

The University of Nottingham  
*Malaysia Campus*

SCHOOL OF COMPUTER SCIENCE

A LEVEL ONE MODULE, SPRING SEMESTER 2018-2019

FUNDAMENTALS OF ARTIFICIAL INTELLIGENCE (COMP 1032)

Time allowed 1.5 hours

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*Candidates must NOT start writing their answers until told to do so*

***Answer ALL questions***

*No calculators are permitted in this examination.*

*Dictionaries are not allowed with one exception. Those whose first language is not English may use a standard translation dictionary to translate between that language and English provided that neither language is the subject of this examination. Subject specific translation dictionaries are not permitted.*

*No electronic devices capable of storing and retrieving text, including electronic dictionaries, may be used.*

***DO NOT turn examination paper over until instructed to do so***

**ADDITIONAL MATERIAL:**      None

**<Proposed Marking Scheme>**

*Annotations Used:*

**K** - Knowledge;  
**C** - Comprehension;  
**A** - Application

## 1. Search (25 marks)

- i. Explain briefly in the context of problem solving that involves searching in a discrete state space the concepts on search space and search tree. **K** [5 marks]

*Answer:*

*The search space is the implicit tree generated from the initial state and the operators.* [2 marks]

*The search tree is the explicit tree generated from a search strategy, which defines the order of state expansion.* [3 marks]

- ii. Consider the following search problem represented by the graph (Figure 1), where S and G are the start and goal states, respectively. The costs associated to arcs connecting the states (vertices) are given in the graph. The following Table 1 lists out three different heuristics ( $h_1$ ,  $h_2$ ,  $h_3$ ), each being a set of heuristic values from the current state to the goal state G. **A, C**

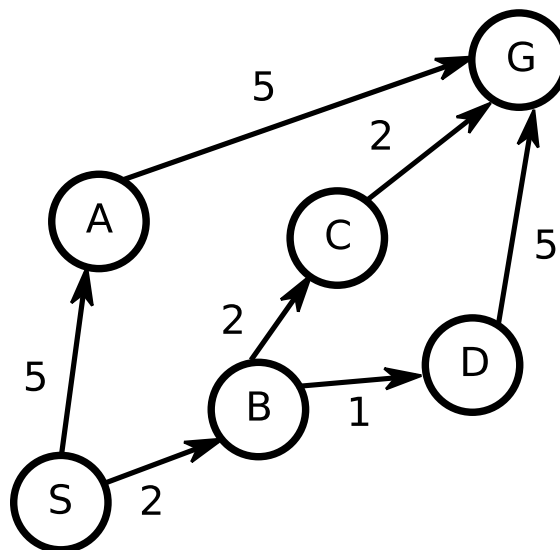


Figure 1

States	$h_1$	$h_2$	$h_3$
S	0	5	6
A	0	3	5
B	0	4	2
C	0	2	5
D	0	5	3
G	0	0	0

Table 1

- a) Greedy and A\* search are two different informed search methods. Explain briefly how A\* search is different from greedy search.

[5 marks]

*Answer:*

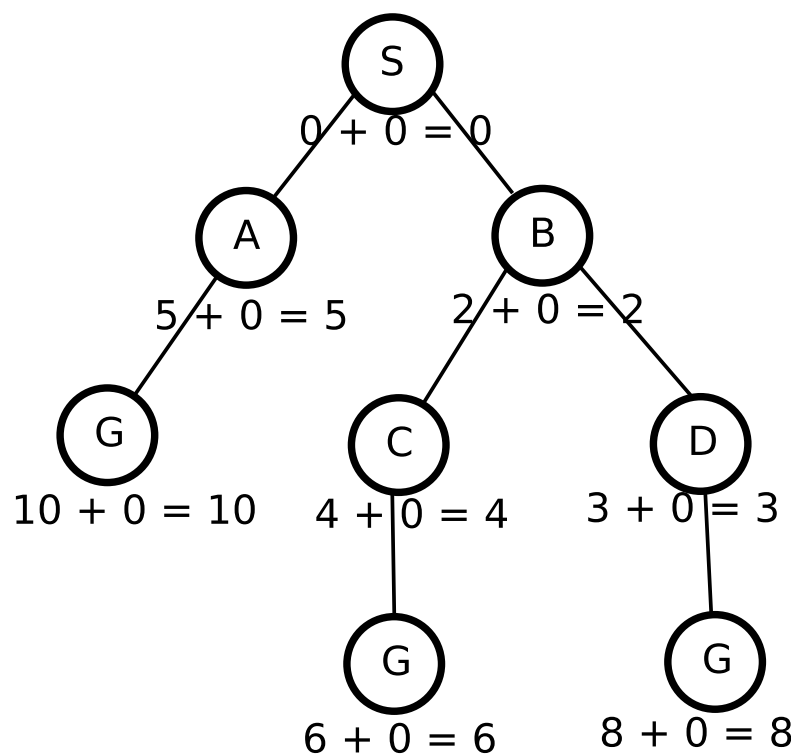
*The difference is mainly with the use of the evaluation function. Greedy search uses only the true cost so far to reach a state  $n$  starting from the initial state,  $g(n)$ . A\* uses (summing) both  $g(n)$  and the estimated cost to the goal state from the current state  $n$ ,  $h(n)$ .*

[5 marks]

- b) Construct the search tree associated to an A\* search, where the evaluation function at state  $n$  is  $f(n)$ , which is a simple sum of the total cost from  $S$  to the current state  $g(n)$  and the heuristic value from the current state to  $G$  given by  $h_1(n)$ . After that, write out the solution path and the total true cost.

[6 marks]

*Answer:*



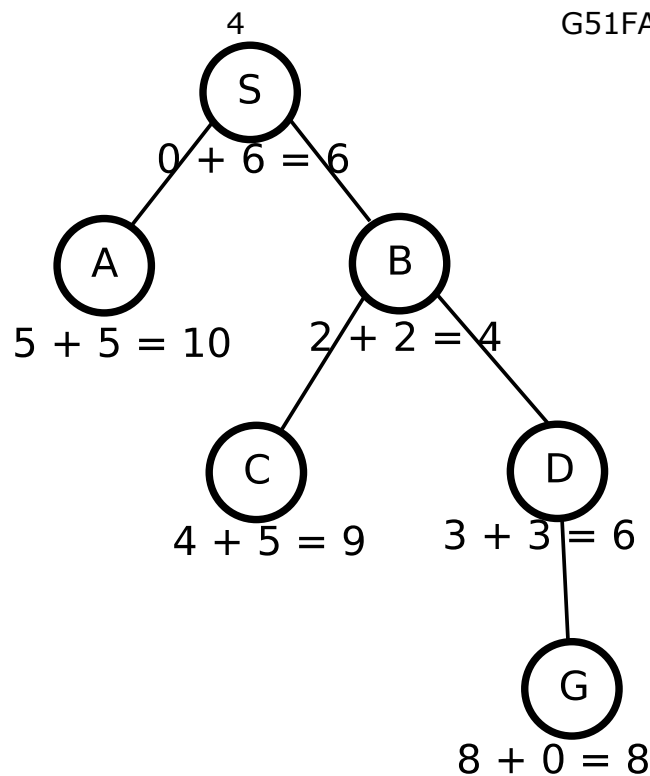
[4 marks]

*S-B-C-G at cost of 6 units.*

- c) Repeat the question above, this time using  $h_3(n)$ .

[6 marks]

*Answer:*



[4 marks]

*S-B-D-G at cost of 8 units.*

[2 marks]

- d) Given your results in (b)-(c), explain briefly if different solutions can be found for using different heuristics, and if so, why.

[5 marks]

*Answer:*

*Different solutions can be obtained. With  $h_3$  a less than optimal solution is found.*

[2 marks]

*Heuristics such as  $h_3$  are not admissible (e.g. overestimate the cost to reach the goal).*

[3 marks]

## 2. Game Playing and Coevolution (25 marks)

- iii. Describe a typical coevolutionary system as a *generate-and-test* procedure. **K, C**

[10 marks]

*Answer:*

*As a general generate-and-test procedure, a coevolutionary system involves*

- 1. Initialize a population of agents (candidate solutions).*
- 2. Evaluate fitness through agent interactions (e.g. pairwise comparisons).*
- 3. Select parents from the populations based on fitness.*

4. Generate offspring from parents in (3).
5. Repeat steps (2-4) until some termination criterion reached.

[2 marks each point]

- iv. Formally explain how Chebyshev's bounds are used for evaluation of performance of strategies produced (evolved) through coevolution (full marks for complete mathematical description). **K, C**

[10 marks]

Answer:

Consider a two-player, win-lose (0-1) game, having a set **S** of  $M$  pure strategies,  $\{1, 2, 3, \dots, M\}$ . The true generalization performance of a strategy  $i$  is

$$G_i = \sum_{j=1}^M P_s(j) G_i(j)$$

where  $P_s(j)$  represents the random selection of  $j \in \mathbf{S}$  as test strategy with some probability,  $G_i(j)$  is the outcome for  $i$  for a game between  $i, j \in \mathbf{S}$ .

[4 marks]

This can be estimated using a random sample  $S_N$  of  $N$  test strategies drawn iid from **S** with probability  $P_s$

$$\hat{G}_i(S_N) = \frac{1}{N} \sum_{j \in S_N} G_i(j)$$

[3 marks]

in which case, one can then apply Chebyshev's Theorem to obtain the following bounds

$$P(|\hat{G}_i - G_i| \geq \varepsilon) \leq \frac{R}{4N\varepsilon^2}$$

for any positive  $\varepsilon$  and  $R = 1$ .

[3 marks]

- v. Consider the case of a game with a win-lose (1-0) outcome. Using an i.i.d. random sample of 25000 test strategies to estimate the generalization performance of a strategy, what statistical claim can you make for a given precision of  $\varepsilon = 0.1$ ? If the variance associated with the game outcomes for the strategy is one-tenth of the worst case, what would be the improved statistical claim you can make?

[5 marks]

Answer:

Greedy search returns the path a-c-f-i-l-m with a total distance of  $200+160+165+250+60=835$  km. A\* search returns the path a-c-f-i-l-m with a total distance of  $200+160+165+250+60=835$  km. Both returns the same total distance

travelled.

[5 marks]

## 2. Artificial Neural Networks and Machine Learning (25 marks)

- i. What are the three activation functions that are used in an artificial neuron? For each activation function, mathematically formulate it and also provide a general graphical representation. **K**

[12 marks]

*Answer:*

$$\text{step}_t(x) = \begin{cases} 1, & x \geq t \\ 0, & x < t \end{cases}$$

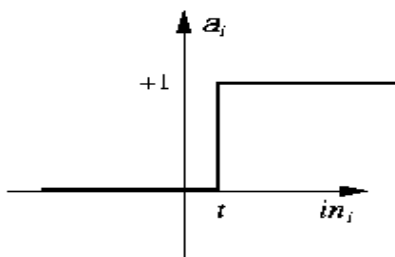
$$\text{sign}(x) = \begin{cases} 1, & x \geq 0 \\ -1, & x < 0 \end{cases}$$

$$\text{sigmoid}(x) = \frac{1}{1 + e^{-x}}$$

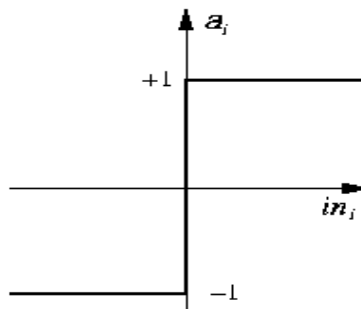
[2 marks for each point]

*In the following figure,  $x = \text{in}_i$  is the weighted sum of activation values of  $j$ s (inputs) to unit  $i$ .*

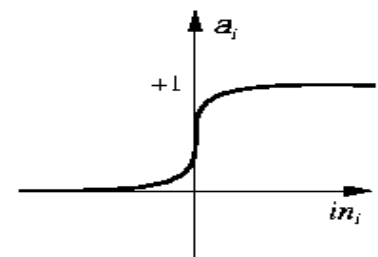
*Sample diagram*



(a) Step function



(b) Sign function



(c) Sigmoid function

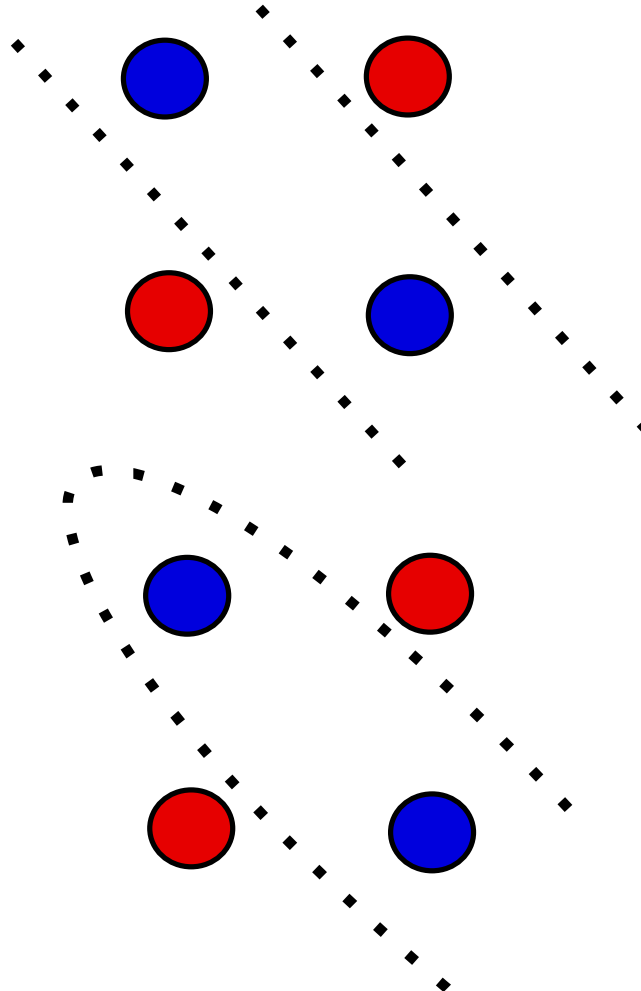
[2 marks for each figure]

- ii. Consider two-dimensional classification problems that are not linearly separable (XOR-like). Explain with the aid of conceptual diagrams two ways in which classifiers such as neural networks solve these problems. **C**

[10 marks]

Answer:

The input-output response separating the two classes of the 2D feature space can be illustrated from the clusters of data points in the figure.



[6 marks]

One can combine neurons with linear activation functions in a multilayer setup so that the two lines partitioned the two classes (top figure).

Or one can use neurons with a nonlinear activation function (bottom figure).

[4 marks]

- iii. Explain briefly the main difference between supervised and unsupervised learning. **K,C**  
[3 marks]

Answer:

The main difference is on data labelling – supervised learning requires having a set of labelled training pair  $(\mathbf{x}, y)$ .

[3 marks]