

# CS 180 Discussion 11

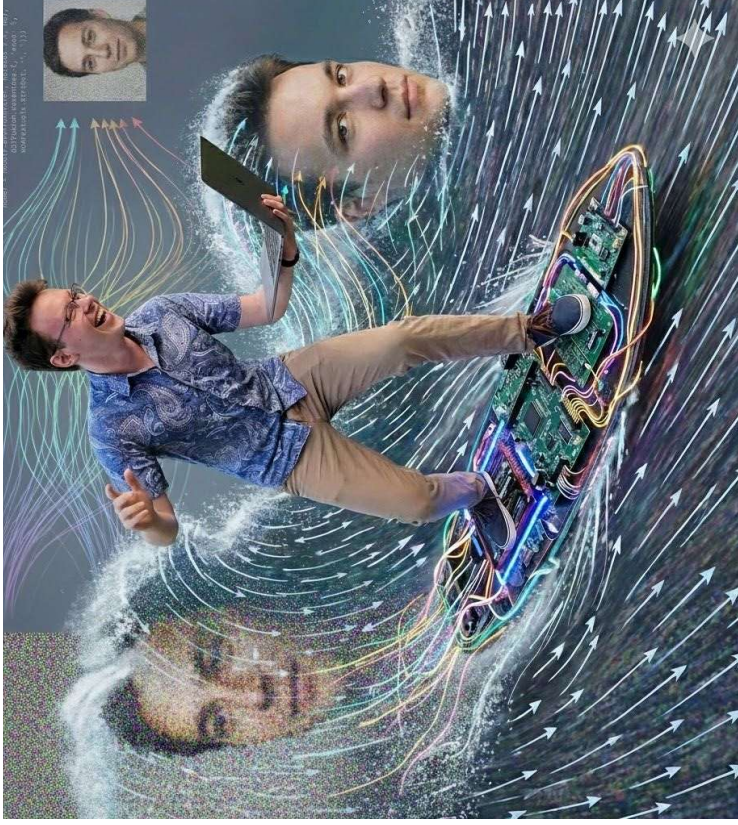
Flow Matching II

# Agenda

Will post discussion + TA evals form soon™. Please fill it out :)

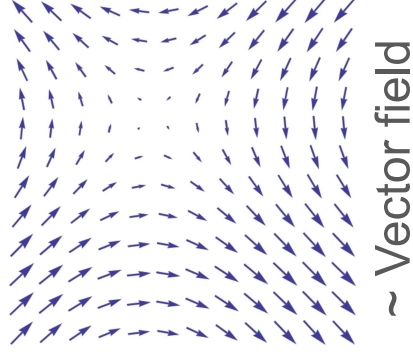
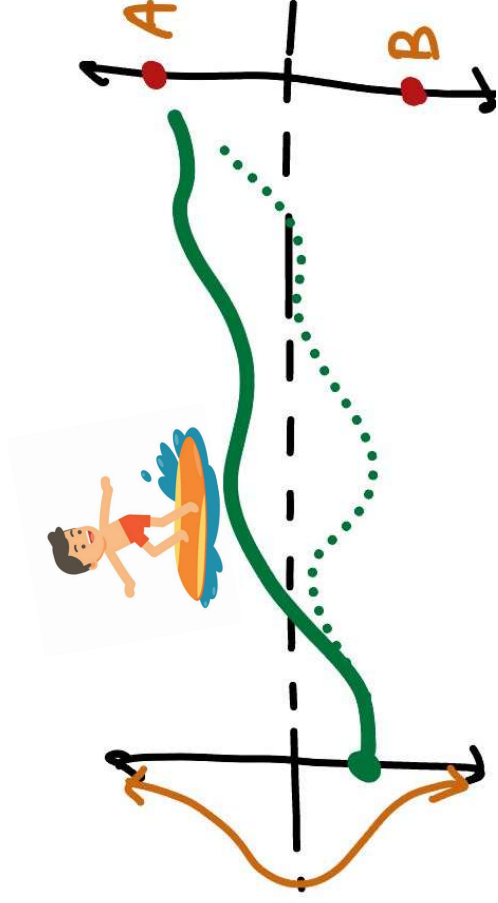
Last short discussion!

1. Flow matching terminology review
2. Sampling
3. AMA if time



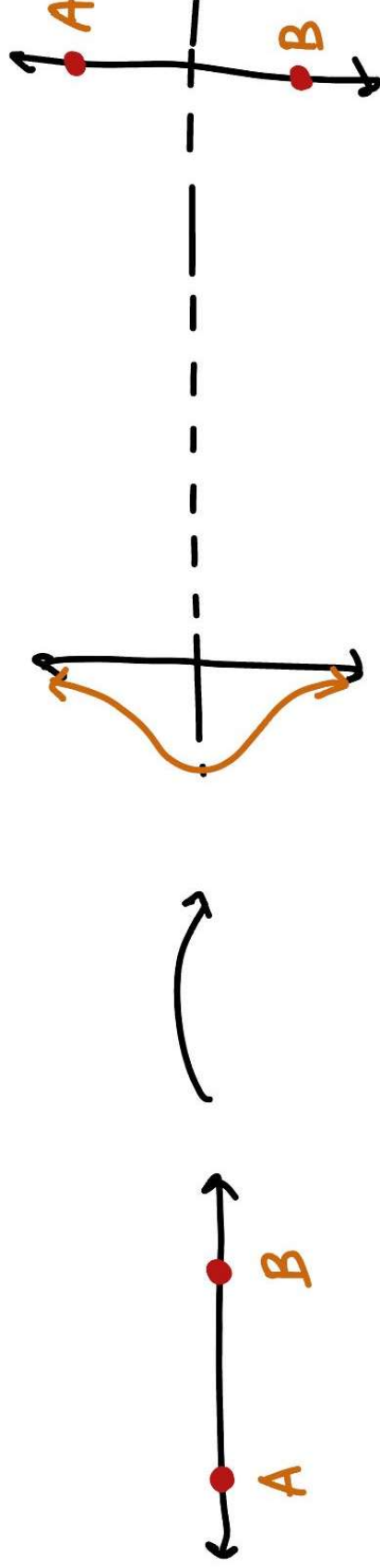
# Flow matching. Geometrically:

- Trying to go from one distribution to another, by following the flow
- Flow matching model supervises this flow



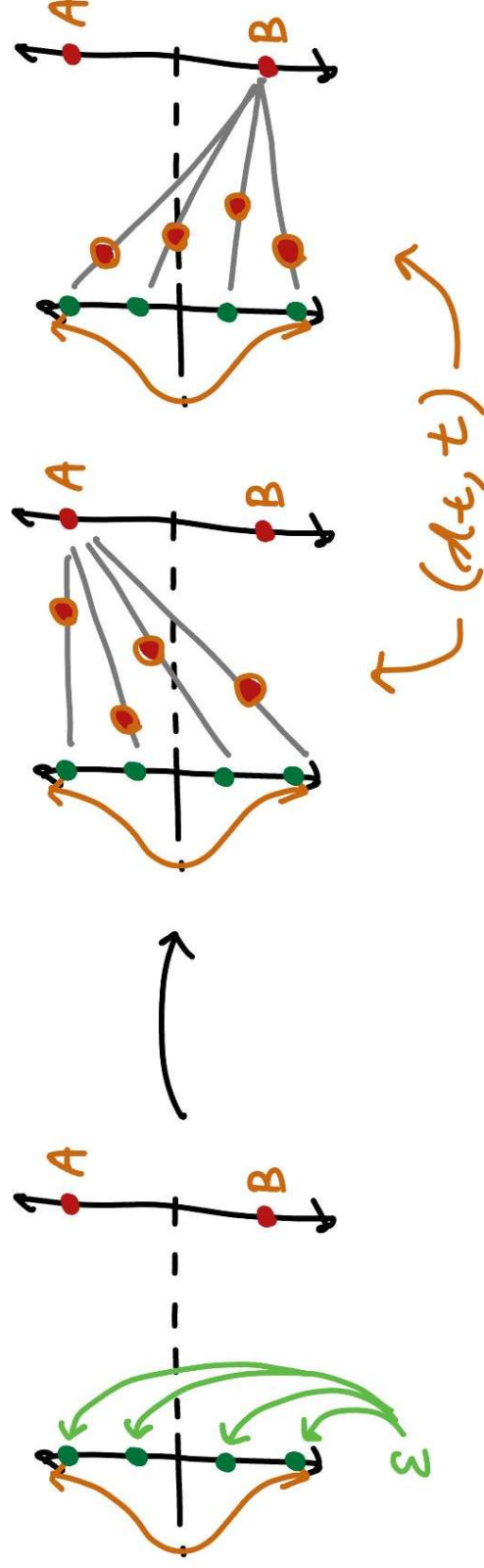
## Flow matching. Recall our 1D distribution:

- **x: (clean or noised) data points. Specify the timestep!**
  - For FM: noise is  $t=0$ , clean data is  $t=1$
- Last time, we modeled a 1D distribution
- The axes here are for: (1) data, and (2) diffusion timesteps



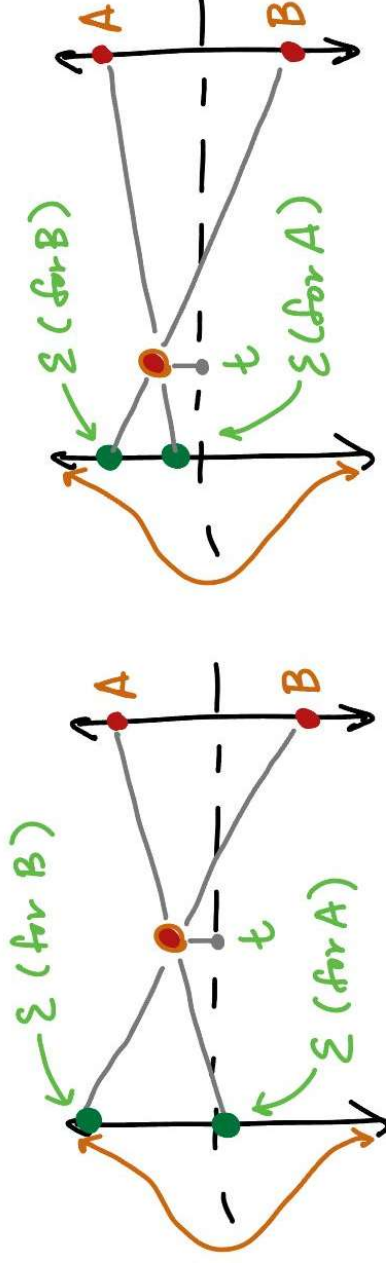
# Flow matching. Generating noisy samples

- **epsilon: noise**
- Sample (epsilon, t), forward noising process from  $x_1$  to get  $(x_t, t)$
- Why linear? Using flow matching model,  $x_t = (1 - t)\epsilon + (t)x_1$



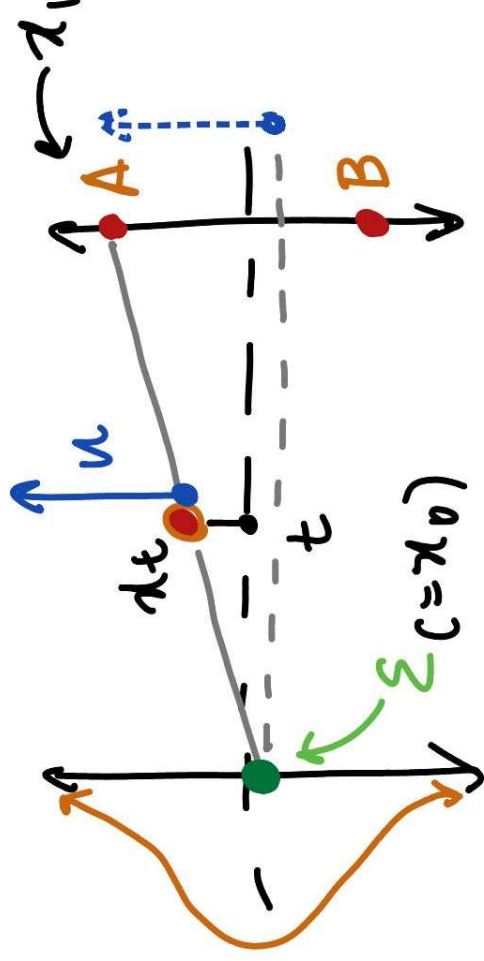
Side note:  $(x_t, t)$  can be sampled from multiple points!

- Just a matter of which epsilon sampled
- Each epsilon has a different likelihood, though (normal distribution)



## Flow matching. velocity

- **u: velocity [towards the clean data sample]**  $u = x_1 - \epsilon$
- Note the magnitude! ( $\sim$  velocity != displacement, ...)



# Flow matching. Term overview

- $x$ : data
  - $x_0$ : fully noised data
  - $x_1$ : clean data (e.g., dataset)
  - $x_t$ : somewhere in between
- epsilon: noise
- $u$ : velocity [towards the data]

$$\begin{aligned}u &= x_1 - \epsilon \\x_t &= (1 - t)\epsilon + (t)x_1 \\&= x_1 - (1 - t)u\end{aligned}$$

$u(x_t, t)$  or  $\epsilon(x_t, t)$  if GT ...  $u_\theta(x_t, t)$  or  $\epsilon_\theta(x_t, t)$  if predicted from network.

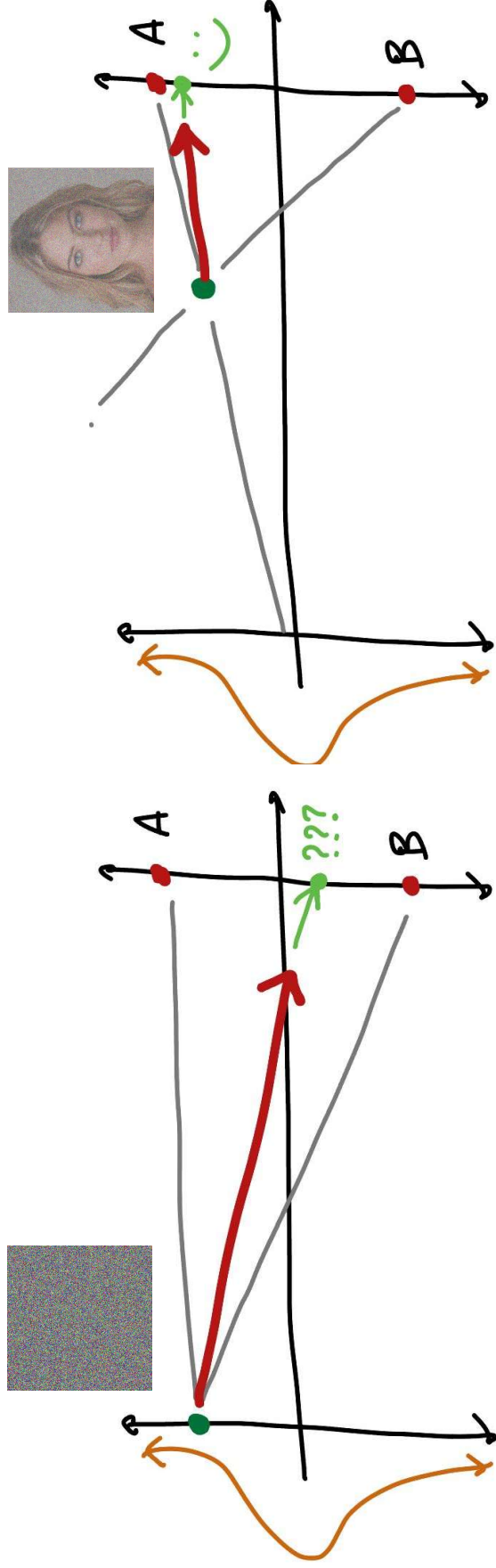


# Sampling. Let's briefly talk about training:

- Model can predict either:  $u$  or  $\epsilon$ 
  - Possible to solve for  $x_1$  given  $(x_t, t)$  and  $u$  OR  $\epsilon$ 
    - Equations:
$$\hat{x}_1 = x_t + (1 - t)u_\theta(x_t, t) \quad \hat{x}_1 = \frac{1}{t}(x_t + (1 - t)\epsilon_\theta(x_t, t))$$
  - Do you see where things could be unstable?
- Training target: point back towards the data sample

# Sampling. Why sample?

- Why? Predictions at small  $t$  are more ambiguous.



# Sampling.

- Many options. One of them is euler integration.

$$x_{t+\Delta t} = x_t + \Delta t \cdot u_{\theta}(x_t, t)$$

- What if your network predicts epsilon?

→ Q2!

$Q1 + Q2 :-)$