Lec-4, 17567, 24.75

Agent k environment interact at each of a seq. of discrete time steps, t = 0, 1, 2, 3...

At time t, agent in a state S_t , takes an action $A_t \in A(S_t)$ $A(S_t) \rightarrow S_t$ possible actions, on agent can take in State S_t . $S_t = S_t \subseteq S_t$

We can say Alst) = Als)

 $S_0 = S_1$ $S_4 = S_1$ actim set is the $S_1 = S_3$ same in all states.

 $S_2 = S_4$ $S_3 = S_2$ t=01234 $S_1S_3S_4S_2S_1$ $A(s_1), A(s_2), A(s_3)$ $A(s_4), A(s_5)$

5 = {s,,s2,s3 s44

Agent receives reward, $R_{+1} \in \mathbb{R} \in \mathbb{R}$ \mathcal{L} finides itself in a new state S_{+1} . $\mathcal{R} = \{1,-1,0.5,7.5...\}$ Finite MOPs:- The sets S, A &R all have a finite number of elements. now, Rt &St are random variables belonging to the set tence we define $b(s', 2|s,a) \triangleq P_{k}\{S_{t}=s', P_{t}=2|S_{t-1}=2, A_{t-1}=a\}$

 $\forall s', s \in S \quad \text{for all } a \in A(s) = A$ The function $\beta(..., ...)$ defines the dynamics of the MDP.

p(s', r|s,a) > Rvs. R+ es+ have well defined discrete prob. distributions dependent only on the preceeding state & action - Markov prob.

may, $\sum_{s'\in S} \sum_{r\in R} p(s',r|s,a) = 1 \quad \forall \quad s\in S, a\in A(s)$

: Conditional prob distribution is also a prob distribution

We can compute state-transition probabilities p(s'|s,a) = Psof st = s'|st-1 = s, At-1 = a} = gwen p(s',s|s,a) = Z p(s',s|s,a) p(s',s|s,a) = 2ER [0,1]E[XIY=y] x p(x/y) $\mathcal{L}(S,a) \triangleq E[Rt|St-1=S,At-1=a]$ = Exp(s|s,a)
rep = Tp(s',r|s,a)
s'es

expected reward for state-action pairs as a two argument function s: SXA - R

$$r(s,a,s) \triangleq E\left[Rt \mid St-1=s, At-1=a, St=s'\right]$$

$$\sum_{EGR} r \cdot \frac{p(s',x|s,a)}{p(s')s,a}$$
expected rewards for state-action-next state triples as a 3-argument function
$$r \cdot S \times A \times S \rightarrow R$$

$$\Rightarrow (x|y,y,y_3) \stackrel{?}{=} p(x,y|y,y_3)$$

$$\Rightarrow (x|y) - \frac{p(x,y)}{p(y,y)} = Use Bayes' theorem.$$