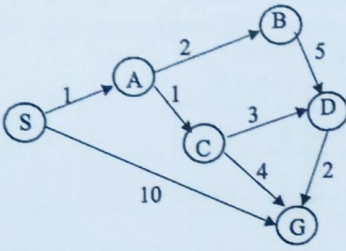


**DECEMBER 2021: END SEMESTER ASSESSMENT (ESA) B TECH V SEMESTER**

**UE19CS303 – Machine Intelligence**

Time: 3 Hrs		Answer All Questions		Max Marks: 100																																																			
1	a.	<p>Consider the graph given below. The heuristic value of all states is given in the table. Use A* algorithm to reach the goal state G from the start state S. Show all the steps.</p> <table><tr><th>State</th><th>h(n)</th></tr><tr><td>S</td><td>5</td></tr><tr><td>A</td><td>3</td></tr><tr><td>B</td><td>4</td></tr><tr><td>C</td><td>2</td></tr><tr><td>D</td><td>6</td></tr><tr><td>G</td><td>0</td></tr></table> 			State	h(n)	S	5	A	3	B	4	C	2	D	6	G	0	5																																				
	State	h(n)																																																					
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D	6																																																						
G	0																																																						
b.	<p>Consider the following dataset of a cooking competition:</p> <table><tr><th>Cook id</th><td>1</td><td>2</td><td>3</td><td>4</td><td>5</td><td>6</td><td>7</td><td>8</td><td>9</td></tr><tr><th>Garnishing of Dish1</th><td>C</td><td>C</td><td>I</td><td>I</td><td>I</td><td>I</td><td>I</td><td>C</td><td>I</td></tr><tr><th>Garnishing of Dish2</th><td>C</td><td>C</td><td>I</td><td>I</td><td>C</td><td>C</td><td>I</td><td>I</td><td>C</td></tr><tr><th>Average time taken(min)</th><td>1.0</td><td>6.0</td><td>5.0</td><td>4.0</td><td>7.0</td><td>3.0</td><td>8.0</td><td>7.0</td><td>5.0</td></tr><tr><th>Selected for next round</th><td>Yes</td><td>Yes</td><td>No</td><td>Yes</td><td>No</td><td>No</td><td>No</td><td>Yes</td><td>No</td></tr></table> <p>(where Garnishing of Dish 1 and Dish2 can be Completed (C) or Incomplete (I)).</p> <p>Which attribute (between garnishing of dish1 and garnishing of dish2) gives the better split based on information gain?</p>			Cook id	1	2	3	4	5	6	7	8	9	Garnishing of Dish1	C	C	I	I	I	I	I	C	I	Garnishing of Dish2	C	C	I	I	C	C	I	I	C	Average time taken(min)	1.0	6.0	5.0	4.0	7.0	3.0	8.0	7.0	5.0	Selected for next round	Yes	Yes	No	Yes	No	No	No	Yes	No	1+2+2	
Cook id	1	2	3	4	5	6	7	8	9																																														
Garnishing of Dish1	C	C	I	I	I	I	I	C	I																																														
Garnishing of Dish2	C	C	I	I	C	C	I	I	C																																														
Average time taken(min)	1.0	6.0	5.0	4.0	7.0	3.0	8.0	7.0	5.0																																														
Selected for next round	Yes	Yes	No	Yes	No	No	No	Yes	No																																														
c.	<p>For the following rational agents determine the PEAS (Performance Measure, Environment, Actuator and Sensors) components. For the environment component also mention the type of environment.</p> <p>(i) Automated Car Driving agent</p> <p>(ii) Online English Tutor agent</p> <p>Is the task environment fully observable for both the above agents?</p>			2+2+1 2																																																			
d.	<p>Define the following terms: (i) Version Space (ii) Inductive Bias (iii) ROC curve (iv) Overfitting (v) Sensitivity</p>			1*5																																																			
2	a.	<p>For the data given below, classify the point x=5.0 based on its 1-, 3- and 5- nearest neighbors (using Manhattan distance and majority vote)</p> <table><tr><th>x</th><td>5.0</td><td>3.0</td><td>4.5</td><td>4.6</td><td>4.9</td><td>5.2</td><td>5.3</td><td>5.5</td><td>7.0</td><td>9.5</td></tr><tr><th>y</th><td>-</td><td>-</td><td>+</td><td>+</td><td>+</td><td>-</td><td>-</td><td>+</td><td>-</td><td>-</td></tr></table> <p>What happens if the value of k is too small or too big?</p>			x	5.0	3.0	4.5	4.6	4.9	5.2	5.3	5.5	7.0	9.5	y	-	-	+	+	+	-	-	+	-	-	1*5																												
	x	5.0	3.0	4.5	4.6	4.9	5.2	5.3	5.5	7.0	9.5																																												
y	-	-	+	+	+	-	-	+	-	-																																													



b	8	<p>Show that <math>\Delta w_{ji} = \eta(t_j - o_j)o_j(1 - o_j)x_{ji}</math> and <math>\Delta w_{ji} = \eta\delta_j x_{ji}</math> where <math>x_{ji}</math> = the <math>i</math>th input to unit <math>j</math>, <math>w_{ji}</math> = the weight associated with the <math>i</math>th input to unit <math>j</math>, <math>net_j = \sum_i w_{ji}x_{ji}</math> (the weighted sum of inputs for unit <math>j</math>), <math>o_j</math> = the output computed by unit <math>j</math>, <math>t_j</math> = the target output for unit <math>j</math>, <math>a</math> = the sigmoid function, <math>outputs</math> = the set of units in the final layer of the network, <math>Downstream(j)</math> = the set of units whose immediate inputs include the output of unit <math>j</math>, <math>\eta</math> is the learning rate and <math>\delta_j</math> is the error term for <math>j</math>th unit.</p>												
c	7	<p>Consider the following dataset with two attributes as Attribute 1 and Attribute 2 and determine the equations of marginal planes and the optimal hyperplane.</p> <table border="1"><thead><tr><th></th><th>Attribute 1</th><th>Attribute 2</th><th>Class Label Y</th></tr></thead><tbody><tr><td>X1</td><td>1</td><td>3</td><td>1</td></tr><tr><td>X2</td><td>3</td><td>1</td><td>-1</td></tr></tbody></table>		Attribute 1	Attribute 2	Class Label Y	X1	1	3	1	X2	3	1	-1
	Attribute 1	Attribute 2	Class Label Y											
X1	1	3	1											
X2	3	1	-1											

3	a	6	<p>Consider a learner <math>L</math> on an instance space <math>X</math> and a hypothesis space <math>H</math> consisting of some class of real-valued function defined over <math>X</math>. <math>L</math> learns an unknown target function <math>f: X \rightarrow \mathbb{R}</math>, where <math>\mathbb{R}</math> is set of real numbers. 'm' training examples are provided to the learner <math>L</math> and the target value of each training example is corrupted by a random noise <math>e_i</math>, which follows a <i>normal distribution</i>. Just to bring you to a comfort zone, each training example is of the form <math>\langle x_i, d_i \rangle</math> where <math>d_i = f(x_i) + e_i</math>. Now, show that the Maximum Likelihood hypothesis <math>h_{ML}</math> is the one that minimizes the sum of the squared errors between the observed training values <math>d_i</math> and the hypothesis prediction <math>h(x_i)</math> under certain assumptions.</p>																																																						
b	3+3+1	<p>Consider the data set shown below:</p> <table border="1"><thead><tr><th>Record</th><th>1</th><th>2</th><th>3</th><th>4</th><th>5</th><th>6</th><th>7</th><th>8</th><th>9</th><th>10</th></tr></thead><tbody><tr><td>A</td><td>0</td><td>0</td><td>0</td><td>0</td><td>0</td><td>1</td><td>1</td><td>1</td><td>1</td><td>1</td></tr><tr><td>B</td><td>0</td><td>0</td><td>1</td><td>1</td><td>0</td><td>0</td><td>0</td><td>0</td><td>1</td><td>0</td></tr><tr><td>C</td><td>0</td><td>1</td><td>1</td><td>1</td><td>1</td><td>1</td><td>1</td><td>1</td><td>1</td><td>1</td></tr><tr><td>Class</td><td>+</td><td>-</td><td>-</td><td>-</td><td>+</td><td>+</td><td>-</td><td>-</td><td>+</td><td>+</td></tr></tbody></table> <p>Estimate the conditional probabilities for <math>P(A +)</math>, <math>P(B +)</math>, <math>P(C +)</math>, <math>P(A -)</math>, <math>P(B -)</math>, and <math>P(C -)</math>. Use the estimate of conditional probabilities obtained so far to predict the class label for a test sample (<math>A=0</math>, <math>B=1</math>, <math>C=0</math>) using the Naive Bayes approach. Why m-estimate of conditional probability in Bayesian classifier is required?</p>	Record	1	2	3	4	5	6	7	8	9	10	A	0	0	0	0	0	1	1	1	1	1	B	0	0	1	1	0	0	0	0	1	0	C	0	1	1	1	1	1	1	1	1	1	Class	+	-	-	-	+	+	-	-	+	+
Record	1	2	3	4	5	6	7	8	9	10																																															
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C	0	1	1	1	1	1	1	1	1	1																																															
Class	+	-	-	-	+	+	-	-	+	+																																															
c	2+5	<p>Hidden Markov Models are characterized by three fundamental problems. Clearly state those three fundamental problems.</p> <p>Say, there are two states <math>S_1</math> and <math>S_2</math> and four observations <math>O_1</math>, <math>O_2</math>, <math>O_3</math> and <math>O_4</math>. The initial state probabilities of <math>S_1</math> and <math>S_2</math> being (0.8, 0.2). Given below are the transition and emission probabilities:</p> <table border="1"><thead><tr><th></th><th><math>S_1</math></th><th><math>S_2</math></th></tr></thead><tbody><tr><td><math>S_1</math></td><td>0.99</td><td>0.01</td></tr><tr><td><math>S_2</math></td><td>0.1</td><td>0.9</td></tr></tbody></table> <table border="1"><thead><tr><th></th><th><math>O_1</math></th><th><math>O_2</math></th><th><math>O_3</math></th><th><math>O_4</math></th></tr></thead><tbody><tr><td>Given <math>S_1</math></td><td>0.2</td><td>0.2</td><td>0.4</td><td>0.2</td></tr><tr><td>Given <math>S_2</math></td><td>0.4</td><td>0.3</td><td>0.1</td><td>0.2</td></tr></tbody></table> <p>Compute the best hidden state sequence given observation (<math>O_1, O_2, O_3</math>).</p>		$S_1$	$S_2$	$S_1$	0.99	0.01	$S_2$	0.1	0.9		$O_1$	$O_2$	$O_3$	$O_4$	Given $S_1$	0.2	0.2	0.4	0.2	Given $S_2$	0.4	0.3	0.1	0.2																															
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4	a	<p>The table below is the distance matrix of 6 objects:</p> <table><tr><td></td><td>A</td><td>B</td><td>C</td><td>D</td><td>E</td><td>F</td></tr><tr><td>A</td><td>0</td><td></td><td></td><td></td><td></td><td></td></tr><tr><td>B</td><td>0.12</td><td>0</td><td></td><td></td><td></td><td></td></tr><tr><td>C</td><td>0.51</td><td>0.25</td><td>0</td><td></td><td></td><td></td></tr><tr><td>D</td><td>0.84</td><td>0.16</td><td>0.14</td><td>0</td><td></td><td></td></tr><tr><td>E</td><td>0.28</td><td>0.77</td><td>0.70</td><td>0.45</td><td>0</td><td></td></tr><tr><td>F</td><td>0.34</td><td>0.61</td><td>0.93</td><td>0.20</td><td>0.67</td><td>0</td></tr></table> <p>Show the final result of hierarchical clustering with complete link on these 6 data objects and also by drawing a dendrogram.</p>		A	B	C	D	E	F	A	0						B	0.12	0					C	0.51	0.25	0				D	0.84	0.16	0.14	0			E	0.28	0.77	0.70	0.45	0		F	0.34	0.61	0.93	0.20	0.67	0	3+2
	A	B	C	D	E	F																																														
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	b	<p>Compute the principal components of following data-</p> <p>CLASS 1</p> <p style="text-align: center;"><math>X = 2, 3, 4</math></p> <p style="text-align: center;"><math>Y = 1, 5, 3</math></p> <p>CLASS 2</p> <p style="text-align: center;"><math>X = 5, 6, 7</math></p> <p style="text-align: center;"><math>Y = 6, 7, 8</math></p>	5																																																	
	c	<p>Differentiate between the following with example: (i) Tournament Selection vs. Rank Selection (ii) Uniform Crossover vs. Masked Uniform Crossover</p>	5																																																	
	d	<p>Explain the working of Particle Swarm Optimization with proper expression for update of velocity and position of the particle and also state the mechanism to control the velocity of the particle in search space.</p>	5																																																	
5	a	<p>Suppose an RNN takes a vector input of length <math>2n</math>, it has one hidden layer with <math>n/2</math> neurons and one output layer with 4 neurons. If the total number of weights be 20(not including bias). Find the value of <math>n</math>.</p>	2																																																	
	b	<p>Suppose Apple has hired you as a Machine Learning Engineer. Initially you have been put onto a project where you will be using different architectures of Recurrent Neural Networks (RNNs). Consider the following tasks and justify which architecture of RNN will be most appropriate for the following:</p> <ul style="list-style-type: none"><li>(i) Sentiment Analysis</li><li>(ii) Machine Translation</li><li>(iii) DNA Sequence Analysis</li><li>(iv) Image Captioning</li><li>(v) Predicting next character in a word</li></ul>	1*5																																																	
	c	<p>Write the expressions for loss function with no regularization, with <math>L_1</math> regularization and <math>L_2</math> regularization. Explain briefly the concept of 'Dropout'.</p>	3+2																																																	

Consider the following 3 input channels as A, B, C:

1	0	1	0	2
1	1	3	2	1
1	1	0	1	1
2	3	2	1	3
0	2	0	1	0

I/p Channel A

1	0	0	1	0
2	0	1	2	0
3	1	1	3	0
0	3	0	3	2
1	0	3	2	1

I/p Channel B

2	0	1	2	1
3	3	1	3	2
2	1	1	1	0
3	1	3	2	0
1	1	2	1	1

I/p Channel C

And the three kernel matrices corresponding to I/p channels A, B and C are as follows respectively:

0	1	0
0	0	2
0	1	0

2	1	0
0	0	0
0	3	0

1	0	0
1	0	0
0	0	2

Bias corresponding to all 3 input channels is +1 and vertical and horizontal stride being 2. Obtain the feature map by applying the convolution operation on inputs and kernels and then apply average pool to get final representation of the input that will be sent to fully connected layer.