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Hidden Markov Models★ Reasoning over Time and Space

⇒ Often, we want to reason about a sequence of observations

{ Robot Localization }

⇒ Need to introduce time (or space) into our models.

⇒ Value of  $X$  at a given time is called the State.



$P(X_1)$

$P(X_t | X_{t-1})$

{ Initial state probability }

Transition probabilities  
on  
dynamics

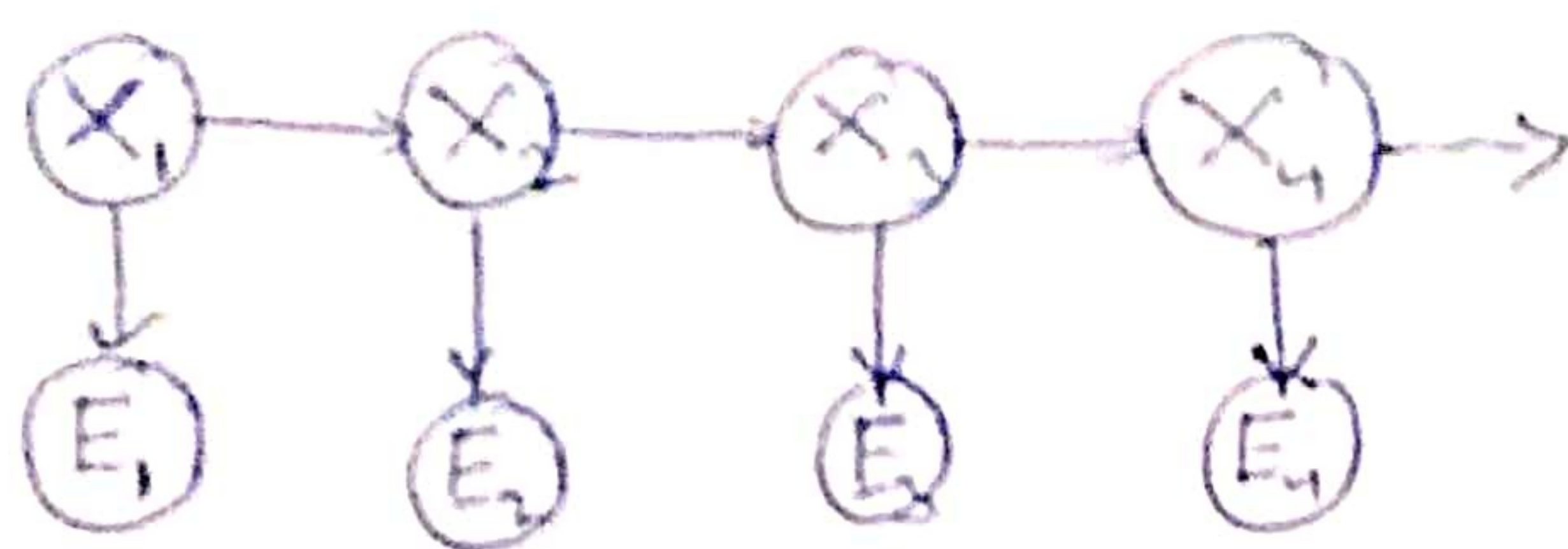
{ Specify how the state  
evolves over time }

⇒ Stationarity Assumption: Transition probabilities stay the same all the time.

⇒ Note that the chain is just a (growable) BN.



## \* Hidden Markov Models (HMMs)



## \* Filtering / Monitoring

$\Rightarrow$  Filtering or monitoring, is the task of tracking the distribution

$$B_t(x) = P_t(x_t | e_1, \dots, e_t) \quad (\text{The belief state})$$

over time

$\Rightarrow$  We start with  $B_1(x)$  in a initial setting, usually uniform

$\Rightarrow$  As time passes, as we get observations, we update  $B(x)$

$\Rightarrow$  The Kalman filter was invented in the 60s and first implemented as a method of trajectory estimation for the Apollo program.

