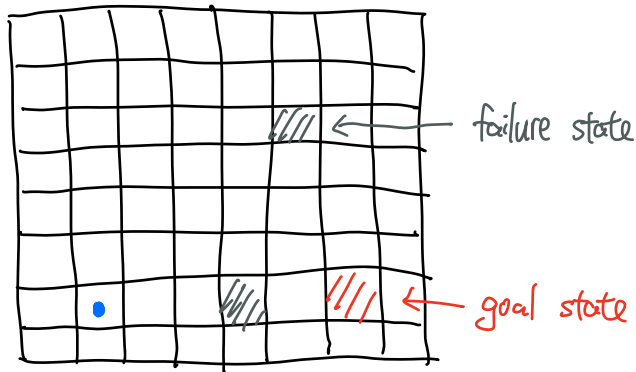


## Review

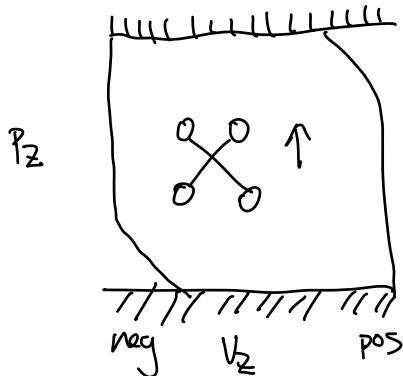
	continuous	Discrete
state	$x(t) \in \mathbb{R}^n$	$X_t$
control	$u(t) \in \mathbb{R}^m$	$U_t$
dynamics	$\dot{x}(t) = f_c(x(t), u(t))$    $\frac{dx}{dt}$	$X_{t+1} = f_D(X_t, U_t)$
trajectory	$X_{x_0, t_0}^u(t)$	$X_{x_0, x_0}^u(t)$ <span style="color: blue;">← discrete timestep</span>

## Grid world



## Reasons for Safety

#1 inevitable collision



$\exists$  states from which collision is inevitable despite your best control

# # uncertainty

Even though we have a mathematical model of the system, it will never capture the behavior of the actual system.

our simple model

$$\dot{P}_z = V_z$$

$$\dot{V}_z = g + k_T u + k_\Phi$$

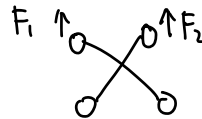
advanced model

$$\dot{P}_z = V_z$$

$$\dot{V}_z = g + k \cos \Phi (F_1, F_2)$$

$$\dot{\Phi} = \omega$$

$$\dot{\omega} = k (F_1 - F_2)$$



This is still not sufficient!

sophistication of model  $\rightarrow$   $\leftarrow$  tractability of analysis

## (model) uncertainty representation

### probabilistic

uncertainty is modeled as a distribution

\* In discrete time, we have

$$x_{t+1} = f(x_t, u_t, d_t) \quad \begin{matrix} \text{r.v.} \\ \uparrow \\ \text{disturbance} \end{matrix}$$

$x_{t+1}$  is a random variable where

$$x_{t+1} \sim P(x_{t+1} | x_t, u_t)$$



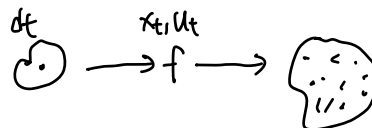
### non-deterministic

uncertainty belongs to a set

$$d_t \in E_t \quad \leftarrow \text{set in } \mathbb{R}^m$$

\* In discrete time, we have

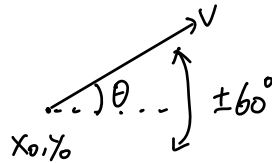
$$x_{t+1} = f(x_t, u_t, d_t) \quad \leftarrow \text{some set (not a point)}$$



$x_{t+1|t}$   $x_{t|t}$

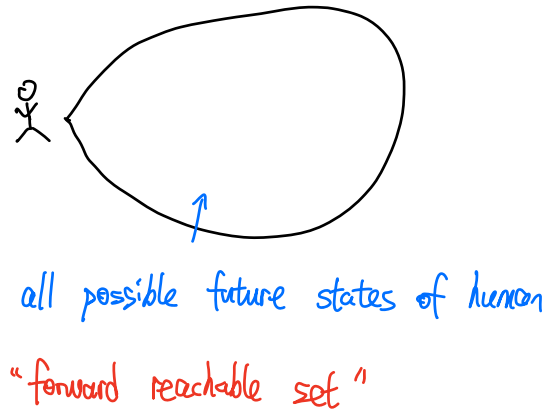
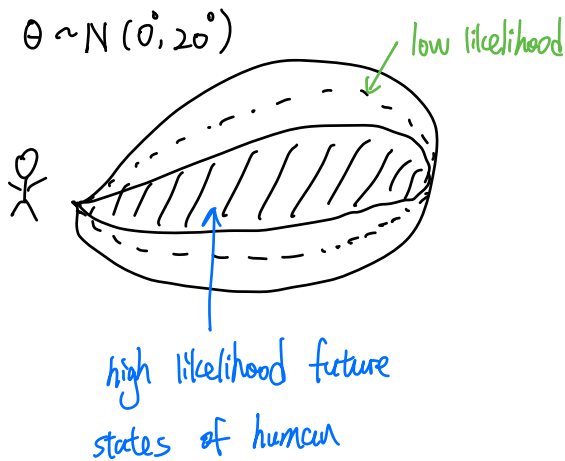
Example: human-robot interaction

$$\begin{bmatrix} \dot{p}_x \\ \dot{p}_y \end{bmatrix} = \begin{bmatrix} v \cos \theta \\ v \sin \theta \end{bmatrix}$$



probabilistic

non-deterministic



1. distribution of uncertainty might not be correct
2. distribution propagation is challenging, especially when the distribution becomes multi-modal  $\rightarrow$  mixture models

1. sets can quickly grow in size resulting in very conservative plans
  - $\hookrightarrow$  closed-loop policies are important
2. no belief update over uncertainty
3. set propagation is very challenging
  - $\hookrightarrow$  level-set methods

