Surname: _____

Given Name: _____

Student Number ____

- This examination paper includes 9 pages (including this cover page and the last blank page) and 5 questions. You are responsible for ensuring that your copy of the papers is complete. Bring any discrepancy to the attention of your invigilator.
- Examination duration is 120 minutes.

Exam Instructions:

• SINGLE VERSION exam

Materials Permitted in the Exam Venue:

- No electronic aids are permitted, e.g. laptops, phones
- McMaster Standard Calculator (Casio FX-991MS/MS+)

Materials to be supplied to the students:

• Scrap paper

Instructions to the students:

• If you think there is an issue with one of the questions, make the best sensible assumption and write your assumption down along with your solution.

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Question	Points	Score
1	45	
2	15	
3	15	
4	10	
5	15	
Total:	100	

- 1. For each multiple choice question, there is only one correct answer. Mark/check your choice clearly.
 - (a) (5 points) We have a labeled data set $(x^i, y^i)_{i=1}^m$ of m points where $y^i \in \{-1, +1\}$. The following is the optimization formulation is for which model?

$$\min_{W} \left(\lambda \|W\|_{2}^{2} + \frac{1}{m} \sum_{i=1}^{m} \max\{0, 1 - y^{i} W^{T} x^{i}\} \right)$$

- Regularized Least Squares
- \bigcirc Hard Margin SVM
- \bigcirc Soft Margin SVM
- Perceptron
- (b) (5 points) Which one is the most reasonable choice for linear classification when data is not linearly separable?
 - \bigcirc Using linear programming
 - \bigcirc Hard Margin SVM
 - \bigcirc Soft Margin SVM
 - Perceptron
- (c) (5 points) Let TP, TN, FP, FN be the number of true positives, true negatives, false positives, and false negatives respectively. The "precision" of a model is defined by
 - $\bigcirc \frac{TP}{TP+TN+FP+FN}$
 - $\bigcirc \frac{TP}{TP+FP}$
 - $\bigcirc \frac{TP}{TP+TN}$
 - $\bigcirc TP+TN$
 - $\bigcirc \frac{1}{TP+TN+FP+FN}$
 - $\bigcirc \frac{TP+FP}{TP+TN+FP+FN}$
- (d) (5 points) Which one of these choices is the most accurate about multiclass SVM?
 - The number of linear classifiers trained for one-versus-all is more than the number of linear classifiers trained for the all pairs strategy
 - \bigcirc There is an end-to-end version of all-pairs SVM that can be trained in end-to-end fashion
 - There is an end-to-end version of one-versus-all SVM that can be trained in an end-to-end fashion

- (e) (5 points) Which one is more accurate?
 - O Universal approximation theorem indicates that you need RELU activation function rather than sigmoid to approximate any function with neural networks.
 - O Any bounded continuous function can be approximated (with any desired accuracy) with a neural network with one layer (i.e., one input layer and no hidden layers).
 - Any sigmoid neural network with 10 layers and any number of parameters can be approximated by a sigmoid network with only 1 hidden layer.
- (f) (5 points) Which is more accurate regarding gradient descent (GD) and stochastic gradient descent (SGD)?
 - Unlike GD, SGD does not require backpropagation since the gradient is computed as the sum of gradients of individual data points.
 - \bigcirc SGD is more likely to get stuck in a local minimum since each update only uses a few data points.
 - \bigcirc SGD makes more updates in an epoch (compared to GD)
- (g) (5 points) Which one is NOT a usual strategy for improving the "vanishing gradient" issue?
 - Using backpropagation for computing the gradient
 - Using RELU activations rather than sigmoid
 - \bigcirc Using a residual architecture (ResNET)
- (h) (5 points) Which one is NOT a usual strategy for regularization in neural networks?
 - \bigcirc Early stopping
 - ⊖ Dropout
 - \bigcirc Weight sharing
 - Using a ResNET architecture
- (i) (5 points) We apply a 3x3 convolution kernel to a 10x10 image with no padding, and stride=2. What would be the size of the output image?

- 2. (15 points) Clearly write TRUE or FALSE besides each choice.
 - Boosting is likely to reduce overfitting.
 - \bigcirc Logistic regression is often used for classification rather than regression.
 - \bigcirc Random Forests combine the idea of boosting and bagging
 - \bigcirc Vapnik-Chrvonenkis dimension (VC dimension) of the class of all linear classifiers in \mathbb{R}^d is infinite.
 - \bigcirc We have trained a fully connected neural network for classifying images and observe that its accuracy on test data is 50%. We would like to improve this to 70%. We also observe that the accuracy on training data is 85%. A sensible solution is to use dropout.

3. In this questions we want to study the relationship between smoking and having a heart attack.

Assume that the ratio of people who smoke to the whole population is C (in other words, $\Pr[x \text{ is a smoker}] = C$ where x is a random person generated from the population).

Assume that the probability of heart attack for smokers during their lifetime is 0.3, but this probability is 0.1 for non-smokers.

(a) (5 points) What is the probability of having a heart-attack for a randomly selected individual (uniform over the whole population, including smokers and nonsmokers) assuming C = 0.3? Show your work and simplify the final solution.

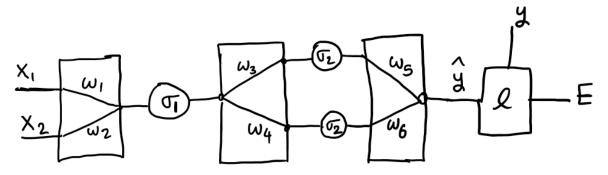
(b) (10 points) We observe that from a randomly selected set of 100 people, 20 of them have experienced heart attack during their lifetime. What is the maximum likelihood estimate for the value of C in this case? Show your work and simplify the final solution.

4. (10 points) We are trying to fit the homogeneous line y = a.x to a non-linear curve, $f(x) = x^2 + 1$ where $a, x, y \in \mathbb{R}$.

Assume that the x values are uniformly distributed on [0, 1]. What should we pick for the value of a in order to minimize the squared error? Show your work and simplify your final answer.

- 5. (15 points) The following neural network has three layers and is used for regression. Here are the details:
 - 1. The activation functions of the first and second layers are ReLU (i.e., $\sigma_1(x) = \sigma_2(x) = max(0, x)$) but there is no activation function in the third layer.
 - 2. The loss function is the squared loss: $l(y, \hat{y}) = (y \hat{y})^2$.
 - 3. $w_1 = 1, w_2 = 1, w_3 = 0.1, w_4 = 0.2, w_5 = -2, w_6 = 5.$
 - 4. The given input is specified by $x = [x_1, x_2]^T = [5, 5]^T, y = 10$

Compute the partial derivative of error E with respect to w_1 for the given input point $(x = [5, 5]^T, y = 10)$. Show your work.



The End