| | | - | | | Р | CS5H002 |
|----------------|--|--|--|--|------------|------------------------------|
| | | | ster Regular / Back Examina | | | |
| | | DAT | TA MINING & DATA WAREH | OUSING | | |
| | | 040 | BRANCH : CSE | 040 | 040 | |
| | | 210 210 | 210 Time : 3 Hours | 210 | 210 | 2 |
| | | | Max Marks: 100 | | | |
| _ | | | Q.CODE : E384 | | | =14.6 |
| An | swe | · | 1) which is compulsory, any from Part-III. | | t-II and a | iny TWO |
| | | I ne figure | s in the right hand margin in | ndicate marks. | | |
| | | 210 210 | 210 Part- I 210 | 210 | 210 | 2 |
| Q1 | | Short Answer Type Que | • | | | (2 x 10) |
| | | List out the operation of C | | | | |
| | • | What are the characterist | tics of data ware house? | | | |
| | c) | Define star scheme. | | | | |
| | d) | | | | | |
| | e) | Mention the purpose of A | | | | |
| | f) | How prediction is differen | | | | |
| | | 2State the role of cluster a | | 210 | 210 | 2 |
| | h) | Write the strengths of hie | | | | |
| | i) | | eps in data transformation? | | | |
| | j) | What do you mean by sp | artial data mining? | | | |
| | | | Part- II | | | |
| Q2 | | Focused-Short Answer | Type Questions- (Answer An | y EIGHT out of TW | ELVE) | (6 x 8) |
| J۷ | | | | | | |
| 42 | a) | What are the Steps involved | ved in data preprocessing? | | | |
| J Z | • | | ved in data preprocessing? three tier architecture of⊵Data Wa | arehousing. | 210 | 2 |
| 42 | • | with neat sketch explain t | | | 210 | 2 |
| 42 | b) | with neat sketch explain t What is decision tree? Ex | three tier architecture of⊵Data Wa | ification problem? | 210 | 2 |
| 42 | b) c) | with neat sketch explain t What is decision tree? Ex | three tier architecture of Data Wa kplain how does it work for classi e steps of knowledge discovery. | ification problem? | 210 | 2 |
| 4 4 | b) c) d) | with neat sketch explain t What is decision tree? Ex Discuss in detail about th | three tier architecture of Data Wa oplain how does it work for classing e steps of knowledge discovery. tion by normalization. | ification problem? | 210 | 2 |
| 4 4 | b) c) d) e) f) | with neat sketch explain to What is decision tree? Explain the Discuss in detail about the Illustrate data transforma Describe about Cognous | three tier architecture of Data Wa oplain how does it work for classing e steps of knowledge discovery. tion by normalization. | ification problem? | 210 | 2 |
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| 4 2 | b) c) d) e) f) g) h) | with neat sketch explain to What is decision tree? Ex Discuss in detail about the Illustrate data transforma Describe about Cognous Discuss different tools ca | three tier architecture of Data Waxplain how does it work for classing e steps of knowledge discovery. It tier to by normalization. Impromptu. It tegorize in data warehouse busing are generated from frequer | ification problem? | 210 | 2 |
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| 4 2 | b) c) d) e) f) g) h) | with neat sketch explain to What is decision tree? Explains in detail about the Illustrate data transforma Describe about Cognous Discuss different tools cat Explain how association of Describe various data cles Explain FP-Growth algoritime? | three tier architecture of Data Waxplain how does it work for classing e steps of knowledge discovery. It is to be normalization. Impromptu. It is to data warehouse busing are generated from frequereaning techniques. | ification problem? ness analysis. nt item sets. | | |
| 44 | b) c) d) e) f) g) h) i) | with neat sketch explain to What is decision tree? Explains in detail about the Illustrate data transforma Describe about Cognous Discuss different tools cat Explain how association of Describe various data cles Explain FP-Growth algoritime? | three tier architecture of Data Waxplain how does it work for classic esteps of knowledge discovery. It is to by normalization. Impromptu. Impromptu. It is to data warehouse busicules are generated from frequer eaning techniques. It is to details. | ification problem? ness analysis. nt item sets. | | |
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| Q 3 | b) c) d) e) f) g) h) i) k) | with neat sketch explain to What is decision tree? Explains to Discuss in detail about the Illustrate data transforma Describe about Cognous Discuss different tools can explain how association to Describe various data clee Explain FP-Growth algorith Discuss K-Nearest neight Write short note on text in Long Answer Type Que | three tier architecture of Data Waxplain how does it work for classic esteps of knowledge discovery tion by normalization. Impromptu. Itegorize in data warehouse busicules are generated from frequer eaning techniques. Ithm in details. 210 bor classification algorithm. Inining. Part-III | ification problem? iness analysis. nt item sets. 210 | | |
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