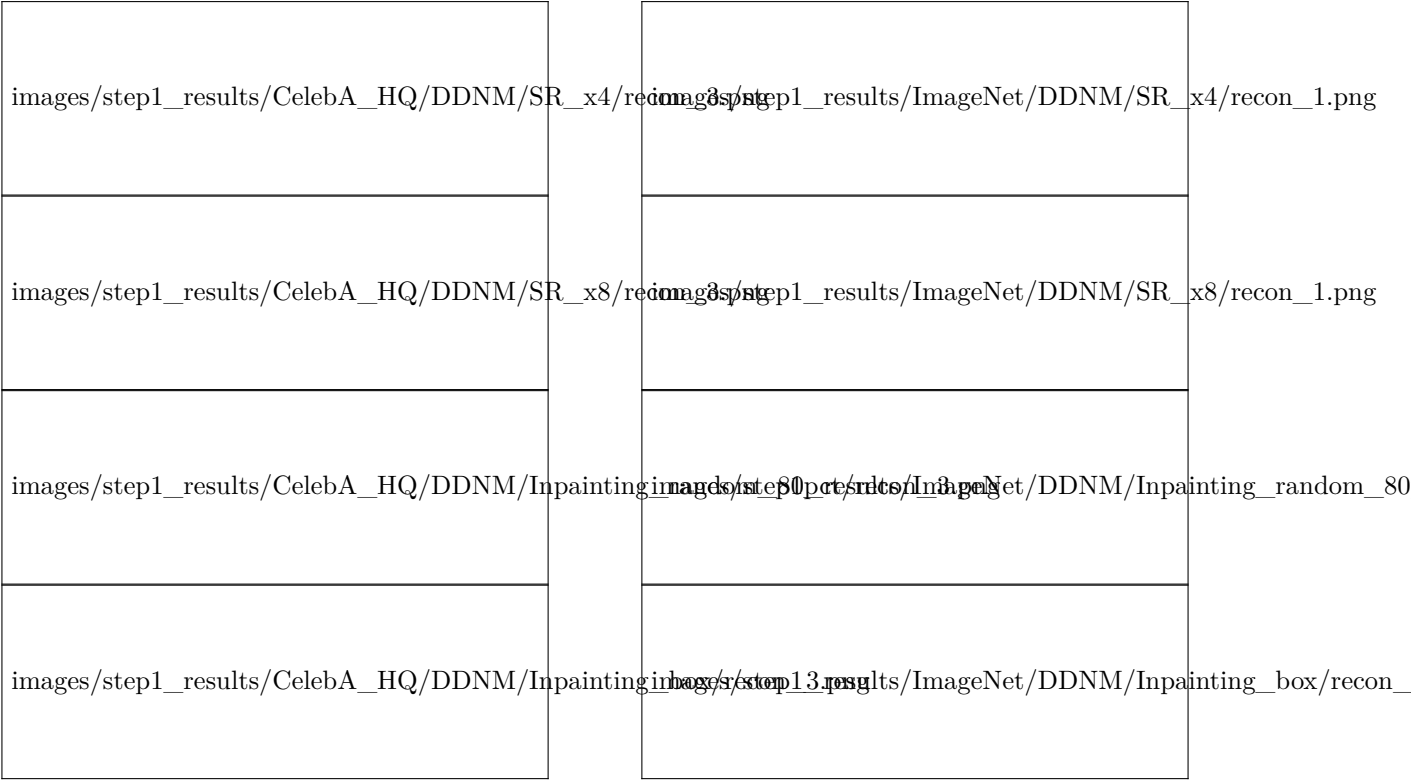


Fig. 6. DDNM Reconstructions on CelebA-HQ for SRx4, SRx8, 80% random inpainting, and 128x128 box inpainting.

Fig. 7. DDNM Reconstructions on ImageNet for SRx4, SRx8, 80% random inpainting, and 128x128 box inpainting.

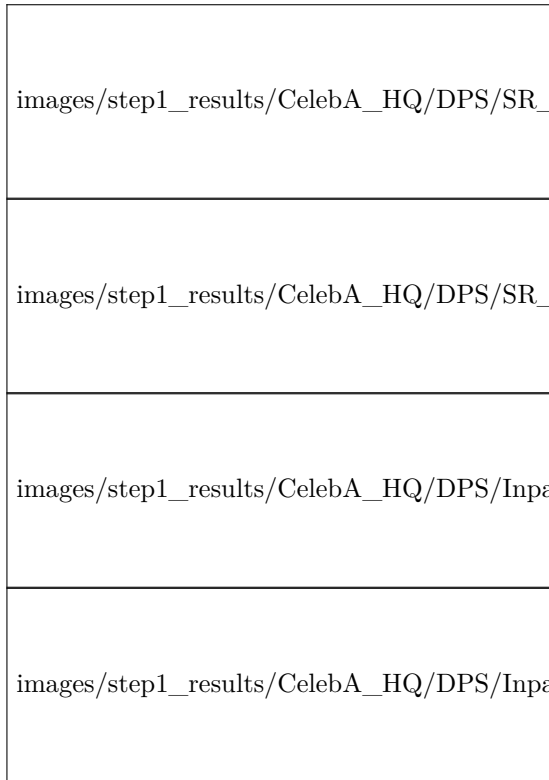


Fig. 8. DPS Reconstructions on CelebA-HQ for SRx4, SRx8, 80% random inpainting, and 128x128 box inpainting with noisy measurements.

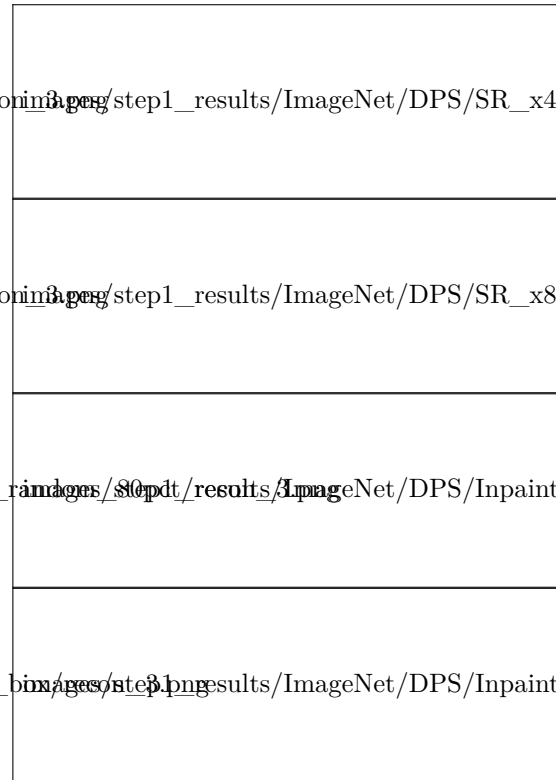


Fig. 9. DPS Reconstructions on ImageNet for SRx4, SRx8, 80% random inpainting, and 128x128 box inpainting with noisy measurements.