

Hi

ChatGPT said:

Hi Rashid! 😊

How can I help you today?

You said:

I want answers for my practical exam question paper

ChatGPT said:

Sure, Rashid — please upload a photo or paste the text of your practical exam question paper here. I'll go through the questions and help you with clear, accurate answers.

You said:

Section A

Computer networks lab

Shall we start?

ChatGPT said:

Yes, let's start! 🖨️

Please share Question 1 from Section A (Computer Networks Lab), and I'll help you with the answer step by step.

You said:

Start a TCP application and monitor the packet flow. Make necessary

assumptions.

ChatGPT said:

Great! Let's break this down:

✓ Question:

"Start a TCP application and monitor the packet flow. Make necessary assumptions."

✓ Assumptions:

We're using Linux or Windows OS

We have tools like:

netcat (nc) for TCP application

Wireshark or tcpdump for packet monitoring

The TCP application is a simple client-server model

✓ Step-by-step Answer:

◆ 1. Start a TCP Server (using netcat)

On Terminal/Command Prompt (Server side):

bash

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nc -l -p 5000

This starts a TCP server listening on port 5000.

◆ 2. Start a TCP Client (on same or different machine):

bash

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nc <server-ip> 5000

Replace <server-ip> with actual IP (e.g., 127.0.0.1 if on same machine)

Type any message after connecting. Example:

bash

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Hello Server!

### ◆ 3. Monitor Packet Flow using Wireshark:

Open Wireshark

Select the network interface (e.g., Ethernet, Wi-Fi, or lo for localhost)

Apply filter:

ini

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tcp.port == 5000

Start capturing

You'll see:

TCP 3-way handshake (SYN, SYN-ACK, ACK)

Data transmission packets

FIN packets if connection is closed

### ◆ 4. Monitor using tcpdump (alternative to Wireshark):

bash

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sudo tcpdump -i any tcp port 5000

✓ Output Observations (What to write in exam):

TCP handshake initiated:

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SYN → SYN-ACK → ACK

Message data:

kotlin

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PSH-ACK (data sent)

Connection termination:

nginx

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FIN → FIN-ACK → ACK

✓ Conclusion:

A TCP application was successfully initiated using netcat. The packet flow including the handshake, data transfer, and termination was monitored using Wireshark/tcpdump.

Let me know if you want a diagram, screenshot labels, or practical write-up format. Ready for Question 2?

You said:

Create a UDP client on a node n1 and a UDP server on a node n2.

ChatGPT said:

Perfect, Rashid — here's the full answer for:

✓ Question 2:

Create a UDP client on a node n1 and a UDP server on a node n2.

✓ Assumptions:

Node n1 and n2 are two systems (or terminals in a network simulator like NS2, GNS3, or Mininet — or even two terminals on the same PC).

We're using Python or netcat (nc) to create UDP client-server.

UDP port chosen: 6000



Answer using Python



On Node n2 (UDP Server):

Create a file called udp\_server.py:

python

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```
import socket
```

```
server_socket = socket.socket(socket.AF_INET, socket.SOCK_DGRAM)
server_socket.bind(("0.0.0.0", 6000))
```

```
print("UDP Server is listening on port 6000...")
```

```
while True:
```

```
    data, addr = server_socket.recvfrom(1024)
```

```
    print(f"Received message from {addr}: {data.decode()}")
```

Run this on Node n2:

bash

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```
python3 udp_server.py
```



On Node n1 (UDP Client):

Create udp\_client.py:

python

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```
import socket
```

```
client_socket = socket.socket(socket.AF_INET, socket.SOCK_DGRAM)
server_address = ("<n2_IP_address>", 6000)
```

```
message = "Hello from Node n1"
client_socket.sendto(message.encode(), server_address)
client_socket.close()
Replace <n2_IP_address> with actual IP of n2 (e.g., 192.168.1.5).
```

Run on n1:

bash

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```
python3 udp_client.py
```

✓ Output (on n2 server terminal):

pgsql

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UDP Server is listening on port 6000...

Received message from ('192.168.1.x', 12345): Hello from Node n1

🔧 Alternative (Using netcat):

◆ On Node n2 (UDP server):

bash

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```
nc -u -l 6000
```

◆ On Node n1 (UDP client):

bash

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```
echo "Hello from n1" | nc -u <n2_IP> 6000
```

✓ Conclusion (Write this in practical answer):

A UDP client was created on Node n1 and a UDP server on Node n2. The client successfully sent a message over port 6000, which the server received without establishing a connection (since UDP is connectionless).

Ready for Question 3?

You said:

Create a simple point to point network topology using two nodes. Make necessary assumptions.

ChatGPT said:

Great Rashid! Let's answer:


✓ Question 3:

"Create a simple point-to-point network topology using two nodes. Make necessary assumptions."

✓ Assumptions:

We're using a network simulator like NS2, NS3, or Mininet.

Here, we'll use NS2 (Network Simulator 2), a common choice in practical labs.

 NS2 Code to Create Point-to-Point Topology with Two Nodes:

tcl

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# Filename: p2p.tcl

# Create simulator object

set ns [new Simulator]

# Open trace file

set tracefile [open out.tr w]

\$ns trace-all \$tracefile

```
# Create two nodes
set n0 [$ns node]
set n1 [$ns node]

# Create a duplex link between n0 and n1
$ns duplex-link $n0 $n1 1Mb 10ms DropTail

# Create a TCP agent and attach it to node n0
set tcp [new Agent/TCP]
$ns attach-agent $n0 $tcp

# Create a sink agent and attach it to node n1
set sink [new Agent/TCPSink]
$ns attach-agent $n1 $sink

# Connect TCP to sink
$ns connect $tcp $sink

# Create a CBR traffic source and attach to TCP
set cbr [new Application/Traffic/CBR]
$cbr attach-agent $tcp
$cbr set packetSize_ 500
$cbr set interval_ 0.005

# Schedule the events
$ns at 0.5 "$cbr start"
$ns at 4.5 "$cbr stop"
$ns at 5.0 "finish"

# Define finish procedure
proc finish {} {
    global ns tracefile
```



```
$ns flush-trace
close $tracefile
exit 0
}
```

# Run the simulation

```
$ns run
```

✓ How to Run:

Save as p2p.tcl

Run in terminal:

```
bash
```

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```
ns p2p.tcl
```

✓ Output (What to write in exam):

Two nodes, n0 and n1, connected via point-to-point link of 1Mb bandwidth and 10ms delay.

TCP agent at n0, TCP sink at n1.

Traffic generated using CBR over TCP.

Data flows from n0 → n1.

✓ Conclusion:

A basic point-to-point topology was created using NS2 with two nodes. One node sent data using TCP over a duplex link, and the other node received it. Traffic behavior was controlled using a CBR source.

Would you like the NAM visualization steps as well? Ready for Question 4?

You said:

Now section 2

Data mining

Shall we start?

ChatGPT said:

Absolutely, Rashid! 🏠

Let's begin Section B: Data Mining.

Please share Question 1, and I'll guide you with the correct practical steps and answers.

You said:

Create an EXCEL file. Convert the EXCEL file to .csv format and prepare it as .arff file. Make necessary assumptions.

ChatGPT said:

Great Rashid! Let's solve this Data Mining Question 1 step by step:

✓ Question:

"Create an EXCEL file. Convert the EXCEL file to .csv format and prepare it as .arff file. Make necessary assumptions."

✓ Assumptions:

We're using MS Excel for file creation.

The data is simple — e.g., student marks dataset.

We will convert it to .csv, then to .arff (used in Weka for data mining).

◆ Step 1: Create an Excel File (.xlsx)

In Excel, enter the following simple data:

Name	Age	Marks	Grade
Alice	20	85	A

Bob	22	75	B
Charlie	21	90	A
David	23	60	C

Then Save As → choose CSV (Comma delimited) (\*.csv)  
Give it a name like: students.csv

◆ Step 2: Convert CSV to ARFF format  
You can do this manually or using Weka.

📄 Manual Conversion:

Open students.csv in Notepad and write an .arff version like this:

arff

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@RELATION students

@ATTRIBUTE Name STRING

@ATTRIBUTE Age NUMERIC

@ATTRIBUTE Marks NUMERIC

@ATTRIBUTE Grade {A, B, C}

@DATA

Alice,20,85,A

Bob,22,75,B

Charlie,21,90,A

David,23,60,C

Save this as: students.arff

🔗 Alternate: Convert CSV to ARFF using Weka:

Open Weka Explorer

Go to Preprocess

Click Open file → choose your students.csv

Review the data

Click Save → Choose ARFF format → Save as students.arff

✓ Conclusion (Write in your answer sheet):

A dataset was created in Microsoft Excel and saved in .csv format. This CSV file was then successfully converted to the .arff format used in Weka by defining attributes and data according to ARFF syntax. The .arff file is now ready for use in data mining algorithms.

✓ Done!

Would you like to proceed to Question 2 of Data Mining now?

You said:

Demonstrate the preprocessing mechanism on a data set of your choice.

Make

necessary assumptions.

ChatGPT said:

Great, Rashid! Let's handle:

✓ Question 2 (Data Mining):

"Demonstrate the preprocessing mechanism on a data set of your choice.

Make necessary assumptions."

✓ Assumptions:

We're using Weka for preprocessing.

The dataset is students.arff (from Question 1).

Preprocessing tasks include:

Handling missing values

Discretization

Normalization

Removing unwanted attributes

☑ Steps to Demonstrate Preprocessing in Weka:

◆ Step 1: Load Dataset

Open Weka Explorer

Go to Preprocess tab

Click Open file → load students.arff

◆ Step 2: Handle Missing Values

(Assume some missing values in dataset)

Go to Filter > unsupervised > attribute > ReplaceMissingValues

Click Apply

☑ This fills missing numerical values with mean and nominal with mode.

◆ Step 3: Normalize Numeric Attributes

(Brings values to 0–1 range)

Go to Filter > unsupervised > attribute > Normalize

Apply → all numeric values (e.g., Age, Marks) will be scaled between 0 and 1

◆ Step 4: Discretize Numeric Attributes  
(Converts continuous to categorical)

Go to Filter > unsupervised > attribute > Discretize

Apply → attributes like Marks become intervals like:

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[60-70), [70-80), [80-90)

◆ Step 5: Remove Unwanted Attributes  
(Example: Remove Name)

Go to Filter > unsupervised > attribute > Remove

Choose attribute index (e.g., 1 for Name) → Apply

✓ Conclusion (Write this in answer sheet):

Preprocessing was demonstrated using the students.arff dataset in Weka. The steps included handling missing values using ReplaceMissingValues, normalizing numeric data using Normalize, converting continuous attributes to categorical using Discretize, and removing unnecessary attributes using Remove. These preprocessing steps are essential for preparing the data for data mining tasks.

✓ Done!

Would you like to continue to Question 3 of Data Mining?

You said:

Implement simple k-means Algorithm to demonstrate the clustering rule on any dataset(s) of your choice.

ChatGPT said:

Excellent, Rashid! Let's answer:

✔ Question 3 (Data Mining):

"Implement simple K-Means algorithm to demonstrate the clustering rule on any dataset(s) of your choice."

✔ Assumptions:

We are using Weka for simplicity.

Dataset used: Iris dataset (built-in in Weka)

It contains flower features: sepal length, sepal width, petal length, petal width.

🔧 Steps to Implement K-Means in Weka:

◆ Step 1: Load the Dataset

Open Weka Explorer

Go to Preprocess tab

Click Open file

Select:

bash

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data/iris.arff

(found inside Weka's data folder)

◆ Step 2: Remove Class Label (optional)

To do unsupervised clustering, you may remove the class attribute (species):

Use Filter > unsupervised > attribute > Remove

Choose attribute index 5 (for class) → Apply

### ◆ Step 3: Apply K-Means Clustering

Go to Cluster tab

Choose Cluster mode: "Use training set"

Click Choose > select:

nginx

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SimpleKMeans

Click on it to configure:

Number of clusters: 3 (since Iris has 3 types)

Distance Function: EuclideanDistance (default)

Click Start

### ◆ Step 4: View Results

Output shows:

yaml

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Cluster centroids:

Cluster#0: ...

Cluster#1: ...

Cluster#2: ...

Instances assigned to clusters:



Cluster 0: 50 instances

Cluster 1: 62 instances

Cluster 2: 38 instances

Each cluster groups similar instances (flowers) based on feature similarity.

✓ Conclusion (Write this in exam):

The Simple K-Means clustering algorithm was applied to the Iris dataset using Weka. The data was grouped into 3 clusters based on petal and sepal attributes. The algorithm used Euclidean distance to compute similarity, and output cluster centroids representing the average point of each cluster. This demonstrates how K-Means can effectively group similar data points without labels.