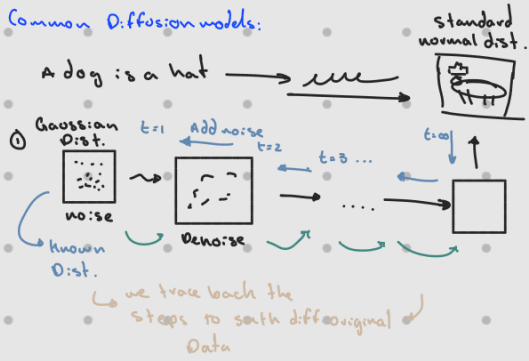
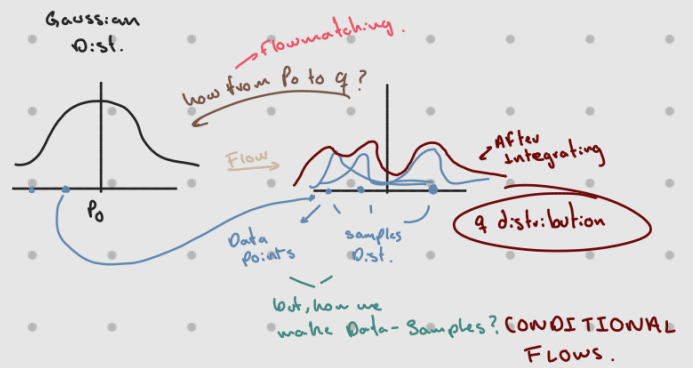
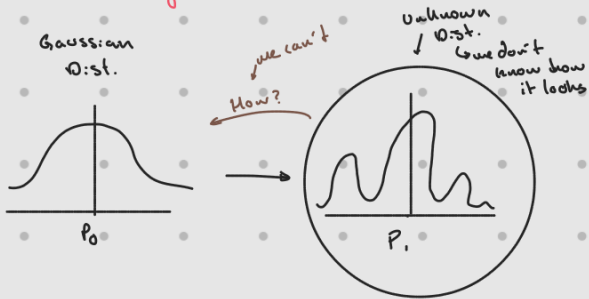


Flow matching.

Common Diffusion models:

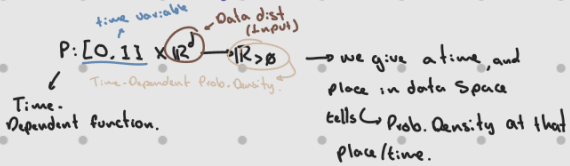


Flow-matching:



Continuous Normalizing Flows

1) Probability Density Path:



2) Time-Dependent vector field.

$v: [0, 1] \times \mathbb{R}^d \rightarrow \mathbb{R}^d$ → same look goes to Data Dist.

3) Time-Dependent diffeomorphic map [Flow]:

$$\phi: [0, 1] \times \mathbb{R}^d \rightarrow \mathbb{R}^d$$

↳ Flow connected to vector field by:

change flow $\rightarrow \frac{d}{dt} \phi_\epsilon(x)$ (time dependent) $\rightarrow v_\epsilon(\phi_\epsilon(x))$ (vector field, space dependent)

↳ Flow at time 0, at any point is that point



but how do we
get from t_0 to
 t_1 dist.

Crossing ODE solver
to solve flows until we
get to $t=x$

$$p_t = [\Phi_t]_* p_0$$

Goal Dist. = Flow * Original Dist.



How To GET VECTOR FIELD?

↳ what Algorithm must learn

Flow-matching:

$$\mathcal{I}_{FM}(\theta) = \mathbb{E}_x \left[\left\| p_\theta(x) - v_\theta(x) - u_\theta(x) \right\|^2 \right]$$

Annotations:
- $p_\theta(x)$: Prob. Dens. path
- $v_\theta(x)$: neural network
- $u_\theta(x)$: vector field (Generates Prob. Dens. path)
- \mathbb{E}_x : how to opt? (red arrow)
- $v_\theta(x)$: match for each pos. and time (blue arrow)
- $u_\theta(x)$: how to opt? (red arrow)

; we can't know, for now, the Prob density of target dist.
↳ we can approximate by samples with Gaussian models.

$$p_i(x) \approx q(x)$$

Constructing p_t, u_t from Conditional

Probability Paths and vector fields: