

R computing for Business Data Analytics

Homework 3 (Due date: November 14, 2014)

Please e-mail your homework (.pdf) and the associated R code (.R) to hchuang.om@gmail.com.

The email title must be **R_HW3_GroupName**. NO late homework will be accepted.

Q1. (20%) The probability density $P(X=k)$ of a random variable $X \sim \text{Poisson-Tweedie family } (\alpha, \beta, \gamma)$ can be calculated through a *double recursion algorithm*.

The first recursion exists in p_k :

$$p_0 = \begin{cases} e^{\beta[(1-\gamma)^\alpha - 1]/\alpha}, & \alpha \neq 0 \\ (1-\gamma)^\beta, & \alpha = 0 \end{cases}, \quad p_1 = \beta\gamma * p_0,$$
$$p_{k+1} = \frac{1}{k+1} \left(\beta\gamma * p_k + \sum_{j=1}^k j * r_{k+1-j} p_j \right), \quad k=1, 2, \dots$$

The second recursion exists in r_j :

$$r_1 = (1-\alpha)\gamma, \quad r_{j+1} = \left(\frac{j-1+\alpha}{j+1} \right) \gamma * r_j, \quad j=1, 2, \dots$$

where $\alpha \in (-\infty, 1]$, $\beta \in (0, +\infty]$, $\gamma \in [0, 1]$

Write a function *PTF* that has four inputs (k, a, b, g) (a for α , b for β , and g for γ) and returns p_k .

Use the function to calculate *PTF*(9, -3, 2, 0.5) (The answer should be close to 0.04235).

Q2. (20%) MLE and Simulation

(a) For the Bernoulli distribution $P(X = x_i | p) = p^{x_i} (1-p)^{1-x_i}$, derive \hat{p}_{MLE} .

(Hint: Solve p for $\ell'(p) = 0$ where $\ell(p) = \log(\prod_{i=1}^n p^{x_i} (1-p)^{1-x_i})$)

(b) Continuing (a), given $p=0.5$, simulate $n=50$, $n=5000$, $n=500,000$ Bernoulli random numbers.

For each simulated sample of size n , calculate \hat{p}_{MLE} from the sample and compare \hat{p}_{MLE} to the TRUE $p=0.5$, what have you observed?

(c) For the exponential distribution $f(x_i) = \lambda e^{-\lambda x_i}$, derive $\hat{\lambda}_{MLE}$ and prove the Markov Property:

$$P(X > s+t | X > s) = P(X > t) \quad (\text{Hint: } F(x) = 1 - e^{-\lambda x})$$

(d) Continuing (c), given $\lambda=0.5$, simulate $n=50$, $n=5000$, $n=500,000$ exponential random numbers.

For each simulated sample of size n , calculate $\hat{\lambda}_{MLE}$ from the sample and compare $\hat{\lambda}_{MLE}$ to the TRUE $\lambda=0.5$, what have you observed?

Q3. (20%) Binomial and Poisson Distributions

- (a) The table below records the historical number of car accidents/week in a district. Create a vector called *car.accident* that stores 109 zeros, 65 ones, 22 twos, 3 threes, and 1 four.
- (b) Apply the *fitdistr()* function in *R* (load the *MASS* library first) to fit the *car.accident* data with the Poisson distribution. What is the value of estimated λ ? What is the log-likelihood?
- (c) Given the estimated λ , use *R* to do the computation and finish the 3rd column of table below. Are the predicted frequencies close to the actual frequency?
- (d) Given the estimated λ , finish the 4th column of table below using *R*.

Car Accidents	Frequency	Poisson($\lambda=???$)	Binomial($n=200, p=\lambda/200$)
0	109	$200 * P(X=0 \lambda)=$	$200 * P(X=0 n, p)=$
1	65	$200 * P(X=1 \lambda)=$	$200 * P(X=1 n, p)=$
2	22	$200 * P(X=2 \lambda)=$	$200 * P(X=2 n, p)=$
3	3	$200 * P(X=3 \lambda)=$	$200 * P(X=3 n, p)=$
4	1	$200 * P(X=4 \lambda)=$	$200 * P(X=4 n, p)=$
>4	0	$200 * P(X>4 \lambda)=$	$200 * P(X>4 n, p)=$

Q4. (20%) Binomial, Poisson, and Normal Distributions

- (a) Finish the second, third, and fourth column of the table below using *R*.
- (b) Based on the probability densities you calculate, generate a plot in which the x-axis is 0:8 and the y-axis lies between [0, 0.25]. The plot should have **three lines with different width and colors**. (red thin line for Binomial, green thick line for Poisson, blue thicker line for Normal).

Defectives (x)	Binomial($n=20, p=0.2$) $P(X=x)$	Poisson($\lambda=4$) $P(X=x)$	Normal ($\mu = 4, \sigma^2 = 4$) $P(x-0.5 < X < x+0.5)$
0			
1			
2			
3			
4			
5			
6			
7			
8			

Q5. (20%) Random Numbers and Monte-Carlo Simulation

(a) Implement the algorithm in the bottom of page 10 in lecture 6. Write a function *runi.congru* that has five arguments (N, A, B, m, seed) (N is the number of random variates to be simulated). After that, generate five uniform random numbers using A=1217, B=0, m=32767, and seed=1. Save the five numbers as a vector u in R and show me the five numbers.

(b) Implement the *inverse transformation* method in the bottom of page 11 in lecture 6. Write a function *rbinom.invtran* that has four arguments (N, n, p, uni) and returns N binomial random numbers. Set n=3, p=0.5, and use the vector u in part (a) as inputs to uni to generate five binomial random variates. Show me the simulated numbers too.

(c) Now, use the function *runi.congru* in (a) to simulate another 50 numbers by setting seed=2 (A=1217, B=0, m=32767 still) and store the 50 numbers in a vector U.

(d) For the zero-truncated Poisson distribution

$$P(X = x | \lambda, x > 0) = \frac{\lambda^x e^{-\lambda}}{x!(1 - e^{-\lambda})}$$

Following the logic of inverse transformation, write a function *rzt pois.invtran* that has three arguments (N, lambda, uni) (lambda for λ) and returns N zero-truncated Poisson random numbers. Set lambda=4 and use the vector U (part (c)) as inputs to uni to generate 50 non-zero Poisson random variates. Show me the simulated numbers too.

(e) Assume the simulated numbers $(x_1, x_2, \dots, x_{50})$ are *counts of arrived customers per group* for a restaurant. Let's say all $x_i \leq 2$ will be assigned to two-people tables, all $2 < x_i \leq 4$ will be assigned to four-people tables, $4 < x_i \leq 6$ will be assigned to six-people tables, and finally, all $6 < x_i$ will be put into private rooms.

Now, based on the simulated 50 groups $(x_1, x_2, \dots, x_{50})$, how many groups of customers will be seated in two-people tables, four-people tables, six-people tables, and private rooms respectively?