

Coversheet for centrally scheduled online non-invigilated exams

Exam information	
Course code and name	COMP4702 / COMP7703 / DATA7703
Course code and name	Machine Learning / Machine Learning for Data Scientists
Semester	Semester 1, 2020
Exam type	Online, non-invigilated
Exam date and time	Please refer to your personalised timetable
Exam duration	Working time (180 minutes) + additional online allowance: 30 minutes = TOTAL exam duration: 3 hrs 30 minutes within a 24 hour window
Exam window	You have a 24 hour window in which you must complete your exam. Once you open your test you will have the duration of the exam to complete and submit/upload your response. The latest possible starting time is 180 minutes prior to the end of this 24 hours.
Late submission of exams	If you have a technical problem, you should collect evidence (photo/screenshot if possible) and make a statement about the incident.
	You should submit as soon as the problem is alleviated. Should the problem extend beyond 30 minutes, you should gather evidence at regular intervals (e.g. a screenshot every 1 hour).
	If you do not submit the exam on time, or within the approved extended time, you will receive '0' mark, unless there is sufficient evidence that the late submission is beyond your control. Without a reason preventing submission there is a 100% penalty.
	Course coordinators decide on the evidence presented whether they will accept a late submission.
	You can appeal to the Associate Dean (Academic).
Reading time	Reading time has not been formally allocated for online exams, however students are encouraged to review and plan their approach for the exam before they start. The total exam time should be sufficient to do this.
Additional time	30 minutes additional time has been incorporated in recognition of the online environment and the different circumstances that students face in their home environments. This includes allowances for network or connection issues.
Weighting	COMP4702 - This exam is weighted at 45% of your total mark for this course. COMP7703 - This exam is weighted at 40% of your total mark for this course.
	DATA7703 - This exam is weighted at 45% of your total mark for this course.
Permitted materials	This is an open book exam – all materials permitted.

	Laptop or desktop computer with software used during the course. Scientific
Daniela Harrisania I. I.	calculator or calculator app.
Required/recommended materials	Paper and writing implements for rough working. If the student chooses to complete hand-written answers for some of the exam, scanning technology (e.g. mobile phone, camera, scanner) will be required to include this material in a pdf submission.
	Answer all questions.
	Part A is to be completed in the test tool.
Instructions	Part B is to be completed in the test tool.
	Part C is to be completed in a document, and uploaded to the Assignment link for the exam.
	Given students may not all undertake the online exam at the same time, or in the same time zone, and that some questions may be randomised, responding to student queries and/or relaying corrections to exam content during the exam will not be feasible.
Who to contact	If you need to make assumptions or have any queries or concerns about a particular exam question, state these at the start of your solution to that question. You may also include queries you may have made with respect to a particular question, should you have been able to 'raise your hand' in an examination room.
	If you experience any technical difficulties during the exam, contact the <u>Library AskUs</u> service for advice. You should also ask for an email documenting the advice provided so you can provide this to the course coordinator immediately at <u>marcusg@itee.uq.edu.au</u> .
	This is an open book exam. You will have access to your own notes, course texts, and other materials.
	The normal academic integrity rules apply.
	You cannot cut-and-paste material other than your own work as answers.
	You are not permitted to consult any other person — whether directly, online, or through any other means — about any aspect of this
Important exam condition information	assessment during the period that this assessment is available. If it is found that you have given or sought outside assistance with this assessment then that will be deemed to be cheating and will result in disciplinary action.
	By undertaking this online assessment you will be deemed to have acknowledged <u>UQ's academic integrity pledge</u> to have made the following declaration:
	"I certify that my submitted answers are entirely my own work and that I have
	neither given nor received any unauthorised assistance on this assessment item".

scription	COMP4702/7703/DATA7703 - Machine Learning		
Ser. (pero.)	Semester One 2020 Final Examination		
structions	Answer all questions. There are 60 marks in total.		
med Test	This test has a time limit of 3 hours and 30 minutes. You will be notified when time expires, and you may conti Warnings appear when half the time, 5 minutes, 1 minute, and 30 seconds remain. [The timer does not appear		
	ots Not allowed. This test can only be taken once. On This test can be saved and resumed later. The timer will continue to run if you leave the test.		
QUESTIC	ON 1	2 points	Save Answer
Regarding	the k-nearest neighbour (kNN) classifier, which of the following statements is incorrect ?		
0 0	en values of k, a majority vote can be used directly to determine the final class prediction.		
	a non-parametric model.		
	sometimes called a "lazy learning" method because no training is required.		
	noice of a distance metric is an important factor in applying kNN.		
O THE OH	loide of a distance metric is an important factor in applying kiviv.		
QUESTIC	DN 2	2 points	Save Answer
When Prince	cipal Component Analysis (PCA) is performed by calculating the eigenvectors and eigenvalues of ance matrix of a dataset, which of the following statements is true?	2 points	Save Answer
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When Pring the covaria	cipal Component Analysis (PCA) is performed by calculating the eigenvectors and eigenvalues of ance matrix of a dataset, which of the following statements is true?	2 points	Save Answer
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When Print the covariation of the print of the eigenvalue of the area of the print of the area of the print o	cipal Component Analysis (PCA) is performed by calculating the eigenvectors and eigenvalues of ance matrix of a dataset, which of the following statements is true? Incipal components are always orthogonal to each other in the transformed feature space, genvalues always sum to equal 1. Inining more principal components the amount of variance explained in the projection is strictly increases. Bea under the scree graph is proportional to the computational complexity of solving the eigensystem. BON 3 The following is not a hyperparameter or experimental factor when training neural networks? Brobability coefficients. Brossize. In initialization range/distribution. BON 4 The fised learning model with high complexity (e.g. the number of parameters in a polynomial model) is trained on a relatively small dataset (i.e. number of data points) the it is most likely to ollowing characteristics:	2 points	Save Answer
When Print the covariation of the print of the eigenvalue of the area of the print	cipal Component Analysis (PCA) is performed by calculating the eigenvectors and eigenvalues of ance matrix of a dataset, which of the following statements is true? Incipal components are always orthogonal to each other in the transformed feature space. It is incipal components to equal 1. In ining more principal components the amount of variance explained in the projection is strictly increases. It is a under the scree graph is proportional to the computational complexity of solving the eigensystem. In a polymomial is not a hyperparameter or experimental factor when training neural networks? In a polymomial is not a hyperparameter or experimental factor when training neural networks? In a polymomial is initialization range/distribution. In a polynomial model with high complexity (e.g. the number of parameters in a polynomial model) is trained on a relatively small dataset (i.e. number of data points) the it is most likely to ollowing characteristics: It is a polynomial model with variance, low training error, high test error.	2 points	Save Answer

QUESTION 5		2 points	Save Answer
Match the following key terms in k-means clumixture model using the EM algorithm:	ustering with their closest corresponding term in fitting a Gaussian		
- • Mean vector.	A. Cluster center.		
- • Data point ownership.	B. Responsibility value.		
- V Local optimization.	C. Expectation maximization.		
- Covariance matrix.	D. Distance metric.		
QUESTION 6		2 points	Save Answer
Support Vector Machines:		- F	
• •	points which fall inside or on the incorrect side of the margin.		
Are trained using stochastic gradient			
	mizing the margin between the discriminant and the training points.		
	e likelihood of each datapoint in the training set.		
Ose a remending to determine the	o inclinious of cault datapoint in the training sec.		
OUESTION 7			
QUESTION 7		2 points	Save Answer
A Bayesian network is designed which im $P(A,B,C,D,E,F,G) = P(A)P(B)P(C A)P(B)$	plements the following probability distribution for a given problem:		
The number of edges in the graph represe	entation of the network is:		
QUESTION 8		2 points	Save Answer
QUESTION 8		2 points	Save Answer
Calculate the overall error rate (as a pe	ercentage, to 2 decimal places) given by the following confusion m		
Calculate the overall error rate (as a pe	ercentage, to 2 decimal places) given by the following confusion m redicted to belong to class i that actually belong to class j:		
Calculate the overall error rate (as a peindicates the number of data points per	ercentage, to 2 decimal places) given by the following confusion m redicted to belong to class i that actually belong to class j:		
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Given the following matrices:

$$H = \begin{pmatrix} 3 & 1 & 4 & 3 \\ 2 & 1 & 4 & 3 \\ 1 & 4 & 1 & 3 \\ 2 & 2 & 3 & 1 \end{pmatrix}, K = \begin{pmatrix} 1 & -1 \\ -1 & 2 \end{pmatrix}$$

where H is the 4×4 input data (e.g. pixel values) and K is a 2×2 kernel, perform a convolution operation (as it is done in convolutional neural networks) on H using K, with a stride of 2 and padding of 1 all around H.

$$\Box \begin{pmatrix} 2 & 3 \\ -1 & -3 \end{pmatrix}$$

$$\begin{array}{c|cccc}
 & 6 & 7 & -3 \\
 & 0 & -5 & 0 \\
 & -2 & -1 & 1
\end{array}$$

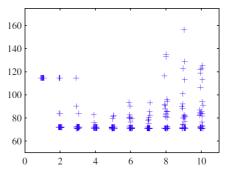
c.

QUESTION 11 5 points Save Answer

In the textbook by Chris Bishop, the following Figure is presented:

- a. What do you think this graph is showing about the training of a multi-layer perceptron? Discuss in a few sentences. (3 marks)
- b. Bishop states that the lowest validation error obtained over all the training runs was for 8 hidden units (he says "validation" in the text but I assume he means the test error values in this graph). Based on the results shown, what value would you choose for the number of hidden units and why? (2 marks)

Figure 5.10 Plot of the sum-of-squares test-set error for the polynomial data set versus the number of hidden units in the network, with 30 random starts for each network size, showing the effect of local minima. For each new start, the weight vector was initialized by sampling from an isotropic Gaussian distribution having a mean of zero and a variance of 10.



5 points

Save Answer

The Figure below is taken from the Hastie et al. textbook. The Bayes error rate here is the lowest possible error that could be achieved for this particular example.

While this Figure presents a specific example, an important, general point can be drawn from it regarding the setting of hyperparameters in Machine Learning models. Discuss in your own words what this general point is by referring to the Figure. (5 marks)

Test Error Curves - SVM with Radial Kernel

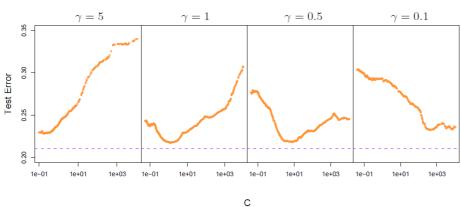


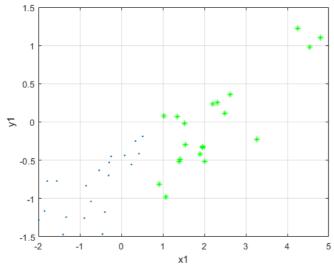
FIGURE 12.6. Test-error curves as a function of the cost parameter C for the radial-kernel SVM classifier on the mixture data. At the top of each plot is the scale parameter γ for the radial kernel: $K_{\gamma}(x,y) = \exp{-\gamma||x-y||^2}$. The optimal value for C depends quite strongly on the scale of the kernel. The Bayes error rate is indicated by the broken horizontal lines.

QUESTION 13

5 points

Save Answer

The following Figure shows a set of data for a classification problem:



Consider using rectangles as a model to classify this data, where a rectangle is specified by [a<=x1<=b; c<=x2<=d]. A rectangle predicts that a point belongs to its designated class if the point lies within or on the perimeter of the rectangle.

- a. Specify a set of model parameter values that will perfectly classify all of the data shown for both classes. (2 marks)
- b. Give an example of a test data point (feature values and class label) that would be incorrectly classified by the models you have specified. (1 mark)
- c. Consider performing logistic regression on the data from Figure 1. How many parameters would the trained linear regression model have? (2 marks)

QUESTION 14 Save Answer

The Figure below is from the textbook by Chris Bishop:

a. Which plot do you think is most representative of overfitting? Explain your choice. (3 marks)

b. Referring to the two green points towards the top of the plots, explain how the K-nearest-neighbour algorithm can be said to be robust to noise for large values of K. (2 marks)

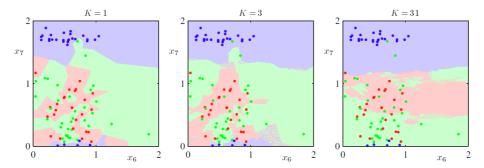


Figure 2.28 Plot of 200 data points from the oil data set showing values of x_6 plotted against x_7 , where the red, green, and blue points correspond to the 'laminar', 'annular', and 'homogeneous' classes, respectively. Also shown are the classifications of the input space given by the K-nearest-neighbour algorithm for various values of K.

QUESTION 15 20 points Save Answer

In the Practical classess for this course, you have implemented and/or applied a range of different machine learning models and algorithms. For this question, you are required to demonstrate your ability to apply one or more of these techniques to a supplied dataset.

Key points:

- You are expected to spend about 60 minutes on this question.
- You can use any code that you developed during the pracs, as well as built-in Matlab and python
 functions/libraries. If you use other libraries then you must reference them and they must be publicly
 available.
- Explain key steps of your analysis (data preprocessing, training, testing, visualization, etc.).
- Present any results and output from your work. Add comments and discussion to demonstrate that you understand
 the results and output.
- State any assumptions you make.
- Don't spend time introducing background concepts or describing the theory of models from the course.
- Make sure your work is understandable and readable but don't spend a lot of time making the presentation neat and polished.

Your answer for this question should be uploaded as a pdf file in the turnitin submission for the exam.