

University of Mumbai

Program: **Computer Engineering**

Curriculum Scheme: Rev2019

Examination: Final Year

Semester: V

Course Code: CSDLO5011 Course Name : Probabilistic Graphical Models

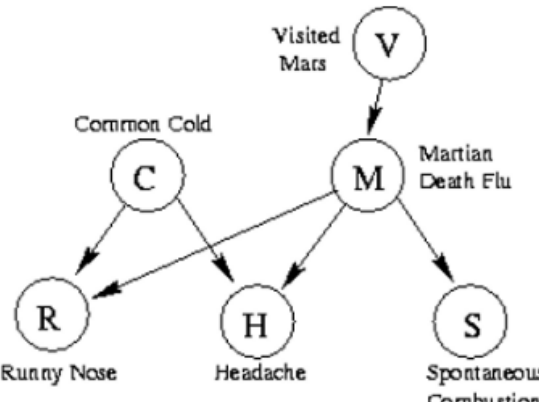
Time: 2.5 hour (150 minutes)

Max. Marks: 80

Q1. All questions compulsory 2 marks each (20 Marks)

| Q1. | A new credit card has been issued to 2000 customers. Of these customers, 1500 hold a Visa, 500 hold an AA card, and 40 hold a Visa and AA card. Find the probability that a random chosen customer holds an AA, given they hold a Visa. | | | | | | | | | | | | |
|-----------|------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------|---|------|----|-----|----|-----|----|-----|----|-----|----|-----|
| Option A: | .0267 | | | | | | | | | | | | |
| Option B: | .02 | | | | | | | | | | | | |
| Option C: | 37.5 | | | | | | | | | | | | |
| Option D: | .3333 | | | | | | | | | | | | |
| | | | | | | | | | | | | | |
| Q2. | Is this a probability distribution? <table border="1" style="margin-left: auto; margin-right: auto;"><thead><tr><th>x</th><th>P(x)</th></tr></thead><tbody><tr><td>75</td><td>.12</td></tr><tr><td>80</td><td>.23</td></tr><tr><td>85</td><td>.42</td></tr><tr><td>90</td><td>.22</td></tr><tr><td>95</td><td>.11</td></tr></tbody></table> | x | P(x) | 75 | .12 | 80 | .23 | 85 | .42 | 90 | .22 | 95 | .11 |
| x | P(x) | | | | | | | | | | | | |
| 75 | .12 | | | | | | | | | | | | |
| 80 | .23 | | | | | | | | | | | | |
| 85 | .42 | | | | | | | | | | | | |
| 90 | .22 | | | | | | | | | | | | |
| 95 | .11 | | | | | | | | | | | | |
| Option A: | No, the sum of p(x) does not equal 1. | | | | | | | | | | | | |
| Option B: | Yes, all p(x) are between 0 and 1. | | | | | | | | | | | | |
| Option C: | No, all p(x) are not between 0 and 1. | | | | | | | | | | | | |
| Option D: | Yes, the sum p(x) is 1. | | | | | | | | | | | | |
| | | | | | | | | | | | | | |
| Q3. | Which one of the following is a discrete random variable? | | | | | | | | | | | | |
| Option A: | Sam's height | | | | | | | | | | | | |

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| Option B: | Sam weight |
| Option C: | Time Sam's runs the 110 m hurdles |
| Option D: | Amount of Sam's brothers and sisters |
| | |
| Q4. | The degree of vertex is.... |
| Option A: | the number of edges attached to the vertex |
| Option B: | the number of lines and edges divided by two |
| Option C: | the number of vertices attached to edges |
| Option D: | the eulers equation |
| | |
| Q5. | A distribution P_{ϕ} is a Gibbs distribution parameterized with a set of factors $\phi_1(D_1); \dots; \phi_m(D_m)$ if it is defined as |
| Option A: | $P_{\phi}(X_1; \dots; X_n) = 1/Z \phi_1(D_1) \times \dots \times \phi_m(D_m)$ |
| Option B: | $P_{\phi}(X_1; \dots; X_n) = 1/z \phi_1(D_1) + \dots + \phi_m(D_m)$ |
| Option C: | $P_{\phi}(X_1; \dots; X_n) = \phi_1(D_1) \times \dots \times \phi_m(D_m)$ |
| Option D: | $P_{\phi}(X_1; \dots; X_n) = Z \phi_1(D_1) \times \dots \times \phi_m(D_m)$ |
| | |
| Q6. | If every Undirected Path from a node in X to a node in Y is D-separated by E, then X and Y are _____ given E |
| Option A: | Independent |
| Option B: | dependent |
| Option C: | conditionally dependent |
| Option D: | conditionally Independent |
| | |
| Q7. | Which of the following Problem is not solved with Hidden Markov Models |
| Option A: | Learning Problem |
| Option B: | Decoding Problem. |
| Option C: | Evaluation Problem |
| Option D: | Encoding Problem |
| | |
| Q8. | Which algorithm is used for solving temporal probabilistic reasoning? |
| Option A: | Depth-first search |
| Option B: | Breadth-first search |
| Option C: | Hidden markov model |
| Option D: | Hill-climbing search |
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| Q9. | <p>Let us define an HMM Model with K classes for hidden states and T data points as observations. The dataset is defined as $X = \{x_1, x_2, \dots, x_T\}$ and the corresponding hidden states are $Z = \{z_1, z_2, \dots, z_T\}$. Please note that each x_i is an observed variable and each z_i can belong to one of classes for hidden state. What will be the size of the state transition matrix, and the emission matrix, respectively for this example.</p> |
| Option A: | K*K,K*T |
| Option B: | K*T,K*T |
| Option C: | K*K,K*K |
| Option D: | K*T,K*K |
| Q10. | <p>Select the option describing Bayesian chain rule for the following graph.</p>  <p>A. $P(V, C, R, M, H, S) = P(R). P(H). P(S). P(C R, H). P(M R, H, S). P(V M)$ B. $P(V, C, R, M, H, S) = P(V). P(C). P(M V). P(R C, M, V). P(H C, M). P(S M)$ C. $P(V, C, R, M, H, S) = P(R). P(H). P(S). P(C R, H). P(M R, H, S). P(V M, R, H, S)$ D. $P(V, C, R, M, H, S) = P(V). P(C). P(M V). P(R C, M). P(H C, M). P(S M)$</p> |
| Option A: | A |
| Option B: | B |
| Option C: | C |
| Option D: | D |

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|--------------------------------------|------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------|---------------------|-------------|-----------|------------|-----------|------------|-----------|------------|-----------|------------|--|
| Q2. (20 Marks Each) | Solve any Four out of Six | 5 marks each | | | | | | | | | | |
| A | Assume that a factory has two machines. Past record shows that machine 1 produces 30% of the items of the output and machine 2 produces rest 70% of the items produced. Further 10% of the items produced by both machine 1 and 2 were defective. If a defective item is drawn at random, What is the probability that the defective item was produced by either Machine 1 or Machine 2 | | | | | | | | | | | |
| B | Given the following data Find the Variance <table style="margin-left: auto; margin-right: auto;"> <tr> <td>x</td> <td>P(x)</td> </tr> <tr> <td>10</td> <td>0.2</td> </tr> <tr> <td>20</td> <td>0.2</td> </tr> <tr> <td>30</td> <td>0.1</td> </tr> <tr> <td>40</td> <td>0.5</td> </tr> </table> | x | P(x) | 10 | 0.2 | 20 | 0.2 | 30 | 0.1 | 40 | 0.5 | |
| x | P(x) | | | | | | | | | | | |
| 10 | 0.2 | | | | | | | | | | | |
| 20 | 0.2 | | | | | | | | | | | |
| 30 | 0.1 | | | | | | | | | | | |
| 40 | 0.5 | | | | | | | | | | | |
| C | Write a short note on Gibbs distribution of Markov Network Model | | | | | | | | | | | |
| D | Define the ways how information can propagate in Bayesian Network with Example. | | | | | | | | | | | |
| E | Define Template Variables and Template Factors with examples. | | | | | | | | | | | |
| F | Write a short note on learning parameters with incomplete dataset (with missing values) with example | | | | | | | | | | | |

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| Q3. (20 Marks Each) | Solve any Two Questions out of Three | 10 marks each |
| A | How does conditional independence and conditional parameterization reduce the number of parameters to represent probability distribution more compactly? Explain with example. | |
| B | Explain Viterbi Algorithm with Example. | |
| C | When parent node and child node become Conditionally Independent? Justify with example scenarios. | |

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| Q4. | Solve any Two Questions out of Three | 10 marks each |
|------------|---------------------------------------------|----------------------|

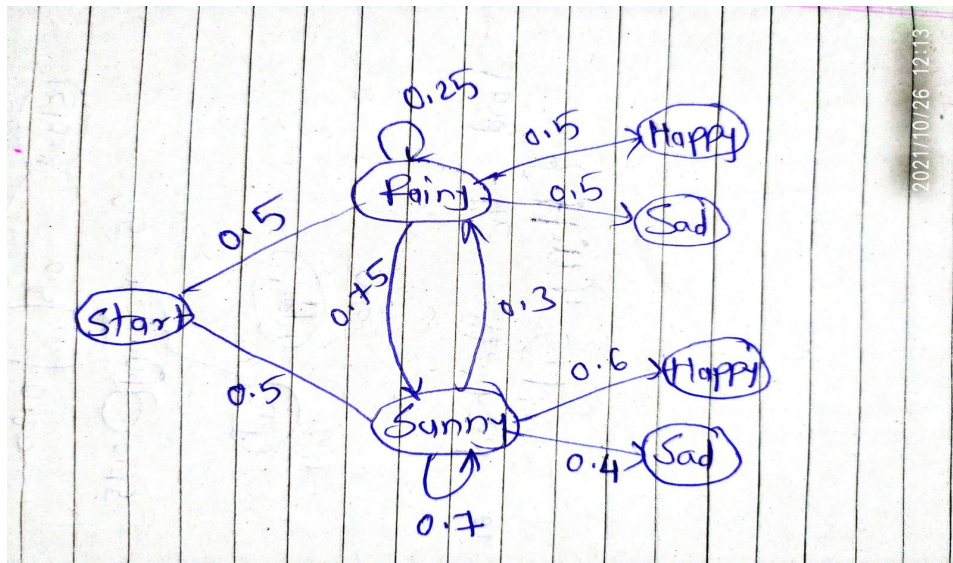
(20 Marks Each)

Today's weather is (outlook=Sunny, Temp=Mild, Humidity=Normal, Windy=False). Is Today's weather suitable for playing Golf?

| Outlook | Temp | Humidity | Windy | Play Golf |
|----------|------|----------|-------|-----------|
| Rainy | Hot | High | False | No |
| Rainy | Hot | High | True | No |
| Overcast | Hot | High | False | Yes |
| Sunny | Mild | High | False | Yes |
| Sunny | Cool | Normal | False | Yes |
| Sunny | Cool | Normal | True | No |
| Overcast | Cool | Normal | True | Yes |
| Rainy | Mild | High | False | No |
| Rainy | Cool | Normal | False | Yes |
| Sunny | Mild | Normal | False | Yes |
| Rainy | Mild | Normal | True | Yes |
| Overcast | Mild | High | True | Yes |
| Overcast | Hot | Normal | False | Yes |
| Sunny | Mild | High | True | No |

A

Given the HMM



B

Find the likelihood of the sequence {Happy,Sad}

C

Given the following Factor table between variables A,B,C.Find the joint distribution function

| A | B | Phi |
|---|---|-----|
| 0 | 0 | 10 |
| 0 | 1 | 5 |
| 1 | 0 | 3 |
| 1 | 1 | 9 |

| B | C | Phi |
|---|---|-----|
| 0 | 0 | 20 |
| 0 | 1 | 15 |
| 1 | 0 | 13 |
| 1 | 1 | 10 |