

Any-resolution Datasets



Dataset Collection



Base Resolution Dataset LSUN [Yu 2015]

- Use standard base resolution dataset to learn structure (~95%) - Add few thousand images at larger, varied resolutions to learn texture (~5%)



Any-resolution Training for High-resolution Image Synthesis Lucy Chai¹, Michaël Gharbi², Eli Shechtman², Phillip Isola¹, Richard Zhang² ¹MIT CSAIL ²Adobe Research

High Resolution Dataset Collected from Flickr



 $G^* = \arg\min_{C} \max_{D} \mathbb{E}_{z,\{x,s,\mathbf{v}\} \sim \mathcal{D}} V(D, G(z, c_{\mathbf{v},s}, x) + \lambda_{teacher} \mathcal{L}_{teacher}(G, G_{fixed}, z) - \frac{\kappa_1}{2} R_1(D, x)$ non-saturating GAN loss teacher reconstruction R1 regularization Images parametrized as continuous surfaces: coordinate grid c controlled by

- scale s, center v, and patch size p
- Transform generator coordinates using sampled patch parameters
- Teacher loss encourages similarity between patch and smaller global image

Arbitrary Scale Synthesis with Continuous Images



- Sampling the coordinates generates image patches at arbitrary scales



- Measure similarity between real/fake patches,

- Correlated with standard FID on ground truth FFHQ

	FFHQ6K	Church	Birds	Mountain
	22.93	83.88	30.19	23.10
AN	16.92	23.04	16.10	19.05
(Ours)	2.96	9.89	6.52	7.99







Base Image



- Super-resolution methods can amplify artifacts or over-smooth the result - Any-res GAN tolerates distortions, can mitigate artifacts in base image





https://chail.github.io/anyres-gan

- Seamless generation of large images despite training with patches

