

Probability Model of Mobile Phone Tariff Selection Based on Poisson Regression Model

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Abstract:

With the deepening of the informatization degree of social life, mobile phones have become an indispensable tool in People's Daily life, study and work. In order to guide people to make more scientific choices, some reasonable models are needed as a reference. The choice of mobile phone tariff is a decision-making problem that people often encounter. For companies, how to set mobile phone rates and which mobile phone rates are more likely to be selected by whom are the decisions they need to make. Determine these problems more effectively in order to research, this article through early to collect data, the Poisson regression model was established, using the largest eigenvalue method and maximum likelihood estimation method to determine the parameters in the model, and finally determine the established model, on the basis of this model can calculate the probability of choosing a mobile phone charges.

Keywords: Poisson regression model; Probability; Maximum likelihood method; Mobile phone charges estimation

I. INTRODUCTION

Suppose Y_1, Y_2, \dots, Y_m are independent of each other. $Y_k \sim P(\lambda_k)$. That is

$Var(Y_k) = E(Y_k) = \lambda_k$. Suppose

$$P(Y_k = y_k) = \frac{e^{-\lambda_k} \lambda_k^{y_k}}{\Gamma(1 + y_k)} \quad (1)$$

$\lambda_k = \exp(x_k^T \beta)$, X_1, X_2, \dots, X_m are independent of each other. $X_k \sim N(\mu_k, \Sigma_k)$ [1-6], $\mu_k, \Sigma_k > 0$ are unknown. x_k is the value of X_k . $x_k = (x_{k1}, x_{k2}, \dots, x_{kp})^T$ is the $p \times 1$ dimensional column vector. $\beta = (\beta_1, \beta_2, \dots, \beta_p)^T$ is a vector of $p \times 1$ dimensional parameters.

Poisson regression model[7-16] is widely used in real life. For example, there are n types of policies, and each type of policy has p rate factors.

II. POISSON REGRESSION MODEL

2.1. Model Establishment

The overall value of (X, Y) is shown in Table 1, where Y represents the sold quantity of various mobile phone charges, and X represents the monthly function fee (yuan), domestic data traffic (G), and call duration (min) of the tariff in sale:

TABLE I. Types of mobile phone charges and sales quantity

Y	7	9	7	11	9	9	4	8	4	8
X	$\begin{pmatrix} 58 \\ 0.15 \\ 150 \end{pmatrix}$	$\begin{pmatrix} 88 \\ 0.3 \\ 350 \end{pmatrix}$	$\begin{pmatrix} 128 \\ 0.6 \\ 650 \end{pmatrix}$	$\begin{pmatrix} 158 \\ 0.6 \\ 900 \end{pmatrix}$	$\begin{pmatrix} 188 \\ 0.6 \\ 1200 \end{pmatrix}$	$\begin{pmatrix} 58 \\ 0.5 \\ 50 \end{pmatrix}$	$\begin{pmatrix} 88 \\ 0.7 \\ 200 \end{pmatrix}$	$\begin{pmatrix} 128 \\ 1 \\ 420 \end{pmatrix}$	$\begin{pmatrix} 158 \\ 2 \\ 510 \end{pmatrix}$	$\begin{pmatrix} 188 \\ 2.5 \\ 600 \end{pmatrix}$

Poisson regression model is constructed as follows:

Suppose Y_1, Y_2, \dots, Y_{10} are independent of each other. $Y_k \sim P(\lambda_k)$. That is $Var(Y_k) = E(Y_k) = \lambda_k$. Suppose

$$P(Y_k = y_k) = \frac{e^{-\lambda_k} \lambda_k^{y_k}}{\Gamma(1 + y_k)} \quad (k = 1, 2, \dots, 10) \tag{2}$$

$\lambda_k = \exp(x_k^T \beta)$, X_1, X_2, \dots, X_{10} are independent of each other. $X_k \sim N(\mu_k, \Sigma_k)$, $\mu_k, \Sigma_k > 0$ are unknown. x_k is the value of X_k . $x_k = (x_{k1}, x_{k2}, x_{k3})^T$ is the eigenvariable column vector. $\beta = (\beta_1, \beta_2, \beta_3)^T$ is a vector of 3×1 dimensional parameters.

2.2. Estimation of parameter unit vector

There are 100 samples $(X^{(i)}, Y^{(i)})$ ($i = 1, 2, \dots, 100$) in the model population (X, Y) . The distribution of sample quantity is shown in Table 2.

TABLE II. Types of mobile phone charges and the frequency

<i>Y</i>	7	9	7	11	9	9	4	8	4	8
<i>X</i>	$\begin{pmatrix} 58 \\ 0.15 \\ 150 \end{pmatrix}$	$\begin{pmatrix} 88 \\ 0.3 \\ 350 \end{pmatrix}$	$\begin{pmatrix} 128 \\ 0.6 \\ 650 \end{pmatrix}$	$\begin{pmatrix} 158 \\ 0.6 \\ 900 \end{pmatrix}$	$\begin{pmatrix} 188 \\ 0.6 \\ 1200 \end{pmatrix}$	$\begin{pmatrix} 58 \\ 0.5 \\ 50 \end{pmatrix}$	$\begin{pmatrix} 88 \\ 0.7 \\ 200 \end{pmatrix}$	$\begin{pmatrix} 128 \\ 1 \\ 420 \end{pmatrix}$	$\begin{pmatrix} 158 \\ 2 \\ 510 \end{pmatrix}$	$\begin{pmatrix} 188 \\ 2.5 \\ 600 \end{pmatrix}$
frequency	6	12	16	10	7	5	10	11	16	7

Let $\hat{\mu}_l = X^{(l)} - \bar{X}$, the values of $\hat{\mu}_l$ are shown in Table 3.

TABLE III. The values of $\hat{\mu}_l$

<i>l</i>	1	2	3	4	5
$\hat{\mu}_l$	$\begin{pmatrix} -69.7 \\ -0.725 \\ -371.3 \end{pmatrix}$	$\begin{pmatrix} -39.7 \\ -0.575 \\ -171.3 \end{pmatrix}$	$\begin{pmatrix} 0.3 \\ -0.275 \\ 128.7 \end{pmatrix}$	$\begin{pmatrix} 30.3 \\ -0.275 \\ 378.7 \end{pmatrix}$	$\begin{pmatrix} 60.3 \\ -0.275 \\ 678.7 \end{pmatrix}$
<i>l</i>	6	7	8	9	10
$\hat{\mu}_l$	$\begin{pmatrix} -69.7 \\ -0.375 \\ -471.3 \end{pmatrix}$	$\begin{pmatrix} -39.7 \\ 0.125 \\ -321.3 \end{pmatrix}$	$\begin{pmatrix} 0.3 \\ 1.125 \\ -101.3 \end{pmatrix}$	$\begin{pmatrix} 30.3 \\ -0.175 \\ -11.3 \end{pmatrix}$	$\begin{pmatrix} 60.3 \\ 1.625 \\ 78.7 \end{pmatrix}$

$$\hat{V}_1 = \sum_{l=1}^{10} \frac{N_l}{100} \hat{\mu}_l \hat{\mu}_l^T = \begin{pmatrix} 1628.91 & 10.6225 & 9579.39 \\ 10.6225 & 0.433725 & 0.3825 \\ 9579.39 & 0.3825 & 84041.31 \end{pmatrix} \quad (3)$$

$$\hat{\Sigma} = \frac{1}{10} \sum_{l=1}^{10} \hat{\mu}_l \hat{\mu}_l^T = \begin{pmatrix} 2197.69 & 16.256 & 13509.71 \\ 16.256 & 0.517625 & 19.399 \\ 13509.71 & 19.399 & 112975.89 \end{pmatrix} \quad (4)$$

$$\hat{V} = \hat{\Sigma}^{-1} \hat{V}_1 \hat{\Sigma}^{-1} = \begin{pmatrix} 0.0067 & -0.1938 & -0.0008 \\ -0.1938 & 7.1993 & 0.0218 \\ -0.0008 & 0.0218 & 0.0001 \end{pmatrix}, \quad (5)$$

The maximum characteristic root of \hat{V} is $\lambda_1=7.2046$, and its corresponding feature vector is

$$\hat{\eta} = \begin{pmatrix} -0.0269 \\ 0.9996 \\ 0.0030 \end{pmatrix}.$$

Estimation of unit vector $\beta' = \frac{\beta}{\|\beta\|}$ of parameter β is $\hat{\beta}' = \hat{\eta} = \begin{pmatrix} -0.0269 \\ 0.9996 \\ 0.0030 \end{pmatrix}$.

2.3. Estimation of parameter vector modulus

The maximum likelihood estimation method is used to estimate the modulus α of parameters β , and the Poisson regression model constructed in this paper can be rewritten as:

$$P(Y_k = y_k | X = x) = \frac{\exp[y_k x_k^T \beta' \alpha - \exp(x_k^T \beta' \alpha)]}{\Gamma(1 + y_k)} \quad (6)$$

Replacing the estimator $\hat{\beta}'$ of the parameter unit vector β' in the above equation, Then we get:

$$P(Y_k = y_k | X = x) = \frac{\exp[y_k x_k^T \hat{\beta}' \alpha - \exp(x_k^T \hat{\beta}' \alpha)]}{\Gamma(1 + y_k)} \quad (7)$$

Starting from the above formula, the joint density can be obtained as:

$$\begin{aligned} f(x, y) &= f_{Y|X}(y) f_X(x) \\ &= \frac{\exp[y_k x_k^T \hat{\beta}' \alpha - \exp(x_k^T \hat{\beta}' \alpha)]}{\Gamma(1 + y_k)} \\ &\prod_{j=1}^{100} \frac{1}{(\sqrt{2\pi})^3} |\Sigma_j|^{-\frac{1}{2}} \exp\left[-\frac{1}{2}(x_j - \mu_j)^T \Sigma_j^{-1} (x_j - \mu_j)\right] \end{aligned} \quad (8)$$

Omitting the irrelevant part, and getting the likelihood function:

$$L(\alpha; y_1, \dots, y_{100}) = \prod_{k=1}^{100} \frac{\exp\left[y_k x_k^T \hat{\beta}' \alpha - \exp(x_k^T \hat{\beta}' \alpha)\right]}{\Gamma(1 + y_k)} \quad (9)$$

The logarithmic likelihood function is

$$\begin{aligned} l(\alpha; y_1, \dots, y_{100}) &= \sum_{k=1}^{100} \ln \frac{\exp\left[y_k x_k^T \hat{\beta}' \alpha - \exp(x_k^T \hat{\beta}' \alpha)\right]}{\Gamma(1 + y_k)} \\ &= \sum_{k=1}^{100} \left[y_k x_k^T \hat{\beta}' \alpha - \exp(x_k^T \hat{\beta}' \alpha) \right] - \sum_{k=1}^{100} \ln [\Gamma(1 + y_k)] \end{aligned} \quad (10)$$

The likelihood equation is

$$\sum_{k=1}^{100} \left[y_k x_k^T \hat{\beta}' - x_k^T \hat{\beta}' \cdot \exp(x_k^T \hat{\beta}' \alpha) \right] = 0 \quad (11)$$

By calculating the data in the paper, the results are shown in Table 4

TABLE IV. Product

Product	x_1^T	x_2^T	x_3^T	x_4^T	x_5^T
$\hat{\beta}'$	-0.9570	-1.0090	-0.8773	-0.9278	-0.8269
Product	x_6^T	x_7^T	x_8^T	x_9^T	x_{10}^T
$\hat{\beta}'$	-0.9100	-0.7635	-0.1743	-2.0089	-0.7445
Product	$6x_1^T$	$12x_2^T$	$16x_3^T$	$10x_4^T$	$7x_5^T$
$\hat{\beta}'$	-5.7423	-12.1078	-14.0373	-9.2780	-5.7880
Product	$5x_6^T$	$10x_7^T$	$11x_8^T$	$16x_9^T$	$7x_{10}^T$
$\hat{\beta}'$	-4.5500	-7.6348	-1.9178	-32.1417	-5.2116

Accordingly, the likelihood equation can be written as

$$\begin{aligned}
 &7 \times (-5.7423) + 9 \times (-12.1078) + 7 \times (-14.0373) + 11 \times (-9.2780) \\
 &+ 9 \times (-5.7880) + 9 \times (-4.5500) + 4 \times (-7.6348) + 8 \times (-1.9178) \\
 &+ 4 \times (-32.1417) + 8 \times (-5.2116) + 100 \times 0.9570 \times \exp(-0.9570\alpha) \\
 &+ 100 \times 1.0090 \times \exp(-1.0090\alpha) + 100 \times 0.8773 \times \exp(-0.8773\alpha) \\
 &+ 100 \times 0.9278 \times \exp(-0.9278\alpha) + 100 \times 0.8269 \times \exp(-0.8269\alpha) \\
 &+ 100 \times 0.9100 \times \exp(-0.9100\alpha) + 100 \times 0.7635 \times \exp(-0.7635\alpha) \\
 &+ 100 \times 0.1743 \times \exp(-0.1743\alpha) + 100 \times 2.0089 \times \exp(-2.0089\alpha) \\
 &+ 100 \times 0.7445 \times \exp(-0.7445\alpha) = 0
 \end{aligned} \tag{12}$$

After calculation, the result is

$$\begin{aligned}
 &658.6686 - 95.70 \times \exp(-0.9570\alpha) \\
 &- 100.90 \times \exp(-1.0090\alpha) - 87.73 \times \exp(-0.8773\alpha) \\
 &- 92.78 \times \exp(-0.9278\alpha) - 82.69 \times \exp(-0.8269\alpha) \\
 &- 91 \times \exp(-0.9100\alpha) - 76.35 \times \exp(-0.7635\alpha) \\
 &- 17.43 \times \exp(-0.1743\alpha) - 200.89 \times \exp(-2.0089\alpha) \\
 &- 74.45 \times \exp(-0.7445\alpha) = 0
 \end{aligned} \tag{13}$$

Let $A = e^\alpha$, the above formula can be transformed into

$$\begin{aligned}
 &658.6686 - 95.70A^{-0.9570} - 100.90A^{-1.0090} - 87.73A^{-0.8773} \\
 &- 92.78A^{-0.9278} - 82.69A^{-0.8269} - 91A^{-0.9100} - 76.35A^{-0.7635} \\
 &- 17.43A^{-0.1743} - 200.89A^{-2.0089} - 74.45A^{-0.7445} = 0
 \end{aligned} \tag{14}$$

Multiply both ends of the equation by $A^{3.0089}$, Then get the equation

$$\begin{aligned}
 &658.6686 \times A^{3.0089} - 95.70A^{2.0519} - 100.90A^{2.0000} - 1.9999 \times 87.73 \\
 &- 92.78A^{2.0811} - 82.69A^{2.1820} - 91A^{2.0989} - 76.35 \\
 &- 17.43 \times A^{2.8346} - 74.45 \times A^{2.2644} - 200.89 \times A
 \end{aligned} \tag{15}$$

Divide both ends of the equation by 200.89, we get

$$\begin{aligned}
 &3.2788A^{3.0089} - 0.4764A^{2.0519} - 0.5023A^{2.0000} - 0.43 \\
 &- 0.4618A^{2.0811} - 0.4116A^{2.1820} - 0.4529A^{2.0989} - 0.37 \\
 &- 0.0868A^{2.8346} - 0.3706A^{2.2644} - A
 \end{aligned} \tag{16}$$

The iterative method is used to obtain $A = 1.3644$, that is $e^\alpha = 1.3644$, then $\hat{\alpha} = 0.3107$.

III. CONSISTENCY ESTIMATION OF PARAMETER VECTORS

According to Slutsky's theorem, the estimation of parameter vector β is consistent estimation:

$$\hat{\beta} = \hat{\alpha} \cdot \hat{\beta} = \begin{pmatrix} -0.0269 \\ 70.9996 \\ 0.0030 \end{pmatrix} = \begin{pmatrix} 0.008 \\ 0.31 \\ 0.0009 \end{pmatrix} \quad (17)$$

VI. CONCLUSION

Based on the analysis of the sales data of the tariff package launched by a mobile company, this paper chooses to use the Poisson model to calculate the probability that customers would choose a certain package. After building the model, the unit vector estimation of parameters is obtained by using the method of maximum eigenvalue and eigenvector, and the modulus estimation of parameters is obtained by using the maximum likelihood estimation method, so as to obtain the consistency estimation of parameters. Through this Poisson regression model, it is easy to determine the probability of each mobile phone tariff package. This model also has a very wide range of applications in all aspects of society.

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