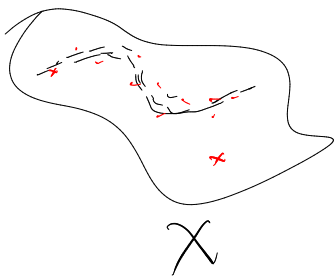
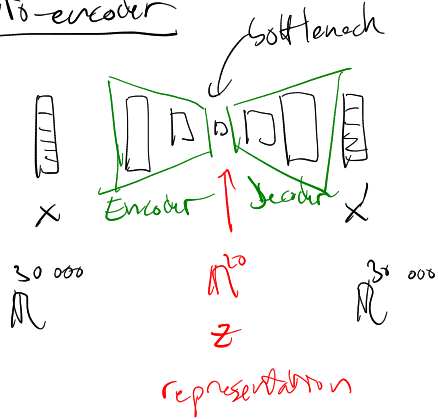


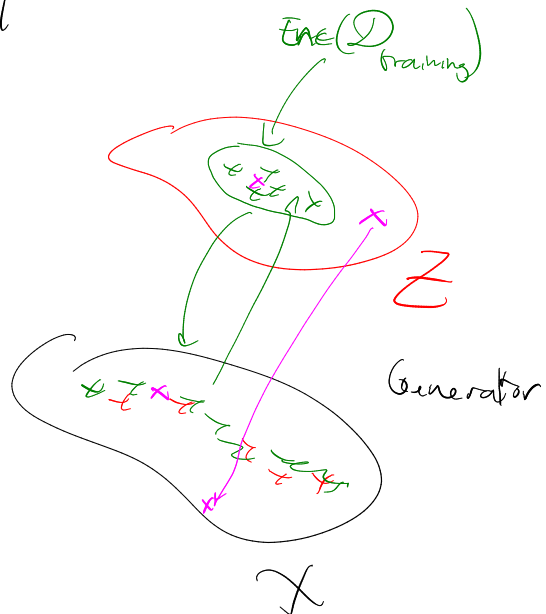
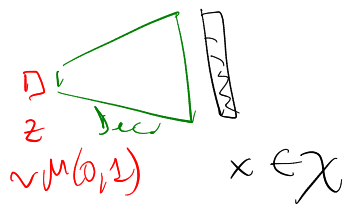
Generative models



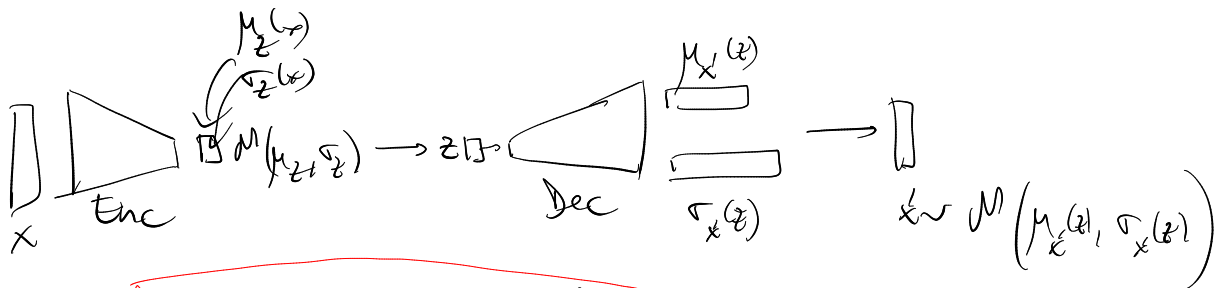
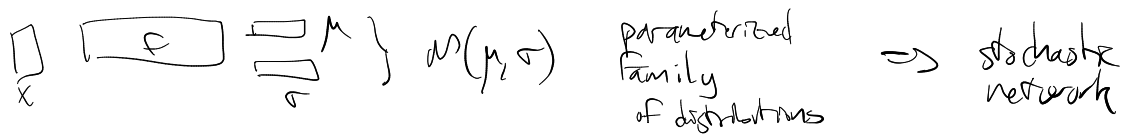
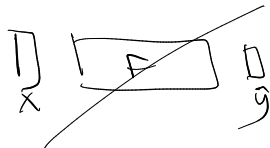
Auto-encoder



Loss: $\|x' - x\|$



Variational Auto-encoder (VAE)

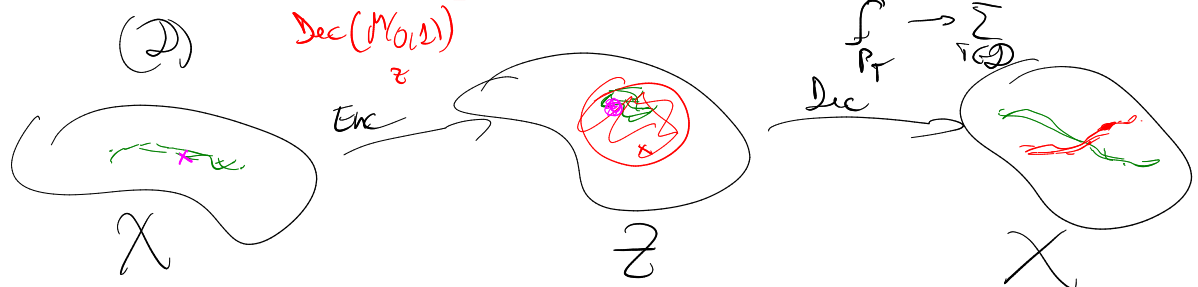


Training criterion:

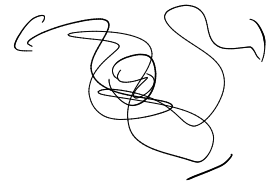
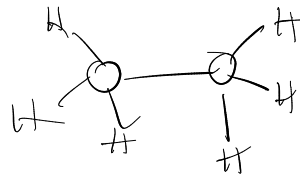
$$KL(P_{\text{target}} \parallel P_{\text{generated}})$$

→ ELBO → Monte Carlo approximation

Generations:
sample $z \sim N(\mu, \sigma)$



ex (not common) given molecule
H3C-C-H3



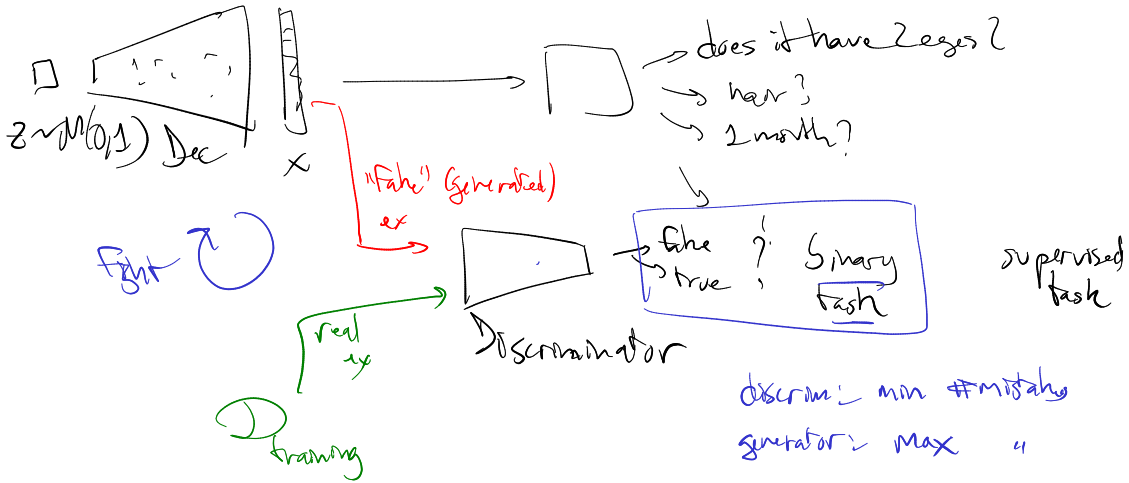
distribution of all 3D conformations

target distribution = Boltzmann distribution

$$P(c) \propto e^{-\beta E(c)}$$

usual ex: $\mathcal{D} \rightarrow$ empirically estimate $p(x)$

Adversarial setting: } GAN: Generative Adversarial Networks
 } DANN: Domain-Adversarial NN



\rightarrow linked to Kolmogorov test

Extension: Wasserstein GAN (WGAN)

Kantorovich-Rubinstein duality

"weight-clipping"

\mathcal{V} -penalty

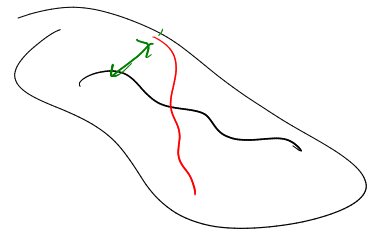
$$+ \left(\|\nabla_x F\| - 1 \right)^2$$

Optimal transport

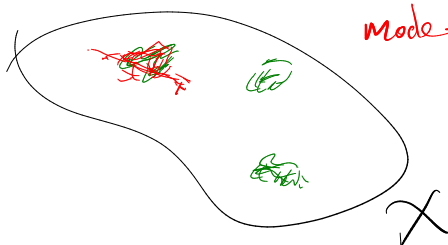
$$OT(P_G, P_T) = \sup_{F \text{ coupling}} E[F(x_G, x_T)]$$

\uparrow generated target

WGAN \in Lipschitz

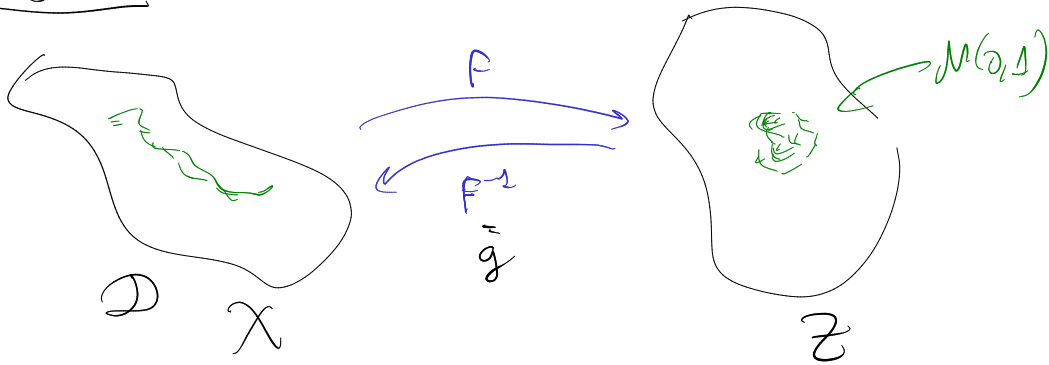


Issue:



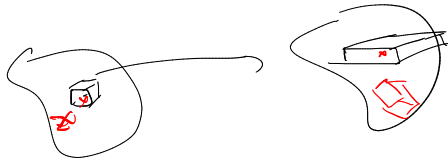
mode-drop:
 no feedback about parts of P_{target} not sampled

Normalizing flows



- find F st. $F(D) \simeq \mathcal{N}(0, I)$
- F is invertible
- \Rightarrow generate by computing $F^{-1}(z \sim \mathcal{N}(0, I))$

$g: Z \rightarrow X$



Jacobian:

$$\frac{dg}{dz} = J(x)$$

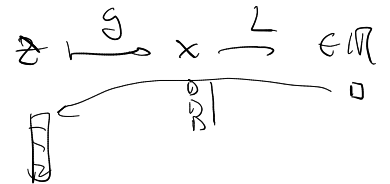
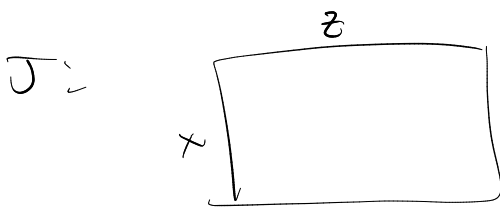
$z \sim p(z)$

$p(x = g(z))$

$= p(z) / \det \frac{dg}{dz}$

optimize $KL(p_{\text{target}} \parallel p_{\text{generated}})$

$-\log p_g(x) = -\log p_m(z) + \log \det J(x)$



$J = \frac{dg}{dz}$



$= \frac{df_1}{df_{12}} \times \frac{df_{12}}{df_{22}} \dots \frac{df_n}{dx}$



$\det J = \prod_l \det \frac{df_l}{df_{l2}}$

