

INFLUENTIAL VARIABLES INDUCING AERONAUTICAL REVENUES AT REGIONAL AIRPORTS

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Article history

Received

12 February 2015

Received in revised form

5 April 2015

Accepted

31 May 2015

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Graphical abstract



Abstract

Maximising revenues is one of the main objectives for regional airports to achieve in order to sustain their business, especially after the introduction of deregulation and privatisation of airports. Optimising aeronautical revenues has always been considered as a straightforward method and airport managers generally overlooked on the importance of daily operational factors that could influence the generation of aeronautical revenues. Thus, the objective of this paper is to validate whether the identified influential factors such as type of flight services offered, type of aircraft that airlines utilised, time of the day, day of the week the flights arrives or departs, and the number of passenger the airlines ferry in and out of the airport have an effect on the generation of aeronautical revenues. This study has adopted a quantitative approach using regression analysis to investigate the influence of these selected variables on the generation of aeronautical revenues. The findings indicate that the selected variables have positive and negative significant influence on the generation of aeronautical revenues at regional airports.

Keywords: Aeronautical revenues, regional airport, regression analysis

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1.0 INTRODUCTION

In this competitive environment, regional airports are expected to be financially stable in order to sustain their operations. Meanwhile, at the same time, regional airports need to fulfil various stakeholders' requirements. The main objective of most regional airports is to maximise their revenues and minimise operational cost in order to sustain their business. Airport revenues mainly stemmed from Aeronautical Revenues (AR) and Commercial Revenues (CR). AR are generated from the airside operations activities, which are directly related to the number of traffic and passengers' movements. CR are derived from terminal concessions, airport car parking, rental car, advertising, commercial development, land use etc.

Currently, most researchers focus more in finding new strategy and solutions to maximise commercial revenues [1-8]. On the other hand, less emphasis are given in searching for solution to maximise AR since it is considered as a straightforward problem. However, it is important to study the factors that influence the generation of AR since it is still the main source of revenues for regional airports [9-10].

2.0 METHODOLOGY

Based on interviews and case study conducted at Rotterdam the Hague Airport in the Netherlands, the external parameters that have been discovered to have influence on the generation of aeronautical

revenues in regional airports are operational modes, traffic types, time of day, day of week, engine type and aircraft MTOW (maximum take-off weight) category [11-12]. There were 29,269 traffic movements at Rotterdam the Hague Airport in 2009.

In this study, Pearson correlation and regression analysis were carried out using SPSS version 20 software to measure the relationship between the independent variables and the dependent variable (output

variable). The type of traffic, the time of day, day of week, engine type and MTOW category were classified as the input or independent variables, as shown in Table 1. Meanwhile, AR (*Total_Rev*) generated per flight was regarded as the output or dependent variable. MTOW was classified into three groups in accordance to Rotterdam the Hague Airport landing charges rate.

Table 1 Independent variables

Variables	Descriptions
Traffic types	Type of flights arrive and depart which includes Schedule flights, Business flights (B_1), Charter flights (C_1), Training flights (T_1) and Others flights (O_1).
Time of day	Day time where flights arrive or depart between 07:00 hrs to 23:00 hrs. Night time (N_1) is between 23:00 hrs to 07:00 hrs.
Day of week	Weekdays are from Monday to Friday. Weekends (W_1) are Saturday and Sunday.
Engine type	Jet engine and Propeller engine (P_1).
MTOW category	Light category is for MTOW below 6,000kg, Medium category is for MTOW between 6,000kg to 20,000kg and Heavy category is for MTOW above 20,000kg

Table 2 Pearson correlation

	Tot_Rev	B_1	C_1	T_1	O_1	N_1	W_1	P_1	MTOW
Tot_Rev	1	-.192**	.030**	-.237**	-.342**	.012*	.035**	-.518**	.733**
B_1		1	-.035**	-.148**	-.233**	.091**	-.058**	-.234**	-.197**
C_1			1	-.035**	-.054**	-.003	.009	-.001	.024**
T_1				1	-.230**	-.050**	.076**	.314**	-.347**
O_1					1	.025**	.080**	.277**	-.415**
N_1						1	.021**	-.111**	.037**
W_1							1	-.034**	.019**
P_1								1	-.743**
MTOW									1

**Correlation is significant at the 0.01 level (2-tailed)

*Correlation is significant at the 0.05 level (2-tailed)

3.0 CORRELATION ANALYSIS RESULT

From Pearson correlation analysis in Table 2, it could be seen that only MTOW has strong positive linear relationship with the total AR where the Pearson correlation coefficient (r) has the value of $r = 0.733$. Business (B_1), Training (T_1) and Others (O_1) flights indicate weak negative correlation with the total AR due to the r values which are small negative values and below 0.6 (weak correlation occurs if $0 \leq r \leq 0.6$). Charter flights (C_1) has weak positive linear relationship with generation of total AR. Similarly, night flights (N_1) and weekends flights (W_1) also indicated weak positive relationship with the generation of total AR. However, the engine propeller type aircraft (P_1) indicates moderate negative relationship with the generation of AR ($r = -0.518$).

4.0 REGRESSION ANALYSIS RESULTS

As for regression analysis, the data was split into two groups which are arrival and departure since arrival and departure movements were derived from different sources for AR generation. When a dummy coding is used in regression analysis, the overall results will indicate whether there is a relationship between dummy variables and dependent variables. The values of the intercept and regression coefficients of the resulted regression model can be obtained using least squares estimation procedure.

Results for arrival model of the regression analysis, with a sample size of 14,637, is shown in Table 3. As $r = 0.990$ (r is denoted as R in Table 3), it indicates that there is a very strong linear relationship between independent variables and dependent variable. In addition, $r^2 = 0.981$ (*R-squared*) means that 98.1% of the variance in the total AR can be predicted from the independent variables. The prediction is supported by the adjusted r^2 (*Adjusted R-squared*), which is a corrected goodness-of-fit (model accuracy) measure

for linear models. The value of $r^2 = 0.981$, which is very close to 1, indicates that the model almost perfectly predicts values in the target field. The variable constant represents the y-intercept or the intercept of the regression equation and it is equal to the expected value of the dependent variable whenever the values of the independent variables are equal to zero. The

resulted estimated regression equation for the above result is as follows:

$$\hat{Y} = 64.606 - 55.606B_1 - 16.823C_1 - 92.839T_1 - 82.159O_1 + 378.409N_1 - 4.359W_1 + 30.666P_1 + 10.750MTOW \quad (1)$$

Table 3 Regression statistics for arrival

	Un-standardised Coefficients		Standardised Coefficients		T	Sig.
	β	Std.Error	Beta			
(Constant)	64.606	2.482			26.027	0.000
B_1	-55.606	1.969	-.055		-28.242	0.000
C_1	-16.823	4.283	-.005		-3.928	0.000
T_1	-92.839	1.672	-.092		-55.514	0.000
O_1	-82.159	1.440	-.107		-57.058	0.000
N_1	378.409	2.489	.177		152.061	0.000
W_1	-4.359	.926	-.005		-4.706	0.000
P_1	30.666	1.630	.045		18.813	0.000
MTOW	10.750	.037	.905		292.952	0.000
R		.990				
R-squared		.981				
Adjusted R-squared		.981				
Std. Error of the Estimate		46.90734				

For B_1 variable, the t -test is significant, and the value of β (Beta for un-standardized coefficient) is -55.606, which means that the AR generation decreased significantly more for accepting Business flights than Schedule flights. For C_1 , the β value also has a negative value but the AR generation by Charter flights is 16 times less compared to schedule flights. T_1 , O_1 also has negative value of β , which indicate that Training flights, generate 92 times less AR than Schedule flights and Others flights generate 82 times less AR compared to Schedule flights. The t -test shows all are significant. N_1 gives a positive beta value, which indicates that night flights can earn 378 times more AR compared to day time flights due to the additional landing charges received from night surcharge penalty. However, weekends have negative β , which means that it

contributes less 4.359 of AR compared to weekdays flights because there are more flights during weekdays. P_1 , has positive β value of 30.666, which means that propeller engine aircraft generate more AR than jet engine aircraft. MTOW is the only continuous data in this analysis and it also indicates a positive β value. It shows that there is a direct relationship between the AR and MTOW.

As for departure flights results with sample data size $N = 14,632$ in 2009, Table 4 confirms that, by having R-squared of 0.745, there is a strong linear relationship between the selected variables and the generation of AR. The resulted estimated regression equation for the above result is as given in equation (2) below:

$$\hat{Y} = 371.419 - 632.442B_1 + 134.893C_1 - 272.018T_1 - 466.607O_1 - 197.417N_1 + 133.506W_1 - 180.239P_1 + 29.648MTOW \quad (2)$$

Table 4 Regression statistics for departure

	Un-standardised Coefficients,		Standardised Coefficients,		T	Sig.
	B	Std.Error	Beta			
(Constant)	371.419	33.386			11.125	0.000
B_1	-632.442	26.363	-.169		-23.990	0.000
C_1	134.893	60.329	.009		2.236	0.025
T_1	-272.018	22.578	-.072		-12.048	0.000
O_1	-466.607	19.374	-.164		-24.085	0.000
N_1	-197.417	58.846	-.014		-3.355	0.001
W_1	133.506	12.666	.045		10.540	0.000
P_1	-180.239	21.891	-.071		-8.233	0.000
MTOW	29.648	.493	.672		60.099	0.000
R		.863				
R-squared		.745				
Adjusted R-squared		.745				
Std. Error of the Estimate		633.70996				

The intercept value for the least square equation is 371.419 for departure, which is of higher value than arrival. This means that airport receive more AR generated by departure activities. The charges received from departing passengers are more substantial than aircraft landing charges. Business flights (B_i) has negative value which indicates that for every movement of Business flight, it loses €632.442 per flight because it has lower load factor compared to Schedule flights. Charter flights (C_i) gives a positive value which means that every Charter flight earns additional AR of €134.893 per flight as compared to Schedule flights. Charter flights have higher load factor compared to Schedule flights. Training and Others flights (T_i and O_i) also give negative β value which indicates that they lose €272.02 and €466.61 per each flight respectively compared to Schedule flights. Night time flights (N_i) are unfavourable than day time because passenger do not prefer to depart at late hours except for business executives, where they choose to arrive at their destination early morning. Weekends (W_i) gives positive indication because leisure passengers and travellers prefer to visit friends and relatives or go for holidays during the weekends. Schedule and Charter traffics prefer to depart during weekends and business passengers using schedule services also probably prefer to depart on weekends so they would arrive at their destination before Monday morning. Departing passengers seem to prefer flying in a jet engine aircraft compared to propeller engine aircraft and most jet engine aircrafts have higher seat capacity to carry more passenger per flight. The MTOW of the aircraft definitely has a positive significant on the generation of passenger departing fees since most large aircraft has higher seating capacity which can carry more passenger per flight.

5.0 CONCLUSION

Airports revenues are derived from aeronautical and commercial revenues. Although currently regional airports seem to focus more in finding new resources of generating commercial revenues, aeronautical revenues still remain the core income for regional airports. The generation of aeronautical revenues of regional airports are highly dependent on the number of flights landed and the number of passenger departs at and from their airports, thus it is important for airport management to analyse in detail the variables that contribute to the generation of aeronautical revenues for better planning and managing the airside operations and flights scheduling. Based on the correlation analysis conducted in this study, it was found that all the selected variables (traffic types, time of day, day of week, engine type and MTOW category) are correlated with the generation of aeronautical

revenue, either positively or negatively. The correlation results are also in line with regression results, where all the selected variables are found to be significance predictors of the generation of aeronautical revenues. By confirming these influential variables, it provides useful insights that can assist airport management in becoming more aware on which traffics types, the time of day, day of week, and type of aircraft that would attract more passengers, thus, they could plan and manage better their business operations and scheduling which would generate more aeronautical revenues for regional airports.

Acknowledgement

We would like to thank Universiti Teknologi MARA (UiTM) for the opportunity and support given to us to conduct this research and we are grateful to Ministry of Education Malaysia (KPM) for funding this research (600-RMI/FRGS 5/3 (153/2014))

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