

NATIONAL INSTITUTE OF TECHNOLOGY PATNA

Department of Computer Science and Engineering MID SEMESTER EXAMINATION, Mar 2024 M.Tech.(CSE) 2nd Sem/PhD

Course Name: Data Visualization Techniques

Course Code: CS540203

Max. Marks: 30 Maximum Time: 2 hours

Instruction:

1. Attempt all questions.

2. Assume any suitable data, if necessary.

3. Answer all the questions in the order as appeared in the question paper and write all the sub-parts of a

S.N.	n in one place.		Questio	ns		Marks	CO	BL
							001	D 1
1(a) Briefly	describe the dat		3	COI	Remember			
(b) Describes the following: i) Data Objects ii) Attributes iii) Nominal data type iv) Ratio-based data type								Apply
	do you understa with suitable exa		dependency-orie	ented data? Explain	Multivariate time	3	CO4	Understand
2(a) Explain the foliation 3, 0, 12	n the five-numb lowing data in the 2, 0, 2, 0, 26, 0,	er summ ne form o 7, 5, 5, 2,	f boxplot. 1, 1, 2	Q1, Medium, Q2, Ma		5	CO3	Apply
2(b) Let's s	ay we have recor	rded data	about research	articles having attributen Publisher and Arti	tes like publishers	5	CO2	Apply
and ar	ticle types. Com	Index	Publishers	Article Types	cie types.			
		1	Elsevier	Book				
		2	Elsevier	Book				-
		3	Elsevier	Journal				
		4	Wiley	Journal				
		5	Wiley	Conference				
		6	Wiley	Book				
		7	Elsevier	Book				
		8	Wiley	Book				
genera	ate the figure give to 9, figure size	en below	. Consider the f	the matplotlib? Write to function $f(x)=x^2+8$, wo otation of 90-degree a	here x ranges	6	CO3	Apply
	Output: X ² + 8	0	- C m	input: x	8 6			
3(b) Briefl suitab	ly explain the ca	rtesian co	oordinate system	and polar coordinate	system with	4	CO3	Understar



NATIONAL INSTITUTE OF TECHNOLOGY PATNA

Department of Computer Science and Engineering
MID SEMESTER EXAMINATION, March 2024
B.Tech.(CSE) 4nd Sem

Course Name: Bioinformatics Course Code: CS540210 Max. Marks: 30

Maximum Time: 2 hours

Instructions:

1. Attempt all questions.

2. Assume any suitable data, if necessary.

3. Answer all the questions in the order as appeared in the question paper and write all the sub-parts of a question in one place.

S.N.	Questions	Mar	ks CO	BL
1.a)	Why UniGene is important in biological data retrieval?	2	COI	Remember
b)		1		Remember
c)	What CATH stands for? Why it is used?	2		
6	Differentiate between the following: Transcription and translation mRNA and rRNA tBLASTn and BLASTx RefSeq and GenBank PAM and BLOSUM	5	COI	Remember
3. a	Why tRNA is important protein synthesis?	2	C02 t	Inderstand
b	Can the base sequence of an mRNA be predicted from the amino acid sequence of its polypeptide product?	1		
c)	How BLAST algorithm works?	2		
4. a)	Is DNA synthesis semidiscontinous? If so, why?	2	CO2 U	Inderstand
b)	How progressive method of multiple sequence alignment works? Illustrate it using suitable diagram.	5		
c)	How the scoring function can be implemented using sum of pairs method to align four DNA sequences?	3		
5.	Given a set of sequence pairs, a and b:	5	CO3	Apply
	a: CTCGT b: CTAAGT Determine the "best" global alignment between them via trace-back procedure using Needlemann-Wunsch algorithm			



NATIONAL INSTITUTE OF TECHNOLOGY PATNA

DEPARTMENT OF COMPUTER SCIENCE AND ENGINEERING MID SEMESTER EXAMINATION, March, 2024

Programme: M. Tech (Data Science and Engineering)/PhD Semester: 2nd
Course Code: CS540202 Course Name: Deep Learning

Full marks:30

Answer *all* questions.

The use of *calculator* is allowed.

Q. no.	Question	Marks	СО	BL
1	Discuss the Gradient Descent algorithm to train a MLP in detail with all necessary equations. Consider both the cases of any neuron <i>i</i> when it is present in the output layer and when it is present in the hidden layer.	07	COI	Remembering
	a) Discuss the Hopfield network algorithm (both <i>training</i> and <i>recall</i>) with proper explanation.	05	CO2	Remembering
2	b) Suppose that a five-node Hopfield network has stored the patterns $X^{(1)}=+++++$, $X^{(2)}=+-+-+$, $X^{(3)}=+++$ If the new pattern -+++- is presented to the net, find out the stored pattern to which the net converges.	07	CO2	Application
3	a) Which drawback of discrete Hopfield network has led to the generation of continuous Hopfield network? Why are stochastic neurons used in a continuous Hopfield network? Why specifically was the Boltzmann machine invented where both Boltzmann machine and Hopfield network contain stochastic neurons?	07	CO2, CO3	Analysis
	b) Mathematically derive the Boltzmann learning rule with all necessary equations.	04	CO3	Understanding

NATIONAL INSTITUTE OF TECHNOLGY PATNA DEPARTMENT OF COMPUTER SCIENCE AND ENGINEERING MID-SEMESTER EXAMINATION - MARCH, 2024

M.Tech – Data Science & Engineering CS540213 – Recommendation Systems

Max.Marks:30

IInd Semester

Time: 2 hours

Note: Answer All questions, and all parts of the same question must be answered at the same place

a) We want to design a recommendation system for a restaurant that serves the entire world. The restaurant has over 5000 food items and one million customers, but its rating database has only 10 million ratings. Which of the following would be a better recommendation system, and justify your answer? (i) User-based collaborative filtering. (iii) Item-based collaborative filtering. (iii) Content-based recommendation system. (iv) Popularity-based recommendation system (Discussed in the class) b) Suppose the restaurant is using the recommendation system you suggested in the above (a). Consider a customer who rated only two items, *Rasgulla* and *Gulab Jamun*, and both ratings are five and five on a scale of five. Which of the following items is most likely to be recommended with proper justification: (i) Jalebi (ii) Motichoor Laddu (iii) It depends only on other users' ratings. (iv) Samosa (v) It depends only on other users' ratings. (iv) Samosa (v) It depends only on other item's ratings c) Assume that there is a very close neighbours v ₁ , v ₂ ,v ₄ of u. The users v ₁ , v ₂ ,v ₄ , in the user-based collaborative filtering to predict the rating for the item *f for user *u*. 2 a) Let *n* be the number of users, *m* be the number of items, *p* be the number of nearest neighbours, and *k* be the number of recommendations for each user. Assume that the *f* user has rated all the items and *f* item has rated by all users. Compute the time complexity for finding the similarity between users and similarity between items, and also compute the space complexity for storing similarity values in both cases. Also, find the time complexities for predicting the k recommendations for each user in both user-based collaborative filtering and item-based collaborative filtering. b) Justify the statement, "The cold start problem regarding the new items can be addressed using the content-based recommendation system." c) Consider the following dataset consisting of five features and a class label (likes/dislikes): *Keywor	Q. No				Ques					Marks	CO	В
in the above (a). Consider a customer who rated only two items, *Rasgulla* and *Gulab Jamun*, and both ratings are five and five on a scale of five. Which of the following items is most likely to be recommended with proper justification: (i) Jalebi (ii) Motichoor Laddu (iii) It depends only on other users' ratings. (iv) Samosa (v) It depends only on other item's ratings. (iv) Samosa (v) It depends only on other item's ratings. (iv) Samosa (v) It depends only on other users v ₁ , v ₂ , v _k of u. The users v ₁ , v ₂ , v _k have not rated the target item *t yet. Consider the number of nearest neighbours is *k\$. Propose an efficient solution for the user-based collaborative filtering to predict the rating for the item *t for user *u\$. 2 a) Let *n be the number of users, *m be the number of items, *p be the number of nearest neighbours, and *k\$ be the number of recommendations for each user. Assume that the *jth* user has rated all the items and *ith* item has rated by all users. Compute the time complexity for finding the similarity between users and similarity between items, and also compute the space complexity for storing similarity values in both cases. Also, find the time complexities for predicting the k recommendations for each user in both user-based collaborative filtering and item-based collaborative filtering. b) Justify the statement, "The cold start problem regarding the new items can be addressed using the content-based recommendation system." c) Consider the following dataset consisting of five features and a class label (likes/dislikes): Keyword Drums Guitar Beat Orchestra Classical Like Dislike 201 1 0 1 0 1 Like 202 0 0 1 1 0 1 Dislike 204 1 1 0 1 1 Dislike 205 1 1 0 1 1 Dislike 206 1 1 1 Dislike 207 1 1 1 Dislike 208 1 1 1 1 Dislike 209 1 1 1 Dislike 209 1 1 1 Dislike 200 1 1 1 Dislike 200 1 1 1 Dislike 200 200 200 200 200 2		customer following answer? (i) Use (ii) Iten (iii) Con (iv) Pop	r-based on-based ontent-based outent-based outen-based outent-based outent-based outent-based outent-based outen	rating da be a bet collaborati collaborati ed recomm	nt has o tabase ha tter recon ive filteri we filteri mendation	ver 5000 as only 10 ammendation ang. ang. an system.	food items million rat on system,	s and one mings. Which of and justify	illion of the your	3M	CO1 CO2 CO3	Creare, Evandate & AllalyZe
Assume that there is a very close neighbours v_1, v_2, v_k of u. The users v_1, v_2, v_k have not rated the target item t yet. Consider the number of nearest neighbours is k . Propose an efficient solution for the user-based collaborative filtering to predict the rating for the item t for user u . a) Let u be the number of users, u be the number of items, u be the number of nearest neighbours, and u be the number of recommendations for each user. Assume that the u user has rated all the items and u item has rated by all users. Compute the time complexity for finding the similarity between users and similarity between items, and also compute the space complexity for storing similarity values in both cases. Also, find the time complexities for predicting the u recommendations for each user in both user-based collaborative filtering and item-based collaborative filtering. b) Justify the statement, "The cold start problem regarding the new items can be addressed using the content-based recommendation system." c) Consider the following dataset consisting of five features and a class label (likes/dislikes): Keyword Drums Guitar Beat Orchestra Classical Like/ Dislike 201 1 0 1 0 1 Like 202 0 0 1 1 Like 203 0 0 1 1 Dislike 204 1 1 0 1 Dislike 205 1 1 0 1 Dislike		in the ab and Guld Which o justificat (i) Jalet (iv) Sam	ove (a). ab Jamu f the follo ion: ii (ii) Mo osa	Consider <i>n</i> , and boowing iten	a custom oth rating one is mos	recommender who rates are five at likely to li	dation systed only two and five of the recommon only on other contracts.	tem you sugger items, <i>Rasg</i> on a scale of ended with pre-	gulla five. coper	3M	COI CO3	Analyze
a) Let <i>n</i> be the number of users, <i>m</i> be the number of items, <i>p</i> be the number of nearest neighbours, and <i>k</i> be the number of recommendations for each user. Assume that the <i>j</i> th user has rated all the items and <i>i</i> th item has rated by all users. Compute the time complexity for finding the similarity between users and similarity between items, and also compute the space complexity for storing similarity values in both cases. Also, find the time complexities for predicting the k recommendations for each user in both user-based collaborative filtering and item-based collaborative filtering. b) Justify the statement, "The cold start problem regarding the new items can be addressed using the content-based recommendation system." c) Consider the following dataset consisting of five features and a class label (likes/dislikes): Keyword Drums Guitar Beat Orchestra Classical Like/ Dislike 201	-	neighbou collabora	irs is k tive filte	rated the ropos	close ne target iter se an e	ighbours v m t yet. Co fficient so rating for th	onsider the plution for	of u. The users number of near the user-bar	s v ₁ , arest ased		CO1 CO3	Apply
users. Compute the time complexity for finding the similarity between users and similarity between items, and also compute the space complexity for storing similarity values in both cases. Also, find the time complexities for predicting the k recommendations for each user in both user-based collaborative filtering and item-based collaborative filtering. b) Justify the statement, "The cold start problem regarding the new items can be addressed using the content-based recommendation system." c) Consider the following dataset consisting of five features and a class label (likes/dislikes): Keyword Drums Guitar Beat Orchestra Classical Like/ Dislike 201	a	nearest n Assume	the number ighbour that the j	s, and k be the user has	rs, <i>m</i> be to be the number rated al	he number nber of rec l the items	of items, <i>p</i> ommendat and <i>i</i> th ite	be the number ions for each up has rated by	iser.	3M	CO3	Remember & understand
c) Consider the following dataset consisting of five features and a class label (likes/dislikes): Keyword Song Id Drums Guitar Beat Orchestra Classical Like/ Dislike	h	and simi storing s predictin	empute the larity bear imilarity g the kattive filter	ween iter values in recomm	mplexity ms, and a both case endations tem-base	for finding also computes. Also, find s for each d collaboration	the similar the the spand the time the user in	rity between u ce complexity e complexities both user-b	for s for ased			understand
Keyword Drums Guitar Beat Orchestra Classical Like/ Dislike) Consider	the follo	g the cont	ent-based	recommen	ndation eve	tem "			CO2	Analyze
201 1 0 1 0 1 Like 202 0 0 1 0 1 Like 203 0 0 1 1 0 Dislike 204 1 1 0 1 1 Dislike 205 1 1 0 1 1 Dislike		Keyword	_	Guitar	Beat	Orchestra	Classical			4M	CO2	Remember, Understand, Apply
202 0 0 1 0 1 Like 203 0 0 1 1 0 Dislike 204 1 1 0 1 Dislike 205 1 1 0 1 Like		201	1	0	1	0	1					nders
203 0 0 1 1 0 Dislike 204 1 1 0 1 Dislike			0	0	1		1					land,
204 1 1 0 1 Dislike			0	0	1	1	0					Appl
205 1 1 0 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1			1	1	0	1	1					Y
the Payes Method, predict the class let 1.0		205	1	1		1	1	T '1				
I leing the Dayes have the class label for the following		Liging th	e Bayes	Method, 1	predict th	e class lab	el for the	following	tont			
samples: (i) $(1, 1, 1, 0, 1)$ (ii) $(1, 0, 1, 0, 0)$		Comples.	(i) (1	, 1, 1, 0, 1) (ii) (1, 0, 1, 0, 0))	onowing two	test		-	

	User Id	Movie_Id	Rating	User_Id	Movie_Id	Rating			
	1	101	3	6	104	5			
	2	102	4	5	101	1			
	4	102	2	5	102	5			
	3	101	5	1	103	3			
	2	103	1	5	104	2			
	3	102	1	1	104	2			
	6	102	2	3	103	5			
	4	103	4	4	104	5			
i)	Convert	the given da	taset into a	utility mat	rix.			2M	CO1
ii)	the siminatings of efficient	ne <i>cosine si</i> ilarities betonly. Based to the reco	ween <i>user</i> on these sommendati	4 and even imilarity va on systems	ery other u lues, which and why?	ser over measure	the raw	3M	C03
iii	for mov	he similarity vie 101 by ediction fun der the num	user 4 usi ction must	ng the suita use the sin	ible predict nilarity valu	ion function for the second se	tion, and	2M	CO3
rat	onsider a strings. Car	cenario when you menter filtering?	re two use	rs have only ssue with	y a small nu this scenar	imber of in us	ser-based		CO3



NATIONAL INSTITUTE OF TECHNOLGY PATNA DEPARTMENT OF COMPUTER SCIENCE AND ENGINEERING

Mid Semester Examination Jan - Mar 2024

M. Tech. (DS) 2nd Semester. Max.Marks: 30

Date: 21-03-24

Time: 2 Hrs. CS540221 – Big Data Analytics

Instructions: 1. Attempt all questions.

1. Attempt any suitable data, if necessary. (Any other Instruction need to provide by the concerned faculty)

Questions Questions	Marks	CO	BL
What is Big Data? Explain 3Vs of Big Data in detail.	[5]	CO1	Remember
a). What is HDFS? Expalin Read and write pipeline on HDFS for a large file. b). Explain MapReduce in detail with all its components.	[5+5]	CO2	Apply Understand Remember
Design MapReduce algorithms to take a very large file of integers and produce as output for the following. Also expain each algorithms with example.	[3*5]	CO2, CO3	Apply Understand
 a) The largest integer. b) The average of all the integers. c) The same set of integers, but with each integer appearing only once. d) The count of the number of distinct integers in the input. e) The set of prime numbers in the input file. 			

National Institute of Technology Patna

End Semester Exam, Session: July-Dec-2023

Program: M.Tech/Ph.D. Subject Name: Natural Language Processing Semester: 1st Department: CSE

Time: 2 hrs Subject Code: CS540201 Full Marks: 30

Assume any missing data and/or conditions. All questions are compulsory. The question paper is of two pages

1. How many operation will be required to convert Intention to execution. [Consider each operations to be of unit cost]. Show the appropriate locations of the operations. [6]

2. Given the following term document matrix and the number of words in each document, compute the TF*IDF score for each word/document. [6]

Term/Doc	Doc1	Doc2	Doc3	Dool	D5	D	-			
Car	2	2002	Docs	D004	Docs	Doc6	Doc7	Doc8	Doc9	Doc10
Cai	3	0	0	5	12	0	0	2	8	1
Auto	8	6	0	12	0	0	0	1	2	10
best	0	1	7	0	0	0	9	1	3	10
	40	1	1	0	1	5	12	0	0	0
Doc Size	40	22	15	38	29	19	47	10	25	26

- 3. Certain named entity recognition system models each word in the input text as one symbol in the following alphabet Σ : [9]
 - A Uppercase word (e.g. IBM)
 - C Capitalized word (e.g. John)
 - f Functional word (e.g. the, and, a, an, in, of, by, ...)
 - a Lowercase word (e.g. will)
 - 9 Number or code (e.g. 12)
 - p Punctuation (e.g., .:;)

For instance, the sentence: Tomorrow, John will be 12 years old. He likes music by Adam and the Ants. would be encoded as C p C a a 9 a a p C a a f C f f C p.

- a. Use the given sample to estimate the probabilities P(xy) and P(xyz) for each observed bigram/trigram.
- b. Compute the smoothing of the obtained probabilities using Laplace's Law. Give also the probability for unseen events.
- c. Compute the language model $P(z|xy) \ \forall x,y,z \in \Sigma$ that would result from using each of the two previous estimations. Compare the results, discussing which option is more suitable to model these sequences.
- 4. Papazom.com also needs to match offers from different suppliers that correspond to the same product, as well as to match user queries with product descriptions. For this, they asked us to propose a similarity model able to establish how similar two product description are. For instance, given the product descriptions. [9]
 - s₁ smartphone Hoewai x23-A with latest super AMOLED display and 64Gb
 - s₂ smartphone x23-A with 64Gb and AMOLED charge indicator
 - s_3 Hoewai smartphone z21-B with super AMOLED display and 32Gb
 - a. Represent each description as a word set, and compute $simjac(s_1, s_2)$, $simjac(s_1, s_3)$, and $simjac(s_2, s_3)$ using Jaccard similarity. Jaccard similarity is $simjac(x_1, x_2) = \frac{|x_1 \cap x_2|}{|x_1 \cup x_2|}$
 - b. Represent each description as a word-bigram set (i.e set elements are not single words, but word bigrams in the sentece), and compute $simcos(s_1, s_2)$, $simcos(s_1, s_3)$, and $simcos(s_2, s_3)$ using Cosine similarity. Cosine similarity is $simcos(x_1, x_2) = \frac{|x_1 \cap x_2|}{\sqrt{|x_1|}\sqrt{|x_2|}}$

c. A Papazon.com user wrote the search Hoewai smartphone AMOLED display. Compute the similarities of this query with s_1 , s_2 , and s_3 with each of the above metrics (unigram Jaccard and bigram Cosine).