



CIT- 315

IA

Artificial Intelligence

Book:

Book - A modern approach - ~~Stuart~~ Stuart J. Russell

Chapters → 1, 2, 7, 8, 16

Lab → python AI → logic → First order logic
→ Predicate logic

AI Project → AI application development

Chapter -

Proprio preuve m'a donné 1.8
similaire - Proprio transfert 1.1

Masud sir

05.05.2024

AI

→ chapter 3

Neural network

Traditional Algorithm vs AI algorithm

↓
train/estimation/
learn \rightarrow all input

data pattern
explore \rightarrow π
learn \rightarrow $\pi \leftarrow \text{astar}(\pi)$

signal \rightarrow process

\rightarrow output \rightarrow signal

\rightarrow IA coding \leftarrow do

We train the algorithm and model

Solving Problem by searching

Introduction —

Goal based agent/problem solving
agent.

Uninformed search algorithm

Informed search algorithm

3.1 Problem solving agent:

1. Intelligent agent - maximise performance

Steps in problem solving using A* algorithm

① Goal Formulation

Observable

② Problem

Known

Fig-3.1

Deterministic \rightarrow

short form \leftarrow output (part 2)

process with IIA

Search

Solution

Execution

07/7

Well defined problem and user
solutions - fire component

1. Initial state

8. Goal Test

2. Action

9. Path cost

3. Applicable

10. Step Cost

4. Transaction model

11. Optimal Solution

5. State space

(1, 2, 3 collectively)

6. Path

↓ decrease total cost initially

7. Goal estate

↓ update to some one

↓ decrease to some one

→ to goal

→ to goal

→ to goal

→ to goal

3.1.1 Uninformed search strategies

blind search →

with more work (I)

without work (II)

3.1.1.1 Breadth first search

1st expand → root node

→ All the successor

→ their successor

FIFO

b^d = branching fact

d = depth

complexity →

$b + b^2 + b^3 + \dots + b^d$ → $O(b^d)$

completeness, optimality, space complexity

3.1.2

Uniform cost search →

no care of steps, care of cost

slower, more work

Total cost c^*
every cost c
steps c^*/c

concept

3.4.3

depth first search \rightarrow not complete, not optimal

$m = \text{maximum-depth}$

$O(b^m) \rightarrow \text{time complexity}$

$O(b \times m) \rightarrow \text{space complexity}$

space complexity ~~varies~~, ~~for~~ ~~depth~~ $\leq m$ ~~problems~~

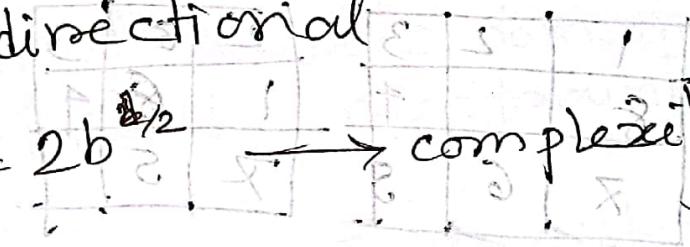
3.4.4 \rightarrow depth limited search.

3.4.5

3.4.6 \rightarrow Bidirectional

$$b^{d/2} + b^{d/2} = 2b^{d/2} \rightarrow \text{complexity} \approx$$

$$2b^{d/2} \leq b^d$$

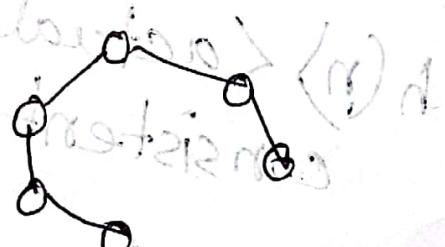


A* search \rightarrow

Heuristic function and its applications

heuristic function

$h(n) = \text{estimate cost of the cheapest path from node } n \text{ to goal node}$



$$\begin{aligned}
 h(n) &= g(n) \\
 f(n) &= g(n) + h(n) \\
 &= h(n) + h_2(n)
 \end{aligned}$$

Estimated		
7	2	11
5	6	
8	3	11

$h_1 = 8$

$$h_2 = 3 + 1 + 2 + 2 + 2 + 3 + 3 + 2$$

$$\approx 18$$

$h(n) = h_1 + h_2 = 26$

$g(n) = \text{depth of the node}$

1	2	3
8		4
X	6	5

$$h(n) = \text{estimated cost}$$

2	8	9
1	6	4
X	5	

heuristic may be optimal

$h(n) > \text{actual cost}$

solutions can be overlooked

$h(n) = \text{actual cost} - \text{best case}$

$h(n) \leq \text{actual cost} + \text{admissible}$

consistent heuristic

consistent \rightarrow

$$h(n) \leq c(n, a, n) + h(n)$$

If $h(n)$ is consistent then the value of $f(n)$ along any path are non decreasing

Chapter - 4

Local search - Global minimum - lowest valley
Global maximum - highest valley

chap - 6

Q.1.3 definition - discrete domain, finite domain, infinite & constraint language

revise function \rightarrow remove inconsistent value

Chap - 7

02.06.2014

Mehlbaum sin

Java point \rightarrow knowledge represented in form of
knowledge based adjacent AT.

knowledge based knowledge system
knowledge based generic knowledge

Inference system

Operations -

Approaches -

knowledge based

What is knowledge representation?
what to represent

Type of knowledge

knowledge cycle

propositional logic

Artificial intelligence with

Syntax

logical connectives

presidues of operations.

Inference
types

13.2.1 Probability - Unconditional, Prior
conditional \rightarrow $P(C|T) / P(C \wedge T)$
Joint

$$\begin{aligned}
 P(a \wedge b) &= P(a) \cdot P(b) \\
 P(a \wedge b) &= P(b \wedge a) \cdot P(a) \\
 P(b \wedge a) &= P(b|a) \cdot P(a) \\
 \Rightarrow P(a \wedge b) P(b) &= P(b|a) P(a) \rightarrow \text{biggest rules.} \\
 \Rightarrow P(a \wedge b) &= \frac{P(b|a) P(a)}{P(b)}
 \end{aligned}$$

Fig 13.3 \rightarrow ***

Malibub sir
Javaatpoint
Propositional logic - ontology, condition,
Proposition - assertive sentence

Atomic proposition - with example

Compound
Logical connectiveness - Negation, conjunction,
Truth tables
Precedence
Logical equivalence
Properties of operation
Limitation
Types of inference rules -

13.1

Independence - Full joint distribution $P(\text{Toothache, Catch, Cavity, weather})$ cloudy

Product rule : $P(A, B) = P(A|B) \cdot P(B)$

$$P(\text{toothache, catch, cavity, cloudy}) = P(\text{cloudy})$$

$$\cdot \underbrace{P(\text{toothache, catch, cavity})}_{\substack{\text{because independent} \\ \text{relation}}} \cdot P(\text{toothache, catch, cavity})$$

\cdot $\underbrace{P(\text{cloudy})}_{\substack{\text{because independent} \\ \text{relation}}} \cdot P(\text{toothache, catch, cavity})$

$$= P(\text{cloudy}) \cdot P(\text{toothache, catch, cavity})$$

$$P(a|b) = P(a) \quad P(b|a) = P(b) \rightarrow \text{if independent}$$

$$P(a \wedge b) = P(a) \cdot P(b)$$

variable - 2

Bayes Rule - $P(b|a) = \frac{P(a|b) \cdot P(b)}{P(a)}$

$$P(Y|X, e) = \frac{P(X|Y, e) \cdot P(Y|e)}{P(X|e)}$$

Using Bayes Rule causing evidence -

$$P(\text{cavity} | \text{toothache} \wedge \text{catch}) = \alpha P(\text{toothache} \wedge \text{catch} | \text{cavity}) \cdot P(\text{cavity})$$

$$P(\text{toothache} \wedge \text{catch} | \text{cavity}) = P(\text{toothache} | \text{cavity}) \cdot P(\text{catch} | \text{cavity})$$

conditional

$$P(\text{cavity} | \text{toothache} \wedge \text{catch}) = P(\text{toothache} | \text{cavity}) \cdot P(\text{cavity})$$

$$\cdot P(\text{catch} | \text{cavity}) \cdot P(\text{cavity})$$

$$P(y|x_1, x_2, \dots, x_n) = \frac{P(x_1|y) P(x_2|y) \dots P(x_n|y)}{P(x)}$$

$$= P(y) \prod_{i=1}^n P(x_i|y)$$

$$P(y) = \text{argmax}_y P(y) \prod_{i=1}^n P(x_i|y)$$

Mahbub Sir

01.09.24

Java point

Knowledge base, Wumpus world

Wumpus moveable

Lab implement \rightarrow Wumpus, Scream, Stench, Breeze, Glitter, Bump, Scratch

02.09.24

Mahbub sir

Wumpus World \rightarrow Java point

Performance

Environment

actuators

sensors

knowledge base wumpus world \rightarrow Java point

4x4 - 16 logic equation

7x9x4 - proposition variable

Table

IC10 for 27/07/2021

Theory to choose direct

09.08

Mahbub Sir

Microprocessor lab \rightarrow report \rightarrow assignment

AI \rightarrow \rightarrow \rightarrow

AI Theory \rightarrow AI related \rightarrow \rightarrow

Microprocessor Theory \rightarrow Scan \rightarrow assignment

Microprocessor \rightarrow Text \rightarrow (পর্যাপ্ত করা
 \rightarrow পরিপূর্ণ প্রশ্ন)

Machine \rightarrow Flag

AI \rightarrow 2 (S10)

\rightarrow 1st order logic + chapter 2

\rightarrow the propositional logic + chapter 1

First order logic \rightarrow

Modern Intelligence \rightarrow logic \rightarrow 1st order

\rightarrow propositional

\forall objects, properties, relations, functions.
individual identity \downarrow distinguish
 \rightarrow between objects

User provides — constant symbol, function symbol

Javatpoint \rightarrow 1st order logic

Syntax

\forall \rightarrow reverseA \rightarrow All \rightarrow universal quantifier

\exists \rightarrow \rightarrow E \rightarrow Existential

Atomic sentence

Complex sentence

Subject, predicates

Quantifiers

Universal Q

Existential Q

Free and bound variable \rightarrow গুরুত্ব করা হবে

Properties of Q

৫ টি sentence রিপ্রেজেন্ট করা হবে



18.1 Ev 10.09.24 Masud Sir

18.1 Evaluating and choosing best hypothesis

[ID] ? এখন independent এবং Gaze assumption
গুরুত্ব করা সমান এবং

Hold out cross-validation

K-fold cross-validation

LOOCV $K=n$

18.1.1 model selection : two task -

- ① model selection $H \}$
- ② Find the best hypothesis
- optimisation

Error Rate of a hypothesis as a problem
the proportion of mistakes it makes \oplus -

$$h(x) \neq y$$

fig 18.9

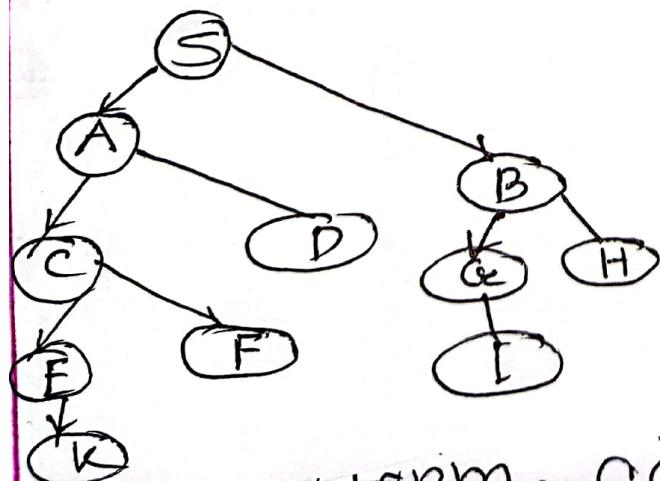
\Rightarrow node রিপ্রেজেন্ট করা হবে এবং optimise

BFS -

- provide solution if any solution exists.
- helps to find the shortest path in goal state since it needs all nodes at ~~same hierarchical~~ ⁿ hierarchical ~~no~~ level before making a move to nodes at lower level.
- If more than one solution exists, then it will provide minimal solution which requires least number of steps

Disadvantages - Lots of memory

- Lots of time (if far away from root)
- inefficient for deeply layered spaces.



$S \rightarrow A \rightarrow B \rightarrow C \rightarrow D \rightarrow G \rightarrow H \rightarrow E \rightarrow F \rightarrow I \rightarrow K$

minimum cost search

Uniformed search strategies -

BFS

- All nodes are expanded in a given depth in search tree before any nodes are expanded at the next level expanded.
- This is achieved by FIFO queue for the frontier.
- Not good for time and space
- Total numbers of nodes generated -
$$b + b^2 + b^3 + \dots + b^d = O(b^d)$$

~~if expands shallower node~~

~~- optimal if path cost is non-decreasing function of the depth of the node~~

Uniform-cost search

- uniform-cost search expands the node n with the lowest path cost $g(n)$
- store frontier as a priority queue ordered by $g(n)$
- Uniform cost search doesn't care about number of steps a path has, but only about the total cost. Therefore it will stuck in infinite loop if there is a path with an infinite sequence of zero cost action.

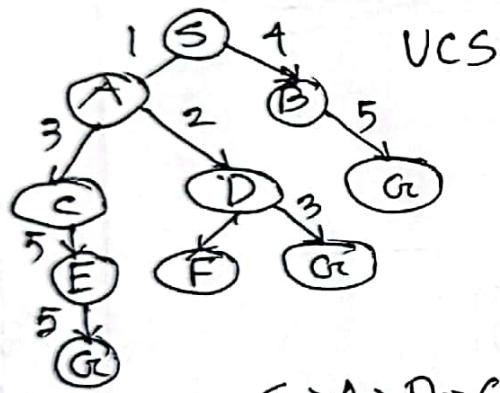
— Optimal because at every state the path with least cost is chosen

— C^* = cost of optimal solution

worst case time and space complexity is $O(b^{1 + [C^*/e]})$

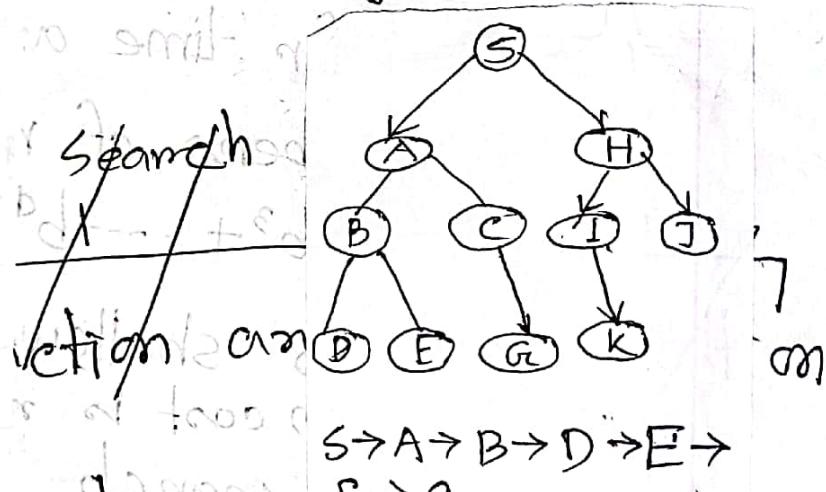
— If all steps are equal $O(b^{1 + [C^*/e]}) = b^{d+1}$

— The goal test is applied to a node when it is selected for expansion rather than when it is first generated.



$S \rightarrow A \rightarrow D \rightarrow G$

Depth-limited search



$S \rightarrow A \rightarrow B \rightarrow D \rightarrow E \rightarrow G$

DFS

Depth-first-search

— expand deepest node

— Use LIFO queue

— variant of depth-search called backtracking search.

— Once node has been expanded, it can be removed from memory as soon as all its descendants have been fully explored. So it uses less memory.

- Combines DFS and BFS benefits in terms of first search and memory efficiency.
 - guarantee to find optimal solution as long as the cost of each edge in the search space is the same
 - Repeats all works of previous steps - drawback
- Unidirectional search
- Replace one single search graph with two small subgraphs in one start from initial vertex and another start from goal vertex and stops when two intersect.
 - It is fast, less memory, helpful for large graph that has no way to make smaller.
 - We should know the goal state in advance
-
- Intersection node.

- There is the possibility that many steps states —
keep re-occurring and there is no guarantee
of finding the solution.

Depth - limited search - It solves infinite path problem.

Depth = λ max
Time complexity $O(b^{\lambda})$
Space complexity $O(b^{\lambda})$

- Space
- Two type failure - standard failure - no solution
- cut off failure - no solution
- It is not optimal even if in depth limit

Iterative deepening depth-first search increases the depth limit

It gradually is found.
$$S_n = (d)b + (d-1)b^2 + \dots + (1)b^d$$

— Worst case the search space \rightarrow is preferred when depth is unknown forward = of

Bi-directional search algorithm to find both initial state and other

— One forward from the goal with drop = 100% — increase complexity

Time complexity = $O(p+q)$ Space = $O(p+q)$

* The graph search version which avoids repeated state and redundant paths

Criterium	Breadth First	Uniform Cost	Depth First	Depth Limited	Iterative	Bidirectional
Complete?	Yes ^a	Yes ^{a,b}	No	No	Yes	Yes ^{a,b}
Time	$O(b^d)$	$O(b^{1+\lceil C^*/\epsilon \rceil})$	$O(b^m)$	$O(b^l)$	$O(b^d)$	$O(b^{d/2})$
Space	$O(b^d)$	$O(b^{1+\lceil C^*/\epsilon \rceil})$	$O(bm)$	$O(bl)$	$O(bd)$	$O(b^{d/2})$
Optimal	Yes ^c $d/(1-b) + d(b)$	Yes ^{c,d}	No	No ^c	Yes ^c	Yes ^{c,d}

b = branching factors

d = depth of shallowest solution

m = maximum depth of search tree

l = depth limit

a = complete if b is finite

b = complete if step costs $\sum \epsilon$ for positive ϵ

c = optimal if step case are all identical

d = if both direction use breadth first search

A* search

$$f(n) = g(n) + h(n)$$

evaluation function

Heuristic function: The cost to reach the node from start node + estimated cost to get from the node to the goal.

Heuristic Function: A heuristic function (algorithm) or simply a heuristic, is a shortcut to solving a problem when there are no exact solutions for it or time the time to obtain the solution is too long.

Heuristic properties -

1. Admissible condition: If an algorithm produces an optimal result, it is considered admissible.
2. Completeness - An algorithm ends with a solution.
3. Dominance Property - If A_1 and A_2 are two heuristic functions, algorithm and have h_1 and h_2 as function respectively; then A_1 will dominate A_2 if h_1 is superior to h_2 for all possible value of node n .
4. Optimality Property: An algorithm will thro thorough, allowable, and dominates the other one algorithm.

Heuristic Function & $h(n)$:

$h(n)$ = Estimate cost of the (n) ⁽ⁿ⁾ cheapest path

from node n to goal node

of total cost of node

of node to goal node

of node to goal node

$$f(n) = g(n) + h(n)$$

$g(n)$ = depth of node

$h(n)$ = number of misplaced tiles

number of tiles not in their correct position

How to make heuristic Function

7	2	1
5	4	6
8	3	1

-	1	2
3	4	5
6	7	8

Start node

Goal node

$$h_1 = 8$$

$$h_2 = 3 + 2 + 2 + 2 + 3 + 3 + 2 + 1$$

$$h(n) = h_1 + h_2 = 8 + 18 = 26$$

number of tiles not in their correct position

Final

1	2	3
8		4
7	6	5

Initial

2	8	3
1	6	4
7		5

A*

2	8	3
1	6	4
7	5	

2	8	3
1		4
7	6	5

2	8	3
1	6	4
7	5	

2	8	3
1	4	
7	6	5

2	8	3
1	8	4
7	6	5

2	8	3
1	4	
7	6	5

2	8	3
1	4	
7	6	5

2	8	3
2	4	
7	6	5

2	3	
1	8	1
7	6	5

1	2	3
8		4
7	6	5

1	2	3
8		4
7	6	5

1	2	3
7	8	1
	6	5

Final state

Optimality of A* search

(1) $h(n) >$ actual cost

→ Optimality Optimum solution can be overlooked, optimum solution is not possible

(2) $h(n) =$ actual cost

→ Best case scenario: If $h(n)$ approximates actual cost, searching uses minimum of node of the goal.

(3) $h(n) <$ actual cost

→ Admissible, consistent heuristics

admissible \rightarrow consistent এবং ফিক্স

consistent \rightarrow admissible নির্মাণ

প্র

$t = 0$	E	S	I
$t = 1$	E	S	I
$t = 2$	E	S	I

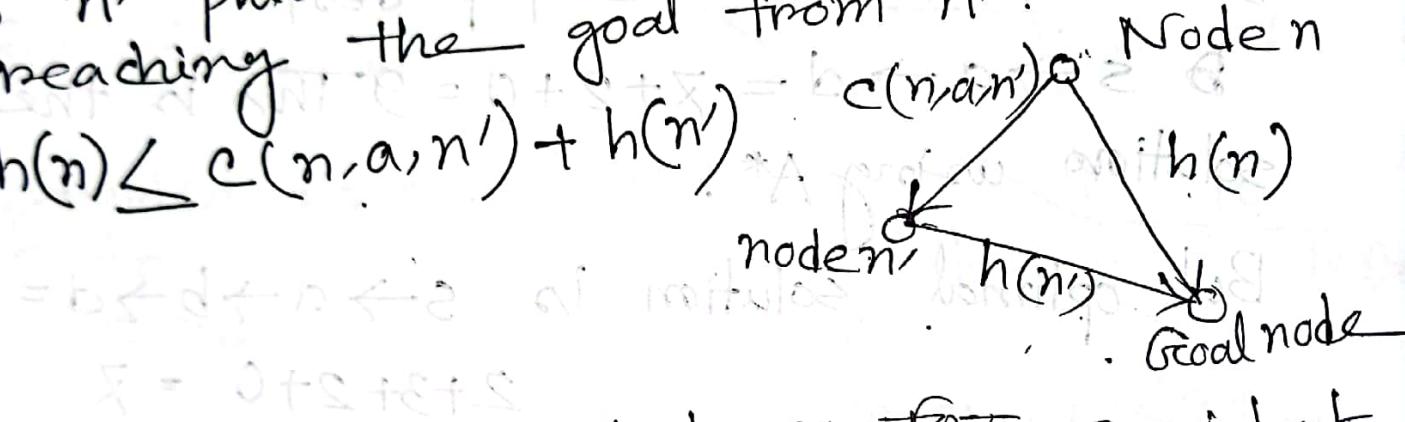
$t = 0$	E	S	I
$t = 1$	E	S	I
$t = 2$	E	S	I

$t = 0$	E	S	I
$t = 1$	E	S	I
$t = 2$	E	S	I

Consistent heuristic —

A heuristic $h(n)$ is consistent (or monotonic) if, for every node n and every successor n' of n generated by any action a , the estimated cost of reaching the goal from n is no longer greater than the step cost of getting to n' plus the estimated cost of reaching the goal from n' .

$$h(n) \leq c(n, a, n') + h(n')$$



admissible \overline{h} consistent \overline{h} , ~~not~~ consistent

\overline{h} admissible $\overline{h} \leq \overline{h}$

Admissible heuristic → An admissible heuristic is that never overestimates the cost to reach the goal.

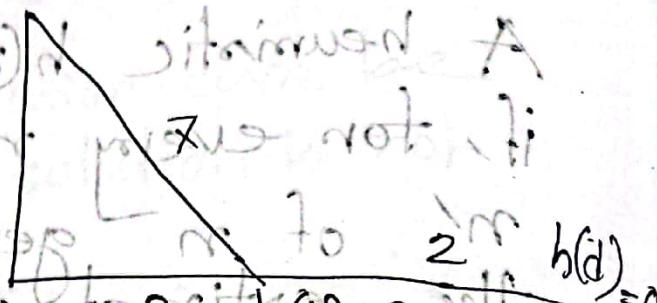
$$h(n) \leq \text{actual cost.}$$

$h(n) > \text{actual cost}$ ~~if $h(s) = 10$ first choice~~

From start

$$s \rightarrow a = 2 + 9 = 11$$

$$s \rightarrow b = 2 + 3 = 10 \quad \text{so } h(a) = 9, h(b) = 3$$



So, we pop b, then explore neighbours a and d.

But a is visited, we pop d

~~$s \rightarrow a \rightarrow d = 2 + 2 + 0 = 4$ This is the solution using A*~~

But optimal solution is $s \rightarrow a \rightarrow b \rightarrow d = 2 + 3 + 2 + 0 = 7$

A* doesn't provide optimal solution because $h(n) > \text{actual cost}$

Heuristic must be admissible and consistent for the optimal solution in the A* search algorithm \rightarrow

We will see an example for that the heuristic is admissible but not consistent.

$$h(s) = 12, \quad h(a) = 11$$

For admissible,

shortest distance from s to dest \rightarrow

$$s \rightarrow a \rightarrow b \rightarrow d = 1 + 1 + 10 = 12$$

For node s, $h(s) = 12 \leq 1 + 1 + 10$ or, $12 \leq 3 + 10$

For node b, $h(b) = 0 \leq 10$

For node a, $h(a) = 11 \leq 1 + 10 = 11$

Therefore this is admissible heuristic

For consistent,

$$h(n) \leq c(n, n') + h(n')$$

We will see

$$s \rightarrow a \quad 12 \leq 1 + 11$$

$$s \rightarrow b \quad 12 \geq 3 + 0 \quad (\text{not satisfied})$$

Therefore heuristic is admissible but not consistent.

If heuristic is admissible but not consistent, it won't provide optimum solution.

If we apply A* algorithm →

Starting node S has two successors,

a and b. S has pushed two values in priority queue $[a = 1 + 1 = 2, b = 3 + 0 = 3]$

We pop b and remained $[a = 12]$

Node b has two successors, a and d. but a has visited already, so we will push the value d. $[a = 1 + 1 = 12, d = 3 + 10 = 13]$

Then pop a and remained $[d = 13]$

$d = 13$ is the solution.

The solution is not optimal. We know here optimal solution is $1 + 1 + 10 = 12$

If heuristic is admissible but not consistent, it may not be provided optimum solution. For optimum solution, Heuristic must be consistent. If it is consistent, it must be admissible.

Most widely used best ~~to~~ first search is A* search. It evaluates nodes by combining $g(n)$, the cost to reach the node and $h(n)$, the cost to get from the node to the goal.

$$f(n) = g(n) + h(n)$$

since $g(n)$ gives the path cost from start node to n , and $h(n)$ is the estimated cost of the cheapest path from n to goal, we have,
 $f(n)$ = estimated cost of the cheapest solution through n .

11.09.24

Maxd sin

18.4.2

From error rates to loss
Error 17. ($\text{spam} \rightarrow \text{nonspam}$) better Error 57.
(nonspam \rightarrow spam)

Expected utility: in means of loss function in ML
Loss function: The loss function $L(x, y, \hat{y})$
is defined as the amount of utility lost by
post predictor $h(x) = \hat{y}$ when the correct
answer is $f(x) = y$

$\text{Log} \rightarrow 1$ loss

theory of learning \rightarrow less important

18.6

18.6.3 fig 2.9 gradient descent
18.6.4 what is structure and optimization
Neural network, backpropagation, algorithm,
backpropagation, algorithm, feed
forward, multilayer, feed

back propagation $\star\star\star$

3 \rightarrow 1
4 \rightarrow 1
6,13 \rightarrow 1
18 \rightarrow 1