# Faculty of Engineering, Mathematics and Science

## **School of Computer Science & Statistics**

**Integrated Computer Science Programme Year 3** 

Hilary Term 2018

ST3009: Statistical Methods for Computer Science

DD MMM YYYY Venue 00.00 – 00.00

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### **Instructions to Candidates:**

Attempt **all** questions.

You may not start this examination until you are instructed to do so by the invigilator.

## **Materials Permitted for this examination:**

Non-programmable calculators are permitted for this examination – please indicate the make and model of your calculator on each answer book used.

- 1. You buy one share of stock in company C for €10. Each day the price of C either increases by €1 with probability p or decreases by €1 with probability 1-p. These changes from day to day are statistically independent. You decide to sell your share if it gains €2 (i.e. reaches a price of €12).
  - (i) What is the probability that you will sell your share exactly 4 days after you buy it? [5 marks]
  - (ii) What is the probability that you sell your share at least 4 days after you buy it? [5 marks]

Suppose now that the daily change in the price of stocks in company C is observed to be related to the change in price of stocks in company D. Namely, the probability that stock in C increases by €1 is equal to 0.2 when the price of stock in company D increases that day, and is equal to 0.1 otherwise.

(iii) State the definition of conditional probability.

[5 marks]

(iv) Describe how marginalisation can be used to calculate the probability of an event E based on knowledge of the conditional probabilities  $P(E|F_1)$ ,  $P(E|F_2)$  and  $P(E|F_3)$  plus the probabilities  $P(F_1)$ ,  $P(F_2)$  and  $P(F_3)$  when events  $F_1$ ,  $F_2$ ,  $F_3$  are mutually exclusive and  $F_1$  U  $F_2$  U  $F_3$  equals the sample space.

[5 marks]

(v) Suppose that the probability that stock in company D increases on a given day is 0.5. Calculate the probability that stock in company C increases that day. [5 marks]

- (i) Prob sell in 4 days = Prob of one increase and decrease followed by two increases = 2p³(1-p). Note that if had four increases then would have sold on day 2 (when had increased by €2).
- (ii) Prob sell in 4 or more days = 1-(Prob sell in day 1 + Prob sell in day 2+Prob sell in day 3). Prob sell in day 1=0 (since need increase of €2 at least). Prob sell in day 2 = p². Prob sell in day 3 = 0 (if have 3 wins then sell on day 2, if a loss and 2 wins then don't sell). So answer is 1-p²
- (iii) For random events E and F,  $P(E|F) = P(E \cap F)/P(F)$
- (iv)  $P(E) = P(E|F_1)P(F_1) + P(E|F_2)P(F_2) + P(E|F_3)P(F_3)$
- (v) P(C increases) = P(C increases|D increases)P(D increases) + P(C increases|D doesn't increase)P(D doesn't increase) = 0.2 x 0.5 + 0.1x(1-0.5)

- 2. Suppose you play a game where four 6-sided fair dice are rolled. Let X be equal to the <u>minimum</u> of the four values rolled (it is ok if more than one dice has the minimal value). It costs €2 to play the game and you win €X.
  - (i) Calculate  $P(X \ge k)$  as a function of k=1,2,...,6.

[5 marks]

- (ii) Assuming you know P(X≥k) for k=1,2,...,6, show how to calculate the PMF of X. [5 marks]
- (iii) State the definition of the expected value.

[5 marks]

(iv) Calculate E[X].

[5 marks]

(v) If you play the game many times do you expect to make a profit (win more than you pay to play the game)? Explain your reasoning. What is the amount cost to play that would make you break even (i.e. have an expected profit of zero)? [5 marks]

- (i)  $P(X \ge 1) = 1$  since dice must come up one or higher.  $P(X \ge 2) = (5/6)^4$  since probability of one dice rolling 2 or higher is 5/6, and since dice rolls are independent the probability that all four dice at 2 or greater is  $(5/6)^4$ . By similar reasoning  $P(X \ge 3) = (4/6)^4$ ,  $P(X \ge 4) = (3/6)^4$ ,  $P(X \ge 5) = (2/6)^4$ ,  $P(X \ge 6) = (1/6)^4$ .
- (ii)  $P(X=6)=P(X\ge6)=(1/6)^4$ .  $P(X=5)=P(X\ge5)-P(X\ge6)$ ,  $P(X=4)=P(X\ge4)-P(X\ge5)$ , etc.
- (iii) For RV X taking values  $x_1, x_2, ..., x_n$  then  $E[X] = x_1 P(X = x_1) + x_2 P(X = x_2) + ... + x_n P(X = x_n)$
- (iv) E[X] = 1.P(X=1)+2P(X=2)+...+6P(X=6) using values from (ii) above.
- (v) The expected profit is E[X]-2. If we play the game N times, for N sufficiently large then our profit is N(E[X]-2) with high probability. To break even the cost to play would be E[X].

- 3. A survey is carried out by selecting n people from the population and asking each person to answer either "yes" or "no" to a question. Let random variable  $Y_i$  take value 1 when the i'th respondent answers "yes" and 0 otherwise. The random variables  $Y_i$  i=1,2,..,n are independent and identically distributed with  $E[Y_i] = \mu$ .
  - (i) Let random variable  $Z = \sum_{i=1}^{n} Y_i$ . Write an expression for E[Z] in terms of E[Y<sub>i</sub>]. Explain your answer. Hint: use the linearity of the expected value. [5 marks]
  - (ii) Using the definition of expectation prove that E[Z/n]=E[Z]/n for n>0. [5 marks]
  - (iii) Using Chebyshev's inequality explain the weak law of large numbers and the behaviour of  $|Z/n \mu|$  as n becomes large. Recall that for random variable X Chebyshev's inequality is:  $P(|X \mu| \ge k) \le E[(X \mu)^2]/k^2$  for any k and  $\mu$ . [5 marks]
  - (iv) Explain what a confidence interval is, using Z/n as an estimate of  $\mu$  as an example. [5 marks]
  - (v) Describe how to use bootstrapping to estimate a confidence interval for Z/n. [5 marks]

- (i)  $E[Z] = E[\sum_{i=1}^{n} Y_i] = \sum_{i=1}^{n} E[Y_i] = n E[Y_i]$  where we use linearity of the expectation to move the E[] inside the sum
- (ii)  $E[Z/n] = \sum_{i=1}^{n} i/n P(Z=i) = (\sum_{i=1}^{n} i P(Z=i))/n = E[Z]/n$
- (iii)  $E[Z/n]=E[Z]/n=E[Y_i]=\mu$ .  $Var(Z/n)=1/n^2Var(Z)=nVar(Y_i)/n^2$ . Plugging these values into Chebyshev's inequality,  $P(|Z/n-\mu|\geq k)\leq Var(Y_i)/nk^2$ . For any value of k>0, as n goes to infinity then  $Var(Y_i)/nk^2$  goes to zero.
- (iv) A confidence interval is an interval [a,b] within which a RV X lies with a specified probability e.g. with probability at least 0.95. This can be written  $P(a \le X \le b) \ge 0.95$ . In the case of Z/n as an estimate of  $\mu$  we might consider the interval  $P(\mu-\epsilon \le Z/n \le \mu+\epsilon)$ . From (iii) we know that  $P(\mu-\epsilon \le Z/n \le \mu+\epsilon)$  goes to 1 for any  $\epsilon > 0$  as n grows large.
- (v) From the observed data Yi, i=1,..,n, draw a sample of m points uniformly at random with replacement. Using this sample calculate estimate Z/n. Repeat to obtain a number of estimates. From the distribution of these estimates we can then estimate a confidence interval for Z/n.

4. Suppose we mark the answers of 200 students to each of 10 exam questions. Let S<sub>ij</sub> be an indicator variable which is 1 if student i answered question j correctly and -1 otherwise. You observe all of the answers for all students. Assume that

$$P(S_{ij}=y | a_i, d_j) = 1/(1+exp(-y(a_i-d_j)))$$

where  $a_i$  is a parameter that represents the students ability and  $d_j$  is a parameter which represents the questions difficulty.

- (i) Give an expression for the log-likelihood of this exam data (the data consisting of the answers by all 200 students). Hint: this is an example of a logistic regression model. [5 marks]
- (ii) Outline how gradient descent might be used to find the maximum likelihood estimates for the unknown parameters a<sub>i</sub> and d<sub>i</sub>. [5 marks]
- (iii) With reference to Bayes Rule explain what is meant by the likelihood, prior and posterior. [5 marks]
- (iv) Explain how the maximum a posteriori (MAP) estimate of a parameter differs from the maximum likelihood estimate. [5 marks]
- (v) How could you incorporate knowledge of the prior probability distribution of parameters a<sub>i</sub> into the above model to obtain a MAP estimate? [5 marks]

- (i) The likelihood of the observed mark data  $s_{ij}$ , i=1,...,200, j=1,...,10 is  $P\left(S_{ij}=s_{ij},i=1,...,200,j=1,...,10|a_i,d_j\ i=1,...,200,j=1,...,10\right)=\prod_{i=1}^{200}\prod_{j=1}^{10}P(S_{ij}=s_{ij}|\ a_i,d_j)$  . The log-likelihood is  $L=Log\prod_{i=1}^{200}\prod_{j=1}^{10}P(S_{ij}=s_{ij}|\ a_i,d_j)=\sum_{i=1}^{200}\sum_{j=1}^{10}\log P(S_{ij}=s_{ij}|\ a_i,d_j)=-\sum_{i=1}^{200}\sum_{j=1}^{10}\log (1+\exp{(-s_{ij}(a_i-d_j))})$
- (ii) The ML estimates select the parameters  $a_i$  and  $d_j$  to maximise the likelihood L. Starting from an intial estimate, these values can be found by iteratively updating the estimates such that L decreases after each update until the decrease in L becomes small enough. We can find updates that decrease L by local search or by taking a step in the direction of the derivatives of L wrt  $a_i$  and  $d_i$ .
- (iii) For random events E and F Bayes Rule is P(E|F)=P(F|E)P(E)/P(F). P(F|E) is the likelihood, P(E) the prior and P(E|F) the posterior.
- (iv) In a MAP estimate the parameter values are selected to maximise the posterior probability P(parameters | data) rather than the likelihood P(data | parameters).
- (v) By Bayes, the posterior is proportional to  $P(S_{ij}=s_{ij},i=1,...200,j=1,...,10|a_i,d_j,s_{ij},i=1,...200,j=1,...,10)P(a_i,i=1,...200)$ . The MAP estimate of the a<sub>i</sub>, i=1,...,200 maximises this.