## Training Normalizing Flows

Maximum likelihood training to find parameters  $\theta$  that make our training samples likely under our parameterized model.

$$\max_{\theta} \mathbb{E}_{\mathbf{x} \sim p_{x}} p_{\theta}(\mathbf{x}) = \min_{\theta} \mathbb{E}_{\mathbf{x} \sim p_{x}} - \log p_{\theta}(\mathbf{x})$$

$$= \min_{\theta} \mathbb{E}_{\mathbf{x} \sim p_{x}} - \log \left[ p_{Z}(T_{\theta}(\mathbf{x})) \middle| \det \frac{\partial T_{\theta}(\mathbf{x})}{\partial \mathbf{x}} \middle| \right]$$

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$$\begin{aligned} \max_{\theta} \mathbb{E}_{\mathbf{x} \sim p_{x}} p_{\theta}(\mathbf{x}) &= \min_{\theta} \ \mathbb{E}_{\mathbf{x} \sim p_{x}} - \log p_{\theta}(\mathbf{x}) \\ &= \min_{\theta} \ \mathbb{E}_{\mathbf{x} \sim p_{x}} - \log \left[ p_{Z}(T_{\theta}(\mathbf{x})) \left| \det \frac{\partial T_{\theta}(\mathbf{x})}{\partial \mathbf{x}} \right| \right] \\ &= \min_{\theta} \ \mathbb{E}_{\mathbf{x} \sim p_{x}} \left[ \frac{1}{2} \| T_{\theta}(\mathbf{x}) \|_{2}^{2} - \log \left| \det \frac{\partial T_{\theta}(\mathbf{x})}{\partial \mathbf{x}} \right| \right] \end{aligned}$$