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ORIGINAL PAPER

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Comparison of Statistical Models of Predict the Factors Affecting the Length of Stay (LOS) in the Intensive Care Unit (ICU) of a Teaching Hospital

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ABSTRACT

Introduction: Modeling can be a useful tool to find out how the distributions of hospital length of stay (LOS) and the factors affecting the length of stay. The present study aims to determine factors affecting the length of stay and selecting suitable statistical models. **Material and Method:** this is a cross – sectional study of 565 patients who were treated in the intensive care unit of Imam Khomeini hospital in Ahwaz. Preliminary data were collected retrospectively through the medical records of all patients admitted on intensive care units of Ahwaz Imam Khomeini Hospital in 2015. Statistical analysis and multivariate regression models were done using of SPSS 21 and STATA 7 software. **Results:** Average length of stay in ICU was 8.16±0.75 days. The Mean and Median age of patients were 58.61±20 and 61 respectively, The Mean LOS for females (16.44±9.37 days) was more than the men (11.5±5.35 days) ($p < 0.01$). The maximum and minimum lengths of stay belonged to patients with endocrine disorders (14.7±3.1 days) and patients with gastrointestinal disorders (5.53±1.1 days) respectively ($p < 0.01$). The goodness of fit for Gamma model showed that this model was more suitable and powerful than Log-normal model to predict the factors affecting the patient's length of stay in intensive care units of hospital. **Conclusion:** Gamma regression model was more robust to predict factors regarding the hospital length of stay. According to Gamma model the key factor in predicting the length of stay in ICU was the type of disease diagnosis. The result of statistical modeling can help managers and policy makers to estimate hospital resources and allocate them for different hospital services.

Keywords: length of stay, statistical modeling, ICU, teaching hospital.

1. INTRODUCTION

Intensive care unit (ICU) is one of the critical parts of a hospital which can reduce the rate of mortality and side effects of hospitalization (1, 2). An increase in the length of stay and the hospital costs are the main problems of admitted patients in intensive care units (3). The LOS is the key indicators that use to plan for hospital resources in making plans for patients (4-6). Clinical, demographic, and managerial factors are impact on the hospital length of stay in ICU (7-14). Since the optimum use of resources plays an important role in enhancing the efficiency of hospitals, reducing the length of stay can led to increase hospital resources efficiency and effectiveness (15-16). To manage hospital beds and patients LOS in ICU we need to model hospital data (17).

Modeling is one of the basic tools used in explaining the medical and health phenomena. These tools determine the distribution of specific variables and their relations with other variables by means of regression statistic methods. Modeling the length of stay is a valuable way to know about the status of distribution of LOS (18). The effecting factors in the length of stay cannot be predicted by using common statistic methods like linear regression method since the useful variable because the distribution of LOS data is normal. Hospital length of stay data have a right skewness (19-20). LOS is a numerical discrete variable; therefore, numerical data distribution should be used to analyze them (17). To model the numerical discrete variable should be used the model which has been a suitable prediction power (18). Statistical models with different distribution can predict the length of stay, as well as the factors influencing it (21). Identifying a robust

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model with a high sensitivity to predict the factors affecting the length of stay can help hospital managers to control prolonged hospital length of stay. The present study aims to determine factors affecting the length of stay and their suitable statistical models.

2. MATERIAL AND METHODS

This is a cross-sectional study of 565 patients treated in the intensive care unit of Imam Khomeini hospital in Ahwaz. Preliminary data were collected retrospectively from medical records of all patients admitted to the intensive care unit from 1 January 2015 to 31 December 2015. To confirm the content validity of data form, we used the expert’s views about questions and variables. In this study According to Kolmogorov-Smirnov test, distribution of the dependent variable data was abnormal and had a right skewness. In this study six percent of the patients an intensive care units died on the first day (LOS=0), so zero data have not been included in modeling. First, data were analyzed by univariate tests. Including Mann-Whitney, kruskal-Wallis, and Spearman coefficient. Then we selected the significant variables to model by using multi-variate regression analysis and stepwise modeling method. In this study significant level was determined 0.01. In this study Log-normal model was used base on formular: In $(y) = 0+1+$, in which X variable data are made from even distribution $(0, 1)$, $\sim N(0, \sigma^2)$, in which, $\sigma^2 = 0.5, 1.0, 1.5, 1=1$ was used, σ was estimated in this way:

$$E(y) = 1;$$

$$E(y|x) = \exp(0+1x+0.5^2)$$

Also Y variable skewness is as follows:

$$(\exp(\sigma^2)+2)(\exp(\sigma^2)-1)^{0.5}$$

And also some data are produced by log-normal distribution with variances of 0.5, 1, 1.5 and 2. Also, GAMMA model was used based on following formula: Variable data from standard GAMMA with density function

In which scale parameter is $b = \exp(\beta_0 + \beta_1x)$. and are chosen in such a way that result will be and they are produced by parameters of 5,0,1,2,4. We used SPSS 21 and STATA 7 to analyze the data modeling.

3. RESULTS

Table 1 shows the demographic and clinical characteristics which influence the patients’ length of stay in intensive care unit. Average length of stay in ICU was 8.16 ± 0.75 days. The Mean and Median age of patients were 58.61 ± 20 and 61 respectively, The Mean LOS for females (16.44 ± 9.37 days) was more than the men (11.5 ± 5.35 days) ($p < 0.01$). The maximum and minimum lengths of stay belonged to patients with en-

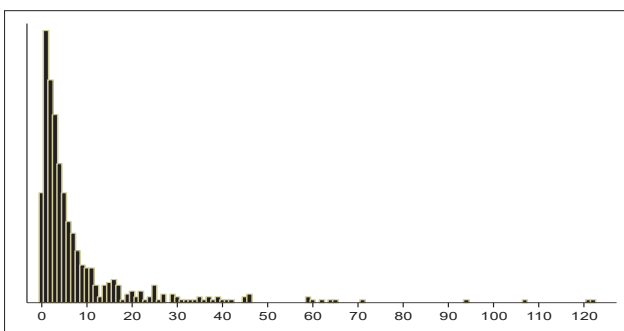


Diagram 1. Data distribution of LOS

P.value	Mean±SD	variables	
0.01>			sex
	5.35±11.5	339 (60%)	male
	9.37±16.44	226 (40%)	female
0.01>			diagnosis
	9.15±12.41	276 (48.8%)	Lung
	5.53±11.44	211 (37.3%)	dyspepsia
	11.26±17.75	60 (10.6%)	nephrology
0.01>			gland
	14.75±31.54	16 (2.8%)	
			Bed sore
0.01>	11.18±14	180 (35.9%)	yes
	7.68±13.61	385 (64.1%)	no
0.01>			diabetes
	11.50±18.69	93 (21.2%)	yes
0.01>	7.70±12.51	472 (78.8%)	no
			hypertension
0.01>	11.05±15.69	97 (21.1%)	yes
	7.77±13.28	468 (78.9%)	no
0.01<			Heart disease
	8.12±12.41	70 (12.3%)	yes
0.01>	8.16±13.93	495 (87.7%)	no
			infection
0.01>	10.67±13.48	184 (38.6%)	yes
	7.43±13.81	381 (61.4%)	no
0.01>			Nephrologic disorder
	11.66±16.89	68 (15.7%)	yes
0.01>	7.82±13.23	497 (84.3%)	no
			smoker
0.01>	11.66±5.69	63 (19.6%)	yes
	7.82±12.48	205 (80.4%)	no

Table 1. Findings related to clinical and demographic variables affecting the length of stay in ICU

doctrine disorders (14.7 ± 3.1 days) and patients with gastrointestinal disorders (5.53 ± 1.1 days) respectively ($p < 0.01$). The lowest age group was 12 while the highest was 99.

The intensive care unit included patients with pulmonary diseases (48.8%), dyspepsia (37.3%), nephrology disorder (10.6%) and endocrine disorders (2.8%). Patients with endocrine disorders (14.7 ± 3.1 days) and those with dyspeptic diseases (5.53 ± 1.1 days) had respectively the highest and lowest length of stay. The result shows that all patients had rehospitalization history. The result of the study showed that 48.3 percent of the patient ICU died. Diagram 1 shows the data distribution of the patients’ length of stay.

The distribution of the patients’ length of stay was from 0 to 122 days. Most patients (16.6%) stayed in intensive care unit 1 day and 6 percent of patients stayed zero (0) day. Ac-

According to Table 2, comparison between log-normal regression model and Gamma model showed that in Gamma model type of diagnosis ($\beta = -.5493982$; $CI = [-0.9413864, -0.15741]$) variable was significant. Factors to predict length of stay in ICU in Log – normal model:

a) Age ($\beta = 0.0200913$; $CI = [0.0121126, 0.0041343]$),

b) Sex ($\beta = 0.5652086$; $CI = [0.3062281, 0.472476]$), and

c) Type of diagnosis ($\beta = -.7543791$; $CI = [-1.298991, -0.2097672]$) were significant factors affecting the length of stay.

Gamma model showed that patient with dyspepsia had a significant shorter length of stay than other patients. According to AIC and BIC test, Gamma model was better than Log-normal model to predict factor affecting the LOS in ICU patients.

4. DISCUSSION

The present study aims to determine factors affecting the length of stay and selecting suitable statistical models. LOS data are right skewness to model skew data; we use Gamma and Log-normal regression to model skewed data (21). In this study, affecting factors LOS were assessed, the result of univariate regression analysis showed that increase of age of patients led to increase in LOS. Some study also showed that increase of patient age led to increase in LOS (22-28).

Most patients hospitalized in intensive care unit were elderly people, so managers and policy makers are recommended to make suitable plans and programs which can better provide necessary services for aging people.

The result of study showed that 35.9% of patient had a bed sore and the patient with Bed sore had longer LOS than other patients. Also some study showed that Bed sore led to increase LOS in ICU (22-24).

So due to the meaningful relation between bed sore and LOS by making suitable plans to control and prevent bed sore we can shorten the patients' length of stay and speed up their recovery. Considering the following points can help prevent and lessen bed sore in patients: Planning and implementing protocol in order to identify the patients in danger before admission to ICU. Paying special attention by nurses and hospital staff to elderly patients, women, patients with long periods of stay, those referred from other hospitals or centers, patients with physical disorders and patients with low consciousness, diabetes, fever, infection, hypertension and respiratory diseases.

Forming and activating a committee or community attended by nurses, doctors, units officials and high ranking hospital managers in order to hold regular meetings in which the bottom cause and the solution of this problem (bed sore) are discussed and units with more problems are identified.

Identifying patients in danger of bed sore before admis-

log-normal			Gamma			
P-Value	SE		P-Value	SE		
0.734	.7482738	-.2542611	Nephrologic disorder
0.564	.1835893	.1057881	0.831	.3490023	-.0745857	diabetes
0.713	.1392989	.051163	0.502	.2782064	-.1866855	Pulmonary infection
0.020*	.1321353	.3062281	0.605	.2362946	.1221271	sex
0.003**	.0040707	.0121128	0.722	.0054999	-.0019566	age
.....	0.086	.0017226	-.0029586	FBS1
.....	0.057	.0084936	.0161602	Na2
.....	(pulmonary) 3
0.006**	.1999976	-.5493982	0.007**	.2778683	-.7543791	(dyspepsia)
0.538	.7623625	.4698147	0.683	.3874695	.1580193	(kidney)
0.187	.2903324	.3831205	0.309	.7532947	-.7669147	(gland)
0.751	4.700433	-1.489488	(skin)
4550.664			1521.322			AIC4
4594.032			1557.232			BIC5

Table 2. Relation between LOS and demographic and clinical variables by using log-normal and Gamma regression. 1. Fast blood sugar, 2. Blood Sodium, 3. Reference Category 4. Bayesian information criterion 5. Akaike information criterion, * $P < 0.05$, ** $P < 0.01$

sion along with a particular attention to the factors creating it can be effective in preventing this problem. It seems that designing and using special forms for examining the risk of bed sore before admission is necessary. Other studies show that bed sore has a meaningful relation with an increase in length of stay in ICU (22-24). According to the result, 38.6% of patients had hospital infections. Univariate analysis showed that patient with hospital infection had significantly longer LOS than other patients. Some study showed that hospital infection led to increase LOS in ICU (25-27).

So because of the high rate of hospital infections and the significance of controlling them to reduce medical loss and the length of stay in ICU it is necessary to identify the factors which cause these hospital infections in order to take efficient steps that consequently lead to a decrease in LOS.

In other studies the meaningful effect of hospital infections on the length of stay in ICU was clearly demonstrated (25-27). In this study, risk factors including (diabetic, hypertension, and smoking) significantly impact on length of stay in ICU. Also some study showed that clinical and demographic variables significantly impact on LOS in ICU and hospital (22-28). Therefore hospital managers and clinicians can use the managerial and clinical methods to manage LOS patients with infection and Bed sores.

In order to decrease length of stay in ICU patients we used Gamma and Log-normal model. According to Log-normal model age, sex, and type of diagnosis were significant factors to predict length of stay in ICU.

According to Gamma model, Type of diagnosis variable was significant to predict length of stay in ICU. In this study BIC and AIC test showed that Gamma model was more robust than Log-normal model to predict affecting factors length of stay. The study by Saki et al showed Gamma model has a better efficiency, sensitivity and prediction ability to predict affecting factors (19).

5. CONCLUSION

The results of this study showed that in comparison to other statistic models Gamma model had stronger ability in

predicting the length of study. Choosing a suitable statistic model will enhance managers and policy makers' ability to organize hospital resources in an optimum way, furthermore comprehensive and exact information about the average length of stay can be useful for managers better physical, financial and strategic hospital planning.

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