



### UNIVERSITY OF COLOMBO SCHOOL OF COMPUTIN

BACHELOR OF SCIENCE IN COMPUTER SCIENCE
BACHELOR OF SCIENCE HONOURS IN COMPUTER SCIENCE
BACHELOR OF SCIENCE HONOURS IN SOFTWARE ENGINEERING

Academic Year 2021/2022 - Third Year Examination - Semester I

SCS3201 - Machine Learning and Neural Computing

TWO (2) HOURS (for both Part A & B)

- Part A -

126

To be completed by the candidate	
Index No:	

### **Important Instructions to candidates:**

- 1. The medium of instruction and question is English.
- 2. If a page or a part of this question paper is not printed, please inform the supervisor immediately.
- 3. Note that questions appear on both sides of the paper. If a page is not printed, please inform the supervisor immediately.
- 4. Write your index number on each and every page of the Question paper.
- Students are required to answer both Part A and Part B in two hours.
- Answer ALL questions. There are 03 questions in Part A and 01 Question in Part B of the paper.
- 7. Part A consists of 11 pages, and it carries 65 marks.
- Any electronic device capable of storing and retrieving text including electronic dictionaries and mobile phones are not allowed.
- 9. Non-Programmable calculators are allowed.

For	Examiner's	use
	only	

# Part A Question Marks 1 2 3 Total

														Index	k No:			• • • • • •
	2							Par	t A									
<u> Duesti</u>	on 1																	
a.	Differenti	ate R	egres:	sion a	$\operatorname{ind} C$	lassif	icatio	n tasl	cs in .	Supei	rvised	Мас	hine .	Learn	ing.			
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b.	Write a l	key d	iffere	nce i	n <i>Ur</i>	isupe	rvised	d Ma	chine	Lea	rning	con	npare	i to	Super	rvisea	l Ma	chine
	Learning	•														1	[03 M	[arlce]
	Answer:							······································	<del></del>			· · · · · · · · · · · · · · · · · · ·					[03 10]	iaiksj
	<u> </u>														· · · · · · · · · · · · · · · · · · ·			
c.	A few san	nples	of the	zoo	datas	et and	l its d	ata at	tribut	es are	e give	n belo	ow. T	he ori	ginal	datas	et co	nsists
	of 101 ar	nimals	s in s	seven	(07)	clas	ses (i	i.e., N	/Iamn	nal, I	Bird,	Repti	le, F	ish, A	Amph	ibian	, Bug	gand
	Invertebra	ite). A	nswe	r the	tollov	ving o	questi	ons u	sing t	he av	ailabi	e info	ormat	ion in	this o	datase	et.	
					T	<b>_</b>	T	Ι.		a)	T	S	T	T		Ι		6)
	Animal	hair	feathers	eggs	m ¥	airborne	aquatic	predator	toothed	backbone	breathes	/enomous	fins	legs	tail	domestic	catsize	Class type
	Name	ے ا	feat	ě	E	airb	adr	prec	too	ack	brea	eno	≔	<u>e</u>	ta	mo	cate	lass

antelope

frog

termite

penguin

crab

catfish

cheetah

	Index No:
	ata into labeled classes what are the attributes that you consider for your classification task? Briefly explain the reasons for
<u> </u>	[03 Mark
Answer:	
	only the given samples (07 samples in total). Write irreleval ributes you discarded above in the Question 1.c.i.) if there is a casens
in such a case. Bitery explain to	[03 Mark
Answer:	
separation of these classes into	different clusters. What is the most reasonable value for K in
	different clusters. What is the most reasonable value for K in explain your justification.
separation of these classes into	different clusters. What is the most reasonable value for K in explain your justification.
separation of these classes into Means algorithm to try? Briefly	different clusters. What is the most reasonable value for K in explain your justification.
separation of these classes into Means algorithm to try? Briefly	different clusters. What is the most reasonable value for K in explain your justification.
separation of these classes into Means algorithm to try? Briefly	different clusters. What is the most reasonable value for K in explain your justification.
separation of these classes into Means algorithm to try? Briefly	different clusters. What is the most reasonable value for K in explain your justification.
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separation of these classes into Means algorithm to try? Briefly	different clusters. What is the most reasonable value for K in explain your justification.
separation of these classes into Means algorithm to try? Briefly	luster the entire dataset with <b>K-Means</b> algorithm to investigate different clusters. What is the most reasonable value for K in explain your justification.  [04 Mark
separation of these classes into Means algorithm to try? Briefly	different clusters. What is the most reasonable value for K in explain your justification.

What is the purpose of a <i>Cost</i>	Function in a supervi	sed learning task	Index No:	• • • • • • • • • • • • • • • • • • • •
	*			[03 Mar
Answer:				
,			******	
Assume a student has run a configuration except the learn three times and the correspon	ing rate. The student a	applied three diff	ferent learning rate	$\alpha_1, \alpha_2, \alpha_3$
With Learning Rate = $\alpha_1$		ng Rate = $\alpha_2$		
With Boarning Rate - 01	with Leanin	ig Rate – $u_2$	With Learn	ning Rate = $\alpha_3$
Cost	Cost		Cost	
		_		
Number of Iterations	Number of It	erations	Number of	Iterations
Which of the three to be the				
Vhich of the three learning ra our reasoning behind the jud	ites $(\alpha_1, \alpha_2, \alpha_3)$ has the general section $\alpha_2$ and $\alpha_3$	he highest and le	owest magnitude?	Briefly expl
	Somonio.			[06 Mar
Answer:				•
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Index No:		
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### Question 2

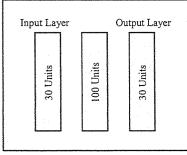
a. Typically *Mean Squared Error* is not used as an evaluation metric in Logistic Regression. Why does Mean Squared Error is inappropriate as an evaluation metric in Logistic Regression?

Answer: [03 Marks]

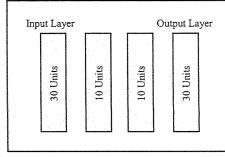
b. It is a known fact that the *Gradient Descent algorithm* may lead to a local optimum during the learning process. With this risk of ending up in a sub-optimal solution, how appropriate to use the *Gradient Decent algorithm* during the learning process when you have a Squared Error function as the cost function? Justify your answer.

	[03 Marks]
Answer:	

c. Consider the following two (02) fully connected feed forward artificial neural network architectures. The architecture 1 has three (03) layers and Architecture 2 has four (04) layers.



Architecture I



Architecture 2

					ndex No:	
Which architec considerations?	ture will lead to a me Explain your answe	ore complex tr er with justific	aining process ations.	with respec	t to the time	and memory
						[04 Marks]
Answer:						
***************						
				**********		
immediately be	of the output layer $a_0$ . The cost function $w_1$ and $w_2$ and $w_3$	. For a given	input $x$ with the s $E$ which is a $\frac{7}{4}$	e expected	value of v	the neuron O
rule for weight p	ming rate of $\alpha$ and parameter $W_4$ accordance	using appropri	iate derivative dient descent c	terms, defin	ne the weigh zation proced	it adjustment dure. [02 Marks]
Answer:						[02 Marks]
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d.

	[05 Mar.
Answer:	
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validation data for testing the model, and the remaining $k-1$ subsets cross-validation process is then repeated $k$ times, with each of the $k$ survalidation data. The $k$ results can then be averaged to produce a single $k$	are used as training data. Itsets used exactly once as
validation data for testing the model, and the remaining $k-1$ subsets cross-validation process is then repeated $k$ times, with each of the $k$ survalidation data. The $k$ results can then be averaged to produce a single $k$	are used as training data. Tubsets used exactly once as estimation.
validation data for testing the model, and the remaining $k-1$ subsets cross-validation process is then repeated $k$ times, with each of the $k$ substantial validation data. The $k$ results can then be averaged to produce a single of $k$ -fold cross-validation.	are used as training data. The subsets used exactly once as estimation.
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e. Using appropriate derivative terms and notations, explain how you are going to derive the necessary

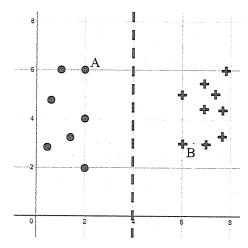
Index No: .....

# Question 3

a. Briefly explain the use of Laplace smoothing in Naïve Bayes classifier.

	[03 Marks]
Answer:	
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b. Assume a hard-margin SVM finds its decision boundary as given in the dashed line below in the diagram.



Draw the new decision boundary in the following plot, if the datapoint A is removed from the dataset and the SVM is retrained.

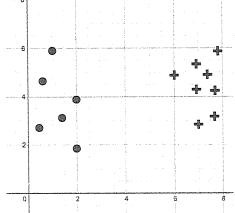


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c. Let's assume the datapoint B is also removed from the dataset after removing the datapoint A in Question 3.b. Draw the new decision boundary if you retrain the SVM with the new dataset.





d. Consider the following table to answer subsequent questions in Question 3.d. The table shows a dataset of COVID-19 test results of 10 individuals against three attributes.

#	Cold -	Temperature	Fatigue	Result
1	No	Normal	No	- Negative
2	No	High	Yes	- Negative
3	No	High	Yes	+ Positive
4	No	Very High	Yes	+ Positive
5	Yes	Normal	No	- Negative
6	Yes	Normal	Yes	+ Positive
7	Yes	High	No	- Negative
. 8	Yes	High	Yes	+ Positive
9	Yes	Very High	No	+ Positive
10	Yes	Very High	Yes	+ Positive

i. Calculate the *Entropy* for the dataset given above considering the target variable Result. Show your calculations clearly.

[03 Marks]

Answer:	
	:

The same of the sa	Index No:
• Entropy(Result   Cold) = 0.93	te, it has been decided to measure the <i>Entropy</i> for <i>Intropy</i> for two cases as given below.
• Entropy(Result   Fatigue) =0.70	
Find the third case Entropy(Result   Temperature) b	y showing your calculation steps clearly.
	[03 Marks]
Answer:	
ii. Calculate the <i>Information Gain</i> for <b>all three cases</b> your calculations clearly.	
ii. Calculate the <i>Information Gain</i> for <b>all three cases</b> your calculations clearly.  Answer:	that were mentioned in the <b>Question 3.d.ii.</b> Show
your calculations clearly.	
Answer:	
your calculations clearly.	
Answer:	

	Index No:
v. Based on the above calculations in Question 3.d, what is the feature split in decision tree construction. State your reasons clearly.	that will be used as the first
	[02 Marks]
Answer:	
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126

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	only	

# Part B Question Marks 4 Total

# Part B

# Question 4

a.	Suppose the two	fuzzy sets F and	G given below	with $X = \{1$	2 3 4 53

$$F = \left\{ \frac{1}{1} + \frac{0.5}{2} + \frac{0.25}{3} + \frac{0.1}{4} \right\}$$

$$G = \left\{ \frac{0.5}{1} + \frac{1}{2} + \frac{0.5}{3} + \frac{1}{4} + \frac{0.5}{5} \right\}$$

Answer the following.

(i) What is G'?

	[02 Marks]
Answer:	
(ii) What is $F \cap G$ ?	
	[02 Marks]
Answer:	
(iii) What is $F - G$ ?	
	[02 Marks]
Answer:	
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b. Consider the following Octave implementation of Mamdani FIS which computes the percentage of investments to be allocated to stocks using four rules.

```
[System]
                                        [Input2]
Name='Investment-Portfolio'
                                        Name='Risk-Tolerance'
Type='mamdani'
                                        Range=[0 10]
Version=2.0
                                        NumMFs=2
NumInputs=2
                                        MFl='Low':'trapmf', [-1 0 2 8]
NumCutputs=1
                                        MF2='High': 'trapmf', [2 8 10 11]
NumRules=4
AndMethod='min'
                                        [Output1]
OrMethod='max'
                                        Name='Percentage-In-Stocks'
ImpMethod='min'
                                        Range=[0 100]
AggMethod='max'
                                        NumMFs=3
DefuzzMethod='gentroid'
                                        MFl='About-Fifteen': 'trimf', [-10 15 40]
                                        MF2='About-Fifty':'trimf',[25 50 75]
[Input1]
                                        MF3='About-Eighty-Five':'trimf',[60 85 110]
Name='Age'
Range=[20 100]
                                        [Rules]
NumMFs=2
                                        1 2, 3 (1) : 2
MFl='Young':'trapmf',[-21 20 40 80]
                                        2 1, 1 (1) : 2
MF2='Old':'trapmf',[40 80 100 101]
                                        -2 -1, 2 (0.5) : 1
                                        -1 -2, 2 (0.5) : 1
```

### The four rules are:

- 1. If (Age is Young) or (Risk-Tolerance is High), then (Percentage-In-Stocks is About-Eighty-Five) (1)
- 2. If (Age is Old) or (Risk-Tolerance is Low), then (Percentage-In-Stocks is About-Fifteen) (1)
- 3. If (Age isn't Old) and (Risk-Tolerance isn't Low), then (Percentage-In-Stocks is About-Fifty) (0.5000)
- 4. If (Age isn't Young) and (Risk-Tolerance isn't High), then (Percentage-In-Stocks is About-Fifty) (0.5000)

Where the number inside the parenthesis represents the rule weight.

Answer the following questions.

(i) List down the five (5) steps in the Mamdani fuzzy inference process.

[05 Marks]

Answer:	

(ii)	Plot the membership functions of the two inputs and the output with labels and	legends. [06 Marks]
Answe	er:	
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		And Andrews (Andrews (Andrew) (Andrews (Andrews (Andrews (Andrews (Andrews (Andrews (Andrews
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Index No:
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(iii) Calculate the approximate value for *Percentage-In-Stocks* when the two inputs *Age* and *Risk-Tolerance* are 40 and 7, respectively. Give the aggregation graph. State any assumptions you make.

[08 Marks]

Answer:

				Index No:	••••••
What are the ty	pical elements of a 1	einforcement lear	ming system?		
					[03 Marks]
Answer:		***************************************	***************************************		[05 1.141.15]
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		*			
<del>~</del>					
Consider the fo	llowing grid world p	roblem.			
S	+5				
<u> </u>	·				
in state (1,1), n	rid squares, identification arked with the letter agent can move in f	r S. The agent re	eceives a reward of	of +5 if it reaches	gent always s the terminal
Give the <b>rewar</b>	d table for this prob	lem. State any as	sumptions you ma	ake.	
		·	• •		[07 Ma
nswer:					
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