

ANALYSIS AND FORECAST OF TOURIST NIGHTS IN CROATIA BY 2022

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Abstract: *The tourist nights variable is one of the most popular measures of tourism demand. In accordance the main objective of this paper is to estimate the number of tourist nights in Croatia by 2022. The analysis of the number of tourist nights in Croatia is based on the methods of descriptive statistics. To estimate the number of tourist nights in Croatia the moving averages methods and method of exponential smoothing are used. This methods are efficient and useful in the short run period. The main finding of this paper points to the fact that the number of tourist nights in Croatia will continue to show a growth trend in the coming period. Findings in this paper provides a good starting point to forecast tourism demand and could be helpful to tourist managers of all levels in making business decisions.*

Key words: *Croatia, tourism, tourist nights, forecast, moving-averages, exponential smoothing*

1. INTRODUCTION

Tourism is a significant source of export revenues for Croatia, it is vital for policymakers to understand the trends and factors that affecting tourism demand. Tourism demand is usually measured in terms of tourists arrivals, in terms of tourist's overnight stays, or in terms of tourist's expenditure. A large number of econometrics studies used, both time series and econometric approaches to forecast tourism demand. Good forecast are of critical in all aspects of tourist business especially for two activities: 1) human resources and 2) capacity.

Subject of this research is tourism demand measured by the number of tourist nights in order to prove the hypothesis that insufficient tourism demand is not a limiting factor for the development of Croatian tourism. Some of the world-famous Croatian tourist destinations such as Dubrovnik, Plitvice Lakes National Park or Krka National Park because of the pronounced seasonality of Croatian tourism have a problem of over-demand. To achieve the purpose and purpose of the research and to prove the set hypothesis in various combinations, several scientific methods were used, from which the method of descriptive statistics and two quantitative forecasting methods are used: moving averages and exponential smoothing method.

This scientific paper has five logically related parts. After the introduction, in the second part of the paper, the theoretical framework of the research, gives a detailed description of the methods of prediction used. The third part of the paper encompasses secondary data that are the subject of analysis in the work and the presentation of the scientific methodology used in the work. After the results of the research and discussion covered by the fourth part of the paper, the conclusion is reached.

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2. THEORETICAL FRAMEWORK OF RESEARCH

Decision making in the tourism industry is often based on the forecast of behaviour of a variable, such as the number of tourist arrivals, number of overnight stays, the price of tourist services, etc. Witt, Song and Anhill (2004) refer to the forecast of tourist demand as a key factor for planning and making decisions in tourism. Mervar & Payne (2007) in their study provide long-run elasticity estimates associated with the aggregate foreign tourism demand for Croatian destinations in the period 1994 – 2004 using the autoregressive distributed lag (ARDL) approach. Foreign tourism demand is appraised by the aggregate number of foreign overnight stays in Croatia. Bahovec, Dumičić and Čeh Časni (2008) in their work present an econometric model of Croatian tourism demand for the period between 1998 and 2007. Tourism demand is estimated by total tourist nights. Multiple Regression analysis was applied. Baldigara, Štambuk and Mamula (2013) endeavor to express a theoretical e-tourism demand model, as a mathematical function that indicates the presence of a relationship between the dependent variable, expressed by the number of tourists' overnight stays and a number of commonly used explanatory variables (income, prices, substitute prices). The authors also research the possibilities of adding some additional supplementary independent variables i.e. the number of internet users, the number of overnights stays that were booked online or the number of online reservations.

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In this paper two simple but very useful quantitative forecasting methods have been used: moving averages and exponential smoothing which are fall into time-series model. Time-series models predict on the assumption that the future is a function of the past.

Moving averages

Moving averages are useful if we can assume that market demands will stay fairly steady over time. A 3-years moving average is found by simply summing the demand during the past 3-years and dividing by 3. With each passing year, the most recent year's data are added to the sum of the previous 2 years and the earliest year is dropped. This practice tend to smooth out short-term irregularities into the data series.

Mathematically, the simple moving average (which serves as an estimate of the next period's demand is expressed as

$$\text{Moving average} = \frac{\sum \text{demand in previous } n \text{ periods}}{n} \quad (1)$$

where n is the number of periods in the moving average.

When a detectable trend or pattern is present, weights can be used to place more emphasis on recent values. This practice makes forecasting techniques more responsive to changes because

more recent periods may be more heavily weighted. Deciding which weights to use requires some experience.

A weighted moving average may be expressed mathematically as

$$\text{Weighted moving average} = \frac{\sum(\text{weight for period } n)(\text{demand in previous } n \text{ periods})}{\sum \text{weights}} \quad (2)$$

Exponential Smoothing

Exponential smoothing is a sophisticated weighted moving-average forecasting method. The basic exponential smoothing formula can be shown as follows:

$$\text{New forecast} = \text{last period's forecast} + \alpha(\text{last period's actual demand} - \text{last period's forecast}) \quad (3)$$

Where α is a weight or smoothing constant, chosen by the forecaster, that has a value between 0 and 1. Equation (3) can also be written mathematically as

$$F_t = F_{t-1} + \alpha (A_{t-1} - F_{t-1}) \quad (4)$$

Where:

F_t – new forecast

F_{t-1} – previous forecast

α – smoothing constant ($0 \leq \alpha \leq 1$)

A_{t-1} – previous period's actual demand

The smoothing constant, α , is generally in the range from 0,05 to 0,5 for business applications. It can be changed to give more weight to recent data (when α is high) or more weight to past data (when α is low).

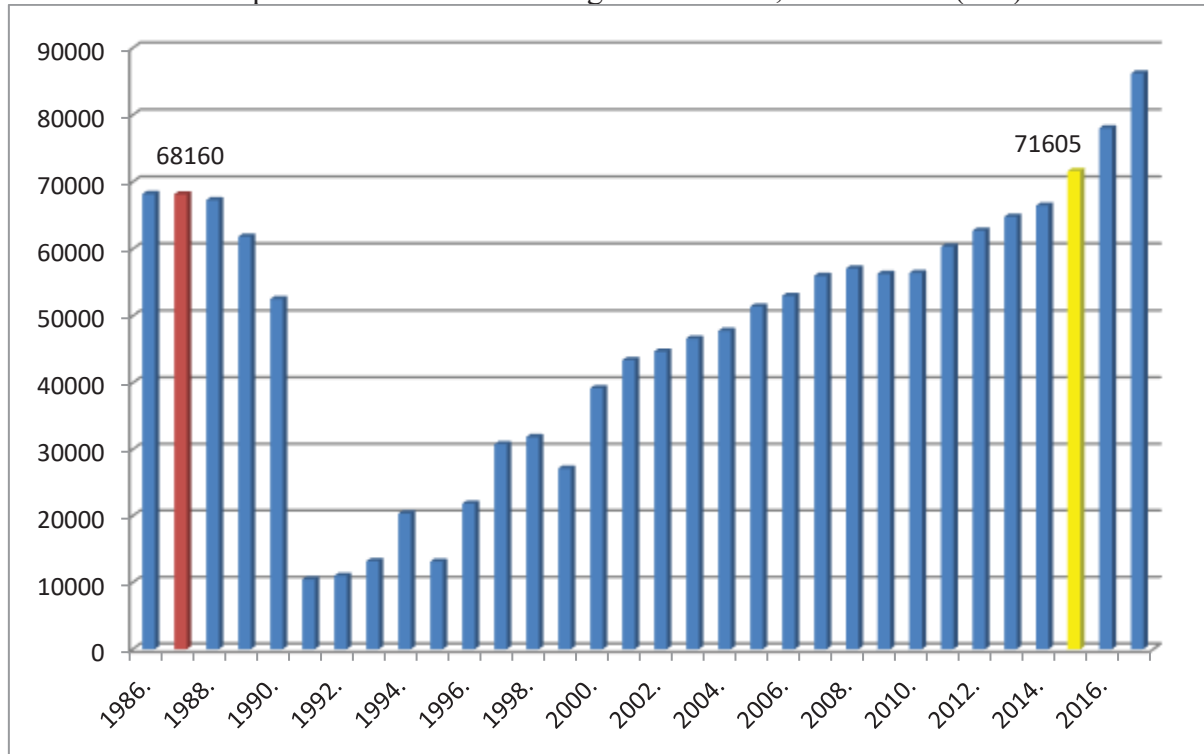
The overall accuracy of any forecasting model-moving average, exponential smoothing, or other – can be determined by comparing the forecast values with the actual or observed values. One of the most popular measures is mean absolute deviation (MAD). This value is computed by taking the sum of absolute values of the individual forecast errors and dividing by the number of periods of data (n):

$$\text{MAD} = \frac{\sum(\text{actual} - \text{forecast})}{n}$$

3. DATA AND RESEARCH METHODOLOGY

According to collected data, regarding the number of tourist nights, it is visible that the Republic of Croatia, in the year of 2015., has not only reached the results from the year 1987. but the number of tourist nights has been increased. (cf. Graph 1).

Graph 1. Number of tourist nights in Croatia, 1986 - 2017 (000)



Source: DZS, Statistical Yearbooks of the Republic of Croatia, different years

Based on the chart data, there is a great fragility of tourism on political, war and economic factors. Tourism demand is negatively affected by the political conflicts in the 1990s. In the post-war period from 1996 to 2017, almost continuous increase in the number of realized tourist nights has been observed. Exceptions to this positive trend occurred only in 1999 because of the military and political conflict in Kosovo, and for two years: 2009 and 2010, which coincide with the outbreak of the great economic crisis in the world as well as in Croatia, of course.

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The results of this paper are based on data on the number of tourist nights for the period from 1996 to 2017. Based on this data, first there will be a descriptive analysis of the number of tourist nights in Croatia for the period 1996 to 2017, and then will be used by 2022 in moving averages and model of exponential smoothing to predict the number of tourist nights in the Republic of Croatia.

4. RESEARCH RESULTS AND DISCUSSION

Based on the collected data (cf. Graph 1), a brief descriptive analysis of the number of tourist nights was made (cf. Table 1).

Table 1. Descriptive statistics of the number of tourist nights to Croatia, 1996 - 2017. (000)

Mean	52446,86
Standard Error	3478,60
Median	54506,00
Standard Deviation	16316,07
Kurtosis	-0,20
Skewness	0,01
Range	64340,00
Minimum	21860,00
Maximum	86200,00
Sum	1153831,00
Count	22,00

In the observed period, the Republic of Croatia had somewhat more than 1,15 billion tourist nights. The average annual number of tourist nights in the observed period was 52.4 million (SD = 16.3). The median value (M) was 54,5 million. The lowest number of tourist nights in the observed period was recorded in 1996, the first post-war year, amounting to only 21,86 million, while the highest number of tourist nights occurred in 2017 with 86.2 million tourist nights. That is an increase of 3,94 times.

Based on formula (1) moving averages are calculated. A 3-year moving average appears on the right (cf. Table 2).

Table 2. Estimate the number of tourist nights with moving average model (000)

Year	Actual tourist nights	3-year moving average	Forecast error
1996	21860		
1997	30775		
1998	31852		
1999	27126	28162,33	-1036,33
2000	39183	29917,67	9265,333
2001	43405	32720,33	10684,67
2002	44692	36571,33	8120,667
2003	46635	42426,67	4208,333
2004	47797	44910,67	2886,333
2005	51421	46374,67	5046,333
2006	53007	48617,67	4389,333
2007	56005	50741,67	5263,333
2008	57103	53477,67	3625,333
2009	56300	55371,67	928,3333
2010	56416	56469,33	-53,3333
2011	60354	56606,33	3747,667
2012	62743	57690	5053
2013	64818	59837,67	4980,333
2014	66484	62638,33	3845,667
2015	71605	64681,67	6923,333
2016	78050	67635,67	10414,33
2017	86200	72046,33	14153,67

2018	78618,33
2019	80956,11
2020	81924,81
2021	80499,48
2022	81126,53
MAD = 5691,46	

Thus we see that forecast for 2018 is 78 618,33 mil. tourist nights. To project demand for tourist nights in the 2018 we sum the 2015, 2016, 2017 and divide by 3.

Based on formula (2) we decided to forecast tourist nights by weighting the past 3 years as follows: 3-last year, 2-two years ago and 1-three years' age. Sum of weights are 6. The results of this weighted-average forecast are as follows:

Table 3. Estimate the number of tourist nights with weighted moving average model (000)

Year	Actual tourist nights	3-year weighted moving average	Forecast error
1996	21860		
1997	30775		
1998	31852		
1999	27126	29827,67	-2701,67
2000	39183	29309,5	9873,5
2001	43405	33942,17	9462,833
2002	44692	39284,5	5407,5
2003	46635	43344,83	3290,167
2004	47797	45449	2348
2005	51421	46892,17	4528,833
2006	53007	49415,33	3591,667
2007	56005	51610	4395
2008	57103	54241,67	2861,333
2009	56300	56054,33	245,6667
2010	56416	56518,5	-102,5
2011	60354	56491,83	3862,167
2012	62743	58365,67	4377,333
2013	64818	60892,17	3925,833
2014	66484	63382,33	3101,667
2015	71605	65305,17	6299,833
2016	78050	68766,83	9283,167
2017	86200	73974	12226
2018		81050,83	
2019		82266,83	
2020		82516,86	
2021		82189,1	
2022		82307,31	

MAD = 4793,13

Moving averages, however, present three problems: 1) Increasing the size of n (the number of periods averaged) does smooth out fluctuations better, but it makes the method less sensitive to real changes in data; 2) Moving averages cannot pick up trends very well. Because they are averages, they will always stay within past levels and will not predict changes to either higher or lower levels. This is, they lag the actual values; 3) Moving averages require extensive records of past data.

Based on formula (4) we estimate the tourist nights with exponential smoothing model. The following table shows the results for $\alpha = 0,8$.

Table 4. Estimate the number of tourist nights with exponential smoothing model (000)

Year	Actual tourist nights	Rounded forecast with $\alpha=0.8$	Forecast error
1996	21860	20000	1860
1997	30775	21488	9287
1998	31852	28917,6	2934,4
1999	27126	31265,12	-4139,12
2000	39183	27953,82	11229,18
2001	43405	36937,16	6467,835
2002	44692	42111,43	2580,567
2003	46635	44175,89	2459,113
2004	47797	46143,18	1653,823
2005	51421	47466,24	3954,765
2006	53007	50630,05	2376,953
2007	56005	52531,61	3473,391
2008	57103	55310,32	1792,678
2009	56300	56744,46	-444,464
2010	56416	56388,89	27,10712
2011	60354	56410,58	3943,421
2012	62743	59565,32	3177,684
2013	64818	62107,46	2710,537
2014	66484	64275,89	2208,107
2015	71605	66042,38	5562,621
2016	78050	70492,48	7557,524
2017	86200	76538,5	9661,505
2018		84267,7	

MAD = 3651,57

On the basis of analysis, an exponential smoothing forecast model is preferred because its MAD is smaller. Based on the forecasts made, a further increase in the number of realized tourist nights in the Republic of Croatia is visible. Presented models show the highest accuracy of the forecast at a time of slowed or unsettled growth, as it was during the time of the global economic crisis in 2009 and 2010. The reasons should be sought in the gap between actual and forecast values. Specifically, the projected values are almost always lagging behind for current values (usually one year). Thus, the forecast values can be taken with reserve, or as a pessimistic variant of the forecast of the number of tourist nights spent.

5. CONCLUSION

Tourism demand in the Republic of Croatia measured by the number of overnight stays has been growing steadily. The number of tourist nights spent over the last twenty years has increased almost four times. According to current knowledge and estimates made in this paper, tourism demand will certainly not be a limiting factor for the development of Croatian tourism. Progress made in this work suggests a further tendency of growth of tourist demand measured by the number of realized tourist nights. The increase of realized tourist nights by 2022 will be higher than the estimated number in this work. Estimated made in this paper represent pessimistic variant of the forecast. Croatia has an invaluable wealth of natural beauty that creates demand for itself. Consequently, the task of tourist workers is increasingly reduced to the management of tourist offer in order to exploit natural beauty in a quality, economic and ecologically sustainable way. The economy of natural beauty should be supported by quality tourism products based on tradition, culture, knowledge and creativity with the full respect and encouragement of entrepreneurship in tourism. Cultural and creative tourism should become a fundamental part of the tourist offer of the Republic of Croatia.

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