

Source Separation with Deep Generative Priors

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Bayesian Source Separation

Unobserved Components: $\mathbf{x} = (\mathbf{x}_1, \dots, \mathbf{x}_k)$

Observed Mixture: \mathbf{m}

Prior (deep generative model): $\mathbf{x} \sim p$

$$\text{Likelihood: } \mathbf{m} \sim \mathcal{N} \left(\sum_{i=1}^n \mathbf{x}_i, \gamma^2 I \right)$$

$$\text{Posterior likelihood: } p(\mathbf{x}|\mathbf{m}) = \frac{p(\mathbf{m}|\mathbf{x})p(\mathbf{x})}{p(\mathbf{m})}$$

Sampling with Langevin Dynamics

Smoothed prior (convolve with Gaussian): $p_\sigma(\mathbf{x})$

$$\text{Smoothed posterior: } p_\sigma(\mathbf{x}|\mathbf{m}) = \frac{p(\mathbf{m}|\mathbf{x})p_\sigma(\mathbf{x})}{p(\mathbf{m})}$$

Innovation Noise: $\varepsilon \sim \mathcal{N}(0, I)$

Noise-Annealed Langevin Dynamics:

$$\mathbf{x}^{(t+1)} \equiv \mathbf{x}^{(t)} + \eta \nabla_{\mathbf{x}} \log p_\sigma(\mathbf{x}^{(t)}|\mathbf{m}) + 2\sqrt{\eta} \varepsilon_t$$

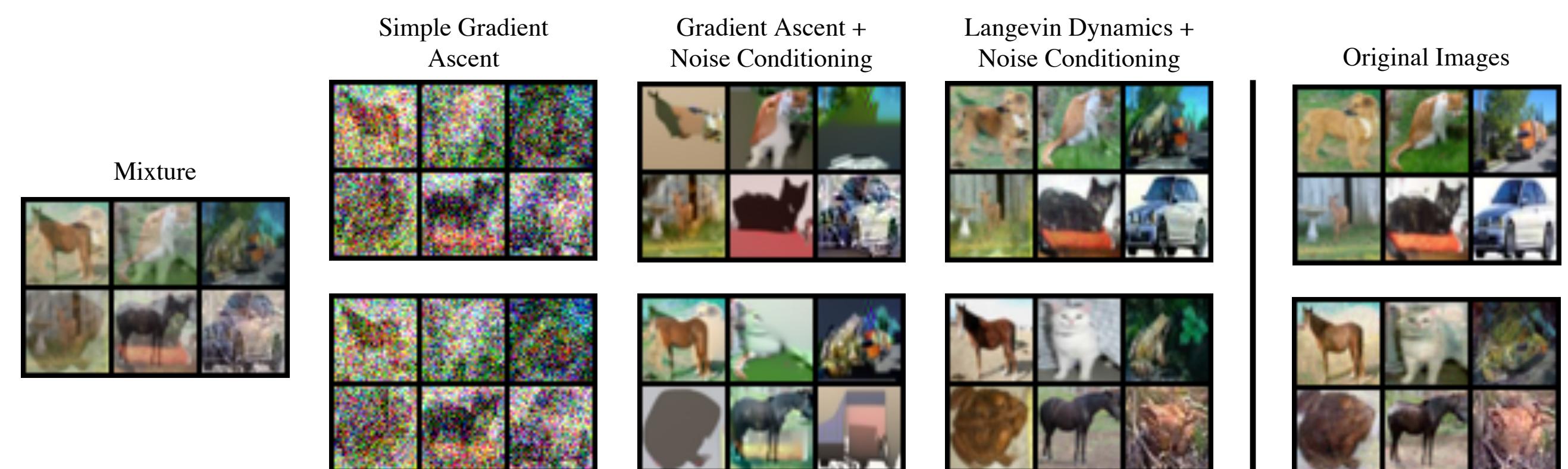
A Robust Practical Algorithm, with a Qualitative Ablation Study

Algorithm 1 BASIS Separation

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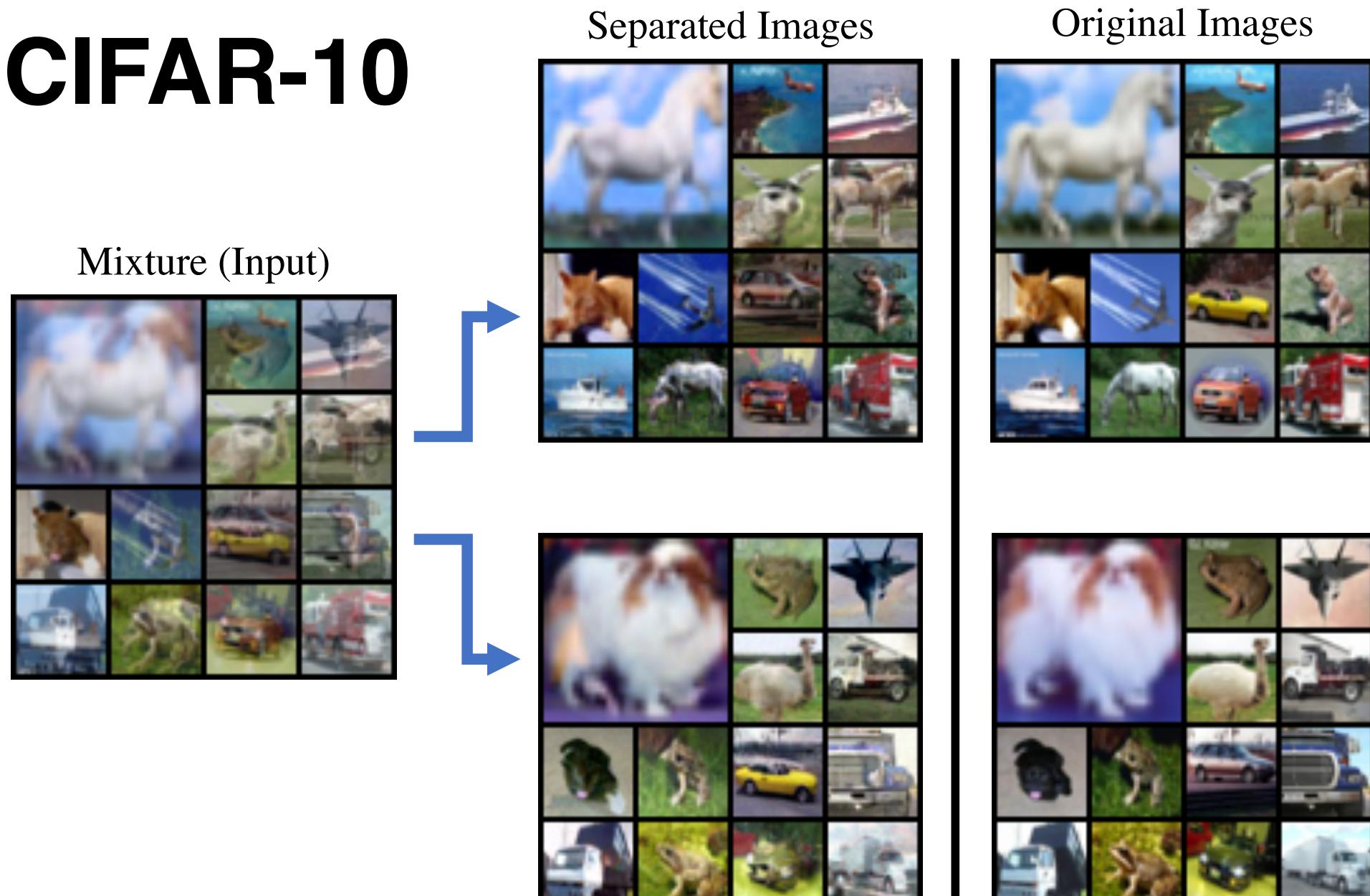
Input:  $\mathbf{m} \in \mathcal{X}, \{\sigma_i\}_{i=1}^L, \delta, T$ 
Sample  $\mathbf{x}_1, \dots, \mathbf{x}_k \sim \text{Uniform}(\mathcal{X})$ 
for  $i \leftarrow 1$  to  $L$  do
   $\eta_i \leftarrow \delta \cdot \sigma_i^2 / \sigma_L^2$ 
  for  $t = 1$  to  $T$  do
    Sample  $\varepsilon_t \sim \mathcal{N}(0, I)$ 
     $\mathbf{u}^{(t)} \leftarrow \mathbf{x}^{(t)} + \eta_i \nabla_{\mathbf{x}} \log p_{\sigma_i}(\mathbf{x}^{(t)}) + 2\sqrt{\eta} \varepsilon_t$ 
     $\mathbf{x}^{(t+1)} \leftarrow \mathbf{u}^{(t)} - \frac{\eta_i}{\sigma_i^2} [\mathbf{m} - g(\mathbf{x}^{(t)})]$ 
  end for
end for

```



Preliminary Separation Results

CIFAR-10



LSUN

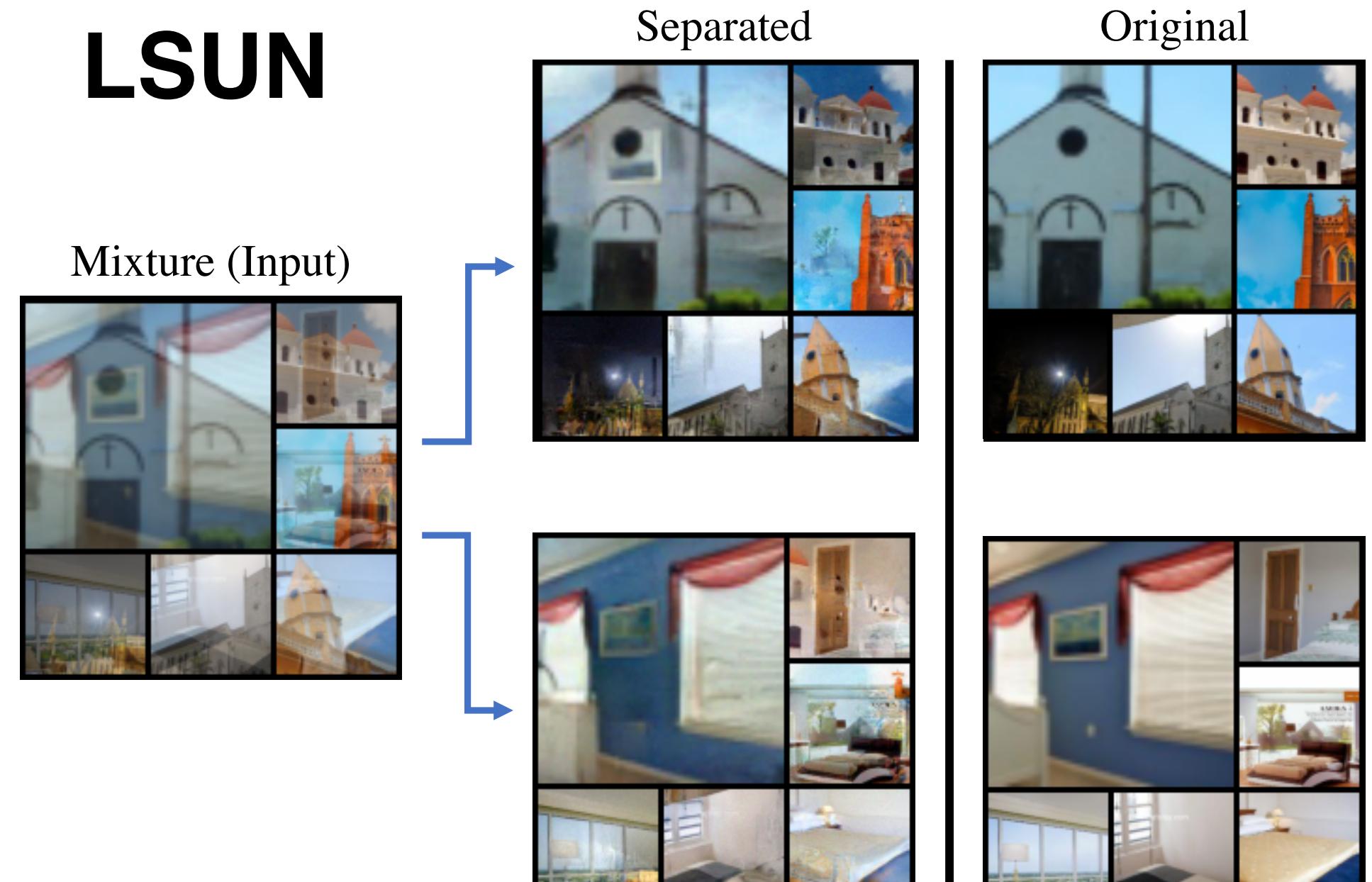


Image Colorization

Image Colorization is a corollary of our work

Color Image: $\mathbf{x} = (\mathbf{x}_r, \mathbf{x}_g, \mathbf{x}_b)$

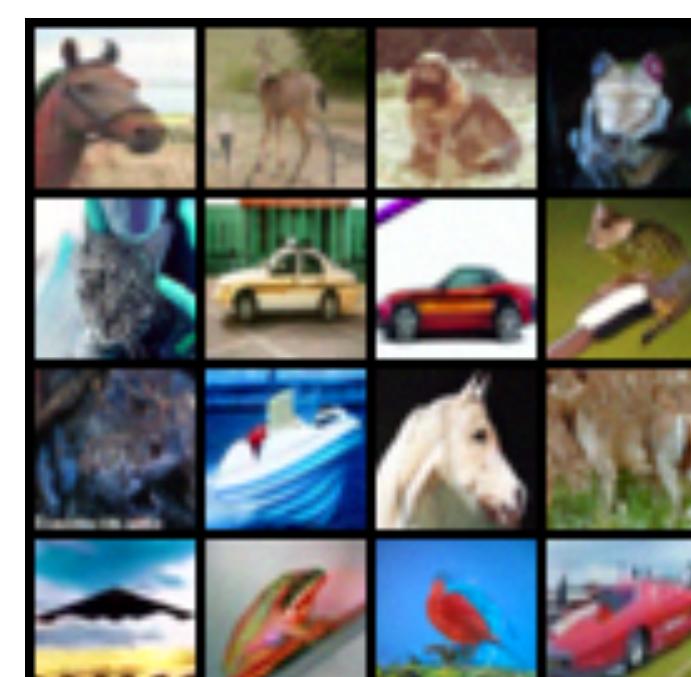
$$\text{Grayscale Image: } \mathbf{m} = \frac{\mathbf{x}_r + \mathbf{x}_g + \mathbf{x}_b}{3}$$

It may be possible to adapt our work to other conditional generation problems

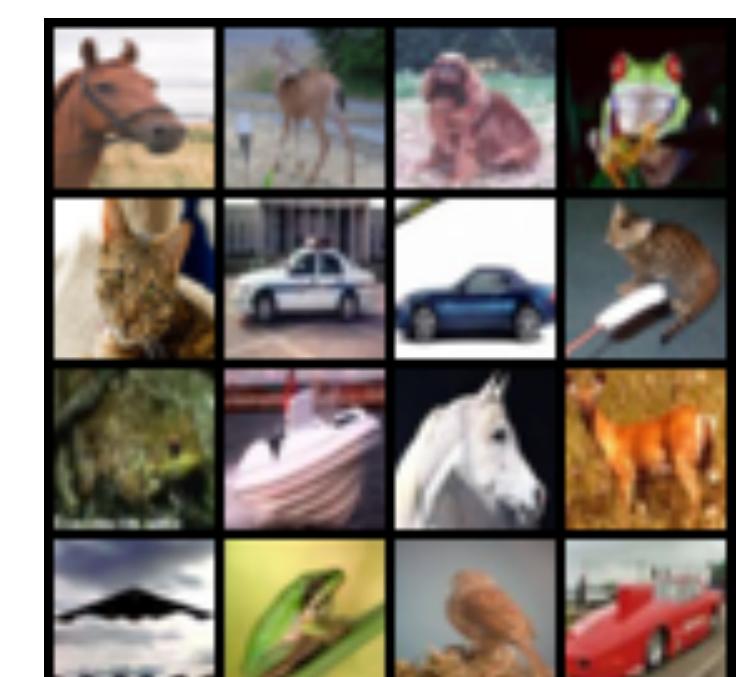
Grayscale (Input)



Colorization



Original



Preliminary Work:

<https://arxiv.org/abs/2002.07942>

GitHub Repo:

<https://github.com/jthickstun/basis-separation>