

Siddhant Mishra-Sharma (MIT/AI FI) Summer School

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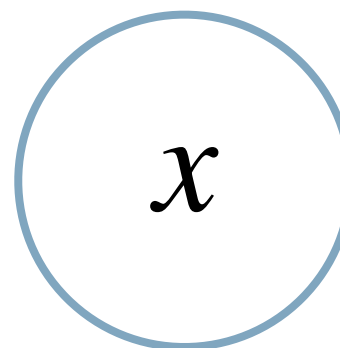
Normalizing flows

Flow

$f(z)$



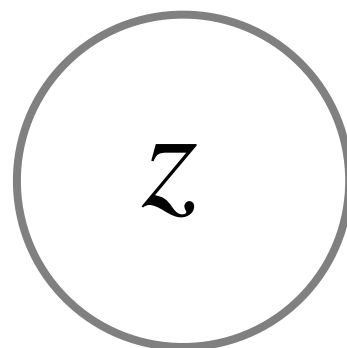
$p(x)$



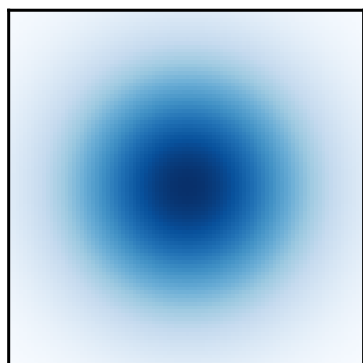
$f^{-1}(x)$



Normalizing



$p(z)$



Data likelihood

$$p(x) = p\left(f^{-1}(x)\right) |\det \nabla f|^{-1}$$


The distribution $p(z)$ should

- *Have an easy-to-evaluate density*
- *Be easy to sample from $z \sim p(z)$*

Typically

$$p(z) = \mathcal{N}(0, \mathbb{I})$$

The function f should be

- *One-to-one*
 - *Differentiable*
 - *Invertible*
- 
- Diffeomorphism
- *Tractable* f^{-1} and $\det \nabla f$

Normalizing flows

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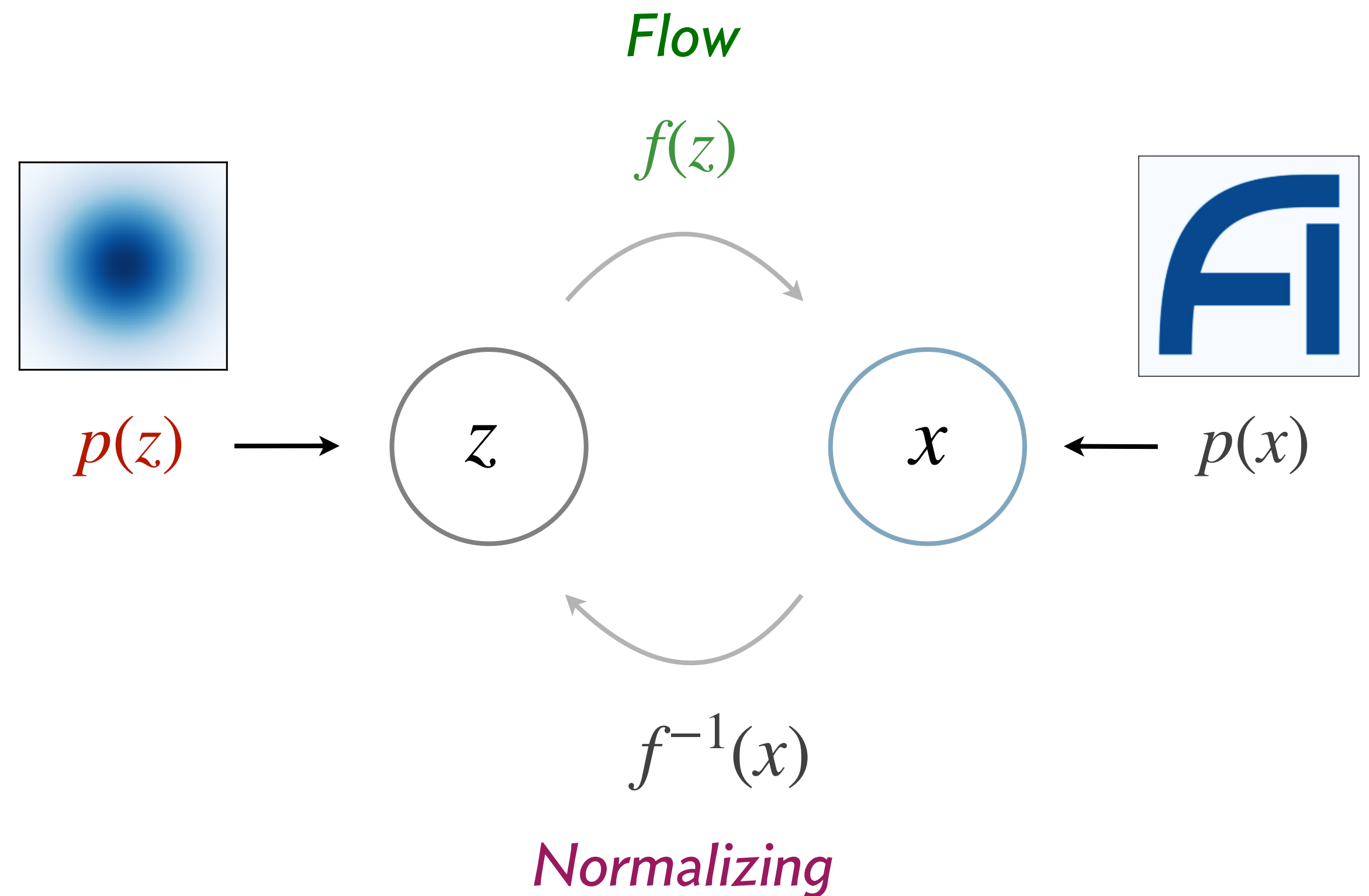
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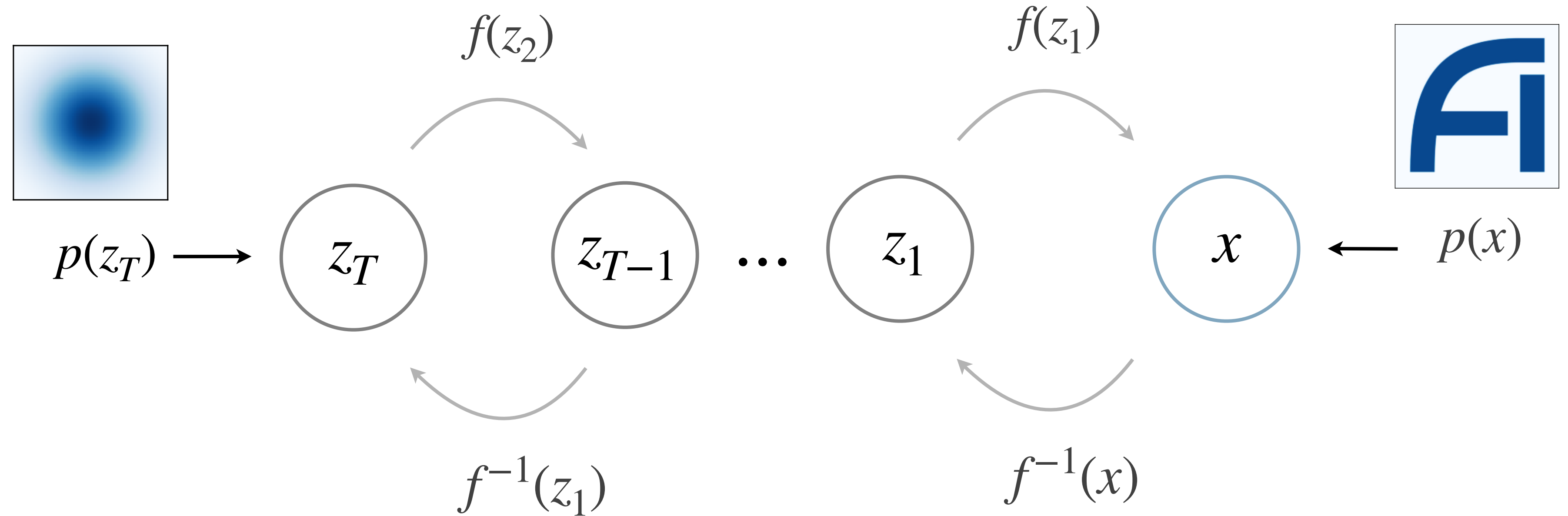


Data likelihood

$$p(x) = p(f^{-1}(x)) |\det \nabla f|^{-1}$$

Normalizing flows

Multiple flow transformation can be easily composed for e.g. more expressivity



Computing $p(x)$: *change-of-variables formula*

$$\int p(x) dx = \int p(z) dz = 1$$

$$p(x) = \pi(z) \left| \frac{dz}{dx} \right| = p(f^{-1}(x)) \left| \frac{df^{-1}}{dx} \right| = p(f^{-1}(x)) |\det \nabla f|^{-1}$$

Train using maximum-likelihood objective

$$\phi^* = \left\langle \arg \max_{\phi} p(f_{\phi}^{-1}(x)) |\det \nabla f_{\phi}|^{-1} \right\rangle_{x \sim p(x)}$$