

Uninformed Search
todo

Heuristic Search
todo

Game Search
todo

Constraint Satisfaction
todo

Logic
todo

Probability & Uncertainty, Bayesian Networks
todo

Intro to ML, Linear Regression, kNN
todo

Decision Trees and Neural Networks
todo

Reinforcement Learning

Markov Property - Future is independent of past given present:
 $\mathbb{P}[S_{t+1}|S_t] = \mathbb{P}[S_{t+1}|S_1, \dots, S_t]$ where S_t is state at time t

We use matrix \mathcal{P} to define transition property from state s to s' , denoted as probability in row s , column s' .

$$\mathcal{P} = \begin{bmatrix} \mathcal{P}_{11} & \dots & \mathcal{P}_{1n} \\ \vdots & & \vdots \\ \mathcal{P}_{n1} & \dots & \mathcal{P}_{nn} \end{bmatrix}, \mathcal{P}_{ss'} = \mathbb{P}[S_{t+1} = s' | S_t = s]$$

Markov Process/Chain - Sequence of states S_1, S_2, \dots satisfying Markov property. Formally defined as tuple $\langle S, \mathcal{P} \rangle$ i.e. (set of states, prob matrix)

Episode - Some sequence of traversed states in a MP

Markov Reward Process - Give states in MP some reward. We “gain” reward R_{t+1} when transitioning from states $S_t \rightarrow S_{t+1}$

placeholder intermediary stuff goes here

	Evaluate Policy, π	Find Best Policy, π^*
MDP Known (Planning probs)	Policy Evaluation	Policy/Value Iteration
MDP Unknown	MC and TD Learning	Q-Learning