Artificial Neural Networks Markov Decision Processes

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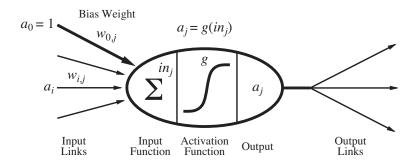
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Outline

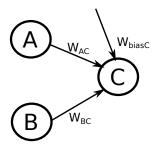
1 Artificial Neural Networks

2 MDP and Learning



Perceptron Design

• Implement $X \vee Y$ using a single perceptron unit. Assume bias input -1.



W_{AC}	1
W_{BC}	1
$W_{ m bias,C}$	0.5

Perceptron Network Design

• Imlement $X \oplus Y$ using a preceptron network.

$\overline{ m W_{AC}}$	1
$ m W_{AD}$	-1
W_{BC}	-1
$\mathrm{W_{BD}}$	1
$W_{ m bias,C}$	0.5
W_{CE}	1
$W_{ m bias,D}$	0.5
W_{DE}	1
$W_{\mathrm{bias,E}}$	0.5

Perceptron Network Design

• Suppose that after training for a long time, the weights of a 3 input (A, B, C) perceptron converge to the values Wa = 1, Wb = 0.5, Wc = 0.5, W0 = -0.75 (bias input is 1). Approximately what Boolean function has the perceptron learned? (Express in logical form as a function of the inputs A, B, and C. You can assume that the inputs take on values of either 0 (false) or 1 (true)). Use step function for activation.

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	Α	В	$^{\rm C}$	W0 (-0.75)	Sum(wx)	G(Sum(wx))
	0	0	0	1	-0.75	0
	0	0	1	1	-0.25	0
	0	1	0	1	-0.25	0
•	0	1	1	1	0.25	1
	1	0	0	1	0.25	1
	1	0	1	1	0.75	1
	1	1	0	1	0.75	1
	1	1	1	1	1.25	1

$$A \vee B \wedge \overline{C}$$

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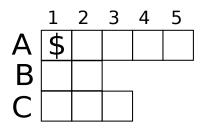
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- Algorithm for solving MDP.
 - Value iteration.

Value Iteration Example



- Actions: forward, back, left, right, idle, pick-up
- Rewards: pick-up treasure: 100, move: -5, idle: none
- Transition: 0.8 move intended, 0.1 move to left, 0.1 move to right
- Formulate the problem as an MDP
- Compute utilities and optimal policy in the first iteration.
- In second iteration.
- Sketch an optimal policy.

Reinforcement Learning

- What if rewards were not known?
- Review TD/Q learning.