## Quiz 1: CS 215

Name: \_\_\_\_\_\_Roll Number: \_\_\_\_\_

Attempt all five questions, each carrying 10 points. Clearly mark out rough work. Useful Information

- 1. Binomial theorem:  $(x+y)^n = \sum_{k=0}^n C(n,k) x^k y^{n-k}$
- 2. The empirical mean of n independent and identically distributed random variables is approximately Gaussian distributed. The approximation accuracy is better when n is larger.
- 3. Defining  $\Phi(x) = \int_{-\infty}^{x} \frac{e^{-x^2/2}}{\sqrt{2\pi}} dx$ , we have the following table:

| n   | $\Phi(n) - \Phi(-n)$ |
|-----|----------------------|
| 1   | 68.2%                |
| 2   | 95.4%                |
| 2.6 | 99%                  |
| 2.8 | 99.49%               |
| 3   | 99.73%               |

- 4. For a non-negative random variable X, we have  $P(X \ge a) \le E(X)/a$  where a > 0.
- 5. For a random variable X with mean  $\mu$  and variance  $\sigma^2$ , we have  $P(|X \mu| \ge k\sigma) \le \frac{1}{k^2}$ .
- 6. Integration by parts:  $\int u dv = uv \int v du$ .
- 7. Gaussian pdf:  $f(x) = \frac{1}{\sigma\sqrt{2\pi}}e^{-(x-\mu)^2/(2\sigma^2)}$
- 8. Poisson pmf:  $P(X = i) = \frac{e^{-\lambda}\lambda^i}{i!}$

Additional space

1. An exponential random variable X has a pdf which is given as  $f_X(x) = \lambda e^{-\lambda x}$  where  $x \in [0, \infty)$  and  $\lambda > 0$ . Derive the pmf of floor(X) and ceil(X). Recall that ceil(X) is the smallest integer greater than or equal to X and floor(X) is the largest integer less than or equal to X. [5+5=10 points]

Solution:  $P(\text{floor}(X) = n) = P(n \le X < n+1) = F_X(n+1) - F_X(n) = (1 - e^{-\lambda(n+1)}) - (1 - e^{-\lambda n}) = F_X(n+1) - F_X(n) = (1 - e^{-\lambda(n+1)}) - (1 - e^{-\lambda n}) = (1 - e^{-\lambda(n+1)}) - (1 - e^{-\lambda n}) = (1 - e^{-\lambda(n+1)}) - (1 - e^{-\lambda(n+1)}) - (1 - e^{-\lambda(n+1)}) = (1 - e^{-\lambda(n+1)}) - (1 - e^{-\lambda(n+1)}) - (1 - e^{-\lambda(n+1)}) = (1 - e^{-\lambda(n+1)}) - (1 - e^{-\lambda(n+1)}) = (1 - e^{-\lambda(n+1)}) - (1 - e^{-\lambda(n+1)}) = (1 - e^{-\lambda(n+1)}) - (1 - e^{-\lambda(n+1)}) - (1 - e^{-\lambda(n+1)}) = (1 - e^{-\lambda(n+1)}) - (1$  $e^{-\lambda n}(1-e^{-\lambda}).$ 

$$P(\text{ceil}(X) = n) = P(n-1 \le X < n) = F_X(n) - F_X(n-1) = (1 - e^{-\lambda(n)}) - (1 - e^{-\lambda(n-1)}) = e^{-\lambda(n-1)}(1 - e^{-\lambda}).$$

2. Verify whether true or false with justification (no credit without it): (a) The minimum of n iid Bernoulli random variables is also a Bernoulli random variable. (If your answer is in the affirmative, what is the parameter of the Bernoulli random variable?). (b) The minimum of n iid geometric random variables is also a geometric random variable. (If your answer is in the affirmative, what is the parameter of the geometric random variable?). Recall that the pmf of a geometric random variable has the form  $P(X=i)=(1-p)^{i-1}p$ . [5+5=10 points]

Solution: Part a:

Let  $Y = min\{X_i\}_{i=1}^n$  where  $X_i \sim Bernoulli(p)$ . Then  $P(Y=1) = \prod_{i=1}^n P(X_i=1) = p^n$ . Also  $P(Y=0) = \prod_{i=1}^n P(X_i=1) = p^n$ .  $1 - P(Y = 1) = 1 - p^n$ . So Y is a Bernoulli random variable with parameter  $p^n$ .

Part b:

Let  $Y = min\{X_i\}_{i=1}^n$  where  $X_i \sim Geometric(p)$ . Then  $P(Y \geq y) = \prod_{i=1}^n P(X_i \geq y) = ((1-p)^y)^n = (1-p)^{ny}$ . Hence  $P(Y < y) = 1 - [(1-p)^{ny}]$ . This is the CDF of a geometric random variable with the parameter  $1-(1-p)^n$ .

3. If  $X \sim \text{Poisson}(\lambda)$ ,  $Y \sim \text{Uniform}(-a, a)$  where a > 0 and Z = X + Y, derive an expression for  $E[(Z - E(Z))^3]$ . Assume X and Y are independent. [10 points]

**Solution:** We have  $E(Z) = E(X) + E(Y) = \lambda$ . We also have  $E(X^2) = \lambda^2 + \lambda$ .

Now LHS =  $E[(X + Y - \lambda)^3] = E[(X + Y)^3 - \lambda^3 + 3(X + Y)\lambda^2 - 3\lambda(X + Y)^2]$ . Now  $E[(X + Y)^3] = E[X^3 + Y^3 + 3X^2Y + 3XY^2]$ . Now  $E[Y] = E[Y^3] = 0$ ,  $E[Y^2] = a^2/3$ , so we have  $E[(X+Y)^3] = E[X^3 + 3XY^2] = E[X^3] + 3\lambda a^2/3 = E[X^3] + a^2\lambda.$ 

We have  $E[X^3] = \sum_{k=0}^{\infty} k^3 \frac{e^{-\lambda} \lambda^k}{k!} = \sum_{k=1}^{\infty} k^2 \frac{e^{-\lambda} \lambda^k}{(k-1)!}$  which upon replacing k by l+1 further yields  $\sum_{l=0}^{\infty} (l+1)^{l+1} (l+1)^{l+1}$ 

$$1)^{2} \frac{e^{-\lambda} \lambda^{l+1}}{l!} = \lambda E((X+1)^{2}) = \lambda (E[X^{2} + 2X + 1]) = \lambda^{3} + 3\lambda^{2} + \lambda.$$

Hence  $E[(X+Y)^3] = \lambda^3 + 3\lambda^2 + \lambda + \lambda a^2$ .

Also  $E[3(X+Y)\lambda^2 - 3\lambda(X+Y)^2] = 3\lambda^3 + 3\lambda^2 E[Y] - 3\lambda E[X^2 + Y^2 + 2XY] = 3\lambda^3 - 3\lambda(\lambda^2 + \lambda + a^2/3 + 0) =$ 

Combining all these results together, we get  $E[(X + Y - \lambda)^3] = \lambda^3 + 3\lambda^2 + \lambda + \lambda a^2 - \lambda^3 - 3\lambda^2 - \lambda a^2 = \lambda$ .

4. Let  $X_1, X_2, ..., X_n$  be independent random variables with the PDF  $f_X(x; a, b) = 1/(b-a)$  if  $a \le x \le b$ and 0 otherwise. Here a < b. Derive the maximum likelihood estimate for a for the special case when b = a + 1. What would be the maximum likelihood estimate of b if the PDF for each of the random variables  $X_1, X_2, ..., X_n$  was  $f_X(x; 0, b) = 1/b$  if  $0 \le x < b$  and 0 otherwise? Notice the strict inequality in the second sub-question. [7+3=10 points]

**Solution:** The value of the likelihood function given all the samples is 1 if  $\forall i, a \leq x_i \leq a+1$ . Hence  $a \leq x_{min}$ and  $a+1 \ge x_{max}$ , i.e.  $a \ge x_{max} - 1$ . As the likelihood is constant in the domain  $\forall i, a \le x_i \le a+1$ , we know that any value lying inside the interval  $[x_{max} - 1, x_{min}]$  is a maximum likelihood estimate for a. This is an example where the MLE is not unique!

For the second part, we have the constraint that for a non-zero PDF, we must have  $x \ge 0$  and x < b (strict inequality). In such a case, we cannot have  $b = x_{max}$  unlike the earlier example. Hence we would want to consider an MLE of  $x_{max} + \delta$  for an infinitesimally small  $\delta$ . However for any  $\delta$  you choose, one can reduce its value (i.e. the value of  $\delta$ ) further and increase the likelihood from  $1/(x_{max}+\delta)^n$  to  $1/(x_{max}+\delta')^n$  where  $\delta' < \delta$ . This is a case where the MLE of b is therefore not defined!

5. A storage device contains the monthly expenses of a group  $\mathcal{G}$  of n individuals in a country. A computer program has read through these records, and has computed and stored in memory the value S = $\sum_{i=1}^{n} \sum_{j=1}^{n} (x_i - x_j)^2$  where  $x_i$  is the monthly expenses of the i<sup>th</sup> individual. Some analysis you wish to perform requires the sample standard deviation of the monthly expenses of the individuals in  $\mathcal{G}$ . However,

the storage device is incredibly slow and you do not have the option of reading any of the data again. How will you compute the standard deviation given S and n? Derive all required formulae if necessary. [10 points] **Solution:** The value of S is actually proportional to the sample variance  $\sigma^2 = \sum_i (x_i - m)^2/(n-1)$ . To see this, consider that  $S = \sum_{i=1}^n \sum_{j=1}^n (x_i - x_j)^2 = \sum_{i,j} (x_i - m + m - x_j)^2 = \sum_{i,j} (x_i - m)^2 + (x_j - m)^2 + 2(x_i - m)(x_j - m) = n \sum_i (x_i - m)^2 + n \sum_j (x_j - m)^2 + 0$ . Here m is the sample mean. Note that the cross-term in the summation above is 0 because  $\sum_i (x_i - m) = 0$  by definition of m. Hence we have  $S = 2n \sum_i (x_i - m)^2 = 2n(n-1)\sigma^2$ . Hence  $\sigma = \sqrt{S/(2n(n-1))}$ .