SEMESTER 2 EXAMINATION 2017/2018

ADVANCED MACHINE LEARNING

Duration: 120 mins

You must enter your Student ID and your ISS login ID (as a cross-check) on this page. You must not write your name anywhere on the paper.

	Question	n Marks
	A1	
Student ID:	B1	
	B2	
ISS ID:	B3	
	Total	

Answer all parts of the question in section A (30 marks) and TWO questions from section B (35 marks each)

This examination is worth 60%. The coursework was worth 40%.

University approved calculators MAY be used.

A foreign language translation dictionary (paper version) is permitted provided it contains no notes, additions or annotations.

Each answer must be completely contained within the box under the corresponding question. No credit will be given for answers presented elsewhere.

You are advised to write using a soft pencil so that you may readily correct mistakes with an eraser.

You may use a blue book for scratch—it will be discarded without being looked at.

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Section A

Question A 1

(a) Briefly describe the type of data where the following learning machines excel: (i) SVMs, (ii) Gradient Boosting and (iii) CNNs. (6 marks)

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$$\mathbf{x} = (x_1, x_2, x_3) \to \vec{\phi}(\mathbf{x}) = (x_1^2, x_2^2, x_3^2, \sqrt{2} x_1 x_2, \sqrt{2} x_1 x_3, \sqrt{2} x_2 x_3)$$

the kernel $K(x, y) = \vec{\phi}(x) \cdot \vec{\phi}(y)$ is equal to $(x \cdot y)^2$. (4 marks)

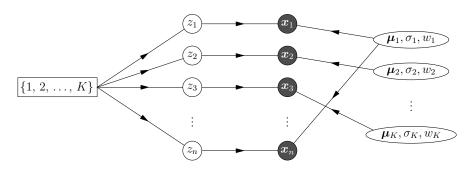
(c) Briefly describe the *random forest* algorithm. Explain why it is often very successful. (5 marks)

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(d) Describe the difficulty of training a many layer multi-layer perceptron.	(5 marks)

$$f(\mathcal{D}|\{z_1, z_2, \ldots, z_n\}) = \prod_{i=1}^n w_{z_i} \mathcal{N}(\boldsymbol{x}_i|\boldsymbol{\mu}_{z_i}, \sigma_{z_i} \mathbf{I}).$$

This can be represented by a graphical model



Draw the equivalent diagram using the plate notation.	(5 marks)
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(f) Show that the beta distribution $\operatorname{Beta}(p|a,b) = p^{a-1} (1-p)^{b-1}/B(a,b)$ is a conjugate prior to the binomial likelihood $\operatorname{Binom}(k|n,p) = \binom{n}{k} p^k (1-p)^{n-k}$. Derive update equations for the parameters of the posterior distribution after observing k successes and n-k failures.. (5 marks)



End of question A1

Q1: (a)
$$\frac{}{6}$$
 (b) $\frac{}{4}$ (c) $\frac{}{5}$ (d) $\frac{}{5}$ (e) $\frac{}{5}$ (f) $\frac{}{5}$ Total $\frac{}{30}$

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Section B

Question B 1

 (a) Explain why choosing the maximum margin dividing plane to the success of SVMs. 	e is so important (5 marks)
b) Sketch how slack variables, ξ_k , are introduced to allow sor lie within the margins.	ne data points to (5 marks)

(c)	Show	
	(i) how the constraints $y_k\left({{m w}^{\sf T}}{m x}_k - b \right) \ge 1$ are changed by introducing the slack variables	
	(ii) how to modify the cost function $\frac{1}{2} \ oldsymbol{w} \ ^2$	
	(iii) the constraints on the slack variables.	
	Describe all the terms used. (5 marks)	
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	Show how the Lagrangian, \mathcal{L} is modified to include the slack variables and give the constraints on any Lagrange multipliers (5 marks)	
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(e) By minimising with respect to the slack variables (i.e. setting $\frac{\partial \mathcal{L}}{\partial \xi_i} = 0$) obtain new constraints for the Lagrange multipliers α_i (5 marks)	
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(f) Write down the general from for (i) a polynomial kernel and (ii) the radial	
basis function kernel (5 marks)	
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(g) Explain why it is important that a kernel is positive semi-definite and give three properties that a positive semi-definite kernel should have. (5 marks)

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End of question B1

Q1: (a)
$$\frac{}{5}$$
 (b) $\frac{}{5}$ (c) $\frac{}{5}$ (d) $\frac{}{5}$ (e) $\frac{}{5}$ (f) $\frac{}{5}$ (g) $\frac{}{5}$ Total $\frac{}{35}$

Question B 2

 a) Sketch a typical CNN from taking in inputs to making sion. Label the layers used. 	(5 marks)
 b) Briefly explain the following terms i) filters ii) feature m iv) max pooling v) fully connected layer 	aps iii) weight sharing <i>(5 marks)</i>
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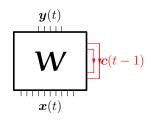
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(c)	Explain what is meant by i) Stochastic Gradient Descent ii)	momentum in
	the context of learning and iii) mini-batches.	(5 marks)

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(d) F	Briefly describe the motivation behind the design of Long-Short Term Mem-
(4)	ory (LSTM) units and how they achieve this. (5 marks)

(e) Consider a recurrent neural network with memory states ${m c}(t)$ as shown below.



Sketch how we can unroll the network in time to learn a sequence

$$(x(1), y(1)), (x(2), y(2)), \dots, (x(4), y(4)).$$

(5 marks)

 f) Explain what linear embedding units do and why they are performing machine learning on languages. 	so important in <i>(5 marks)</i>
) Briefly explain the typical preprocessing steps that are carr ments before the data is feed into a learning machine.	ed out on docu- (5 marks)
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End of question B2

Q2: (a)
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 (b) $\frac{}{5}$ (c) $\frac{}{5}$ (d) $\frac{}{5}$ (e) $\frac{}{5}$ (f) $\frac{}{5}$ (g) $\frac{}{5}$ Total $\frac{}{35}$

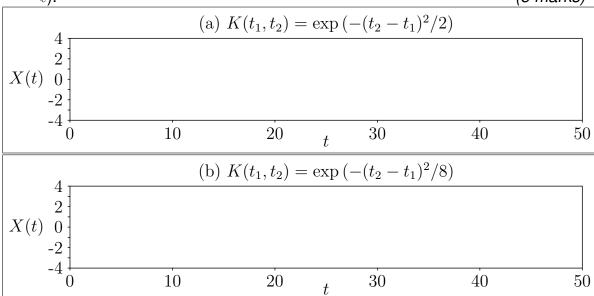
Question B 3

(a) Explain for Gaussian Processes (GP) what is the prior, the likelihood and the posterior. (5 marks)	
	-
(b) Explain what the kernel function represents and how it could be measured empirically from many observations. (5 marks)	
	1

(c) Consider a 1-d Gaussian Process, $\boldsymbol{X}(t)$, with a kernel of the form

$$K(t_1, t_2) = \exp\left(-\frac{(t_2 - t_1)^2}{2\ell}\right).$$

Sketch three Gaussian Processes drawn from the prior with (a) $\ell=1$ and (b) $\ell=2$ (we are not looking for accuracy, but rather the effect of changing ℓ). (5 marks)



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than a full Bayesian solution.	AP solution rather (5 marks)

) Explain why Monte Carlo techniques a ence problems.			(5 marks)
Briefly describe in words the use of the ence.	e MCMC algo	orithm in Ba	yesian infer- (5 marks)
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f) Briefly describe in words the use of the ence.	e MCMC algo	orithm in Ba	yesian infer- (5 marks)

(g) When are probabilistic methods likely to give good results and what is the hurdle in using it?

(5 marks)



End of question B3

Q3: (a)
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 (b) $\frac{}{5}$ (c) $\frac{}{5}$ (d) $\frac{}{5}$ (e) $\frac{}{5}$ (f) $\frac{}{5}$ (g) $\frac{}{5}$ Total $\frac{}{35}$