

Time Series Analysis of Road Traffic Accidents in Zimbabwe

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Abstract In this paper, focus is on finding a suitable model for the annual Zimbabwe Traffic Accident statistics from 1997 to 2013 and to forecast the number of annual traffic accidents likely to occur in future. The Box-Jenkins model building strategy is used. The Augmented Dickey Fuller test showed that the accident data was non-stationary. After first order differencing, the data became stationary. Three ARIMA models were suggested based on the ACF and PACF plots of the differenced series, these were ARIMA(0,1,0), ARIMA(1,1,0) and ARIMA(1,1,1). The model with the smallest corrected Akaike Information Criteria (AICc) and Bayesian Information Criteria (BIC) was chosen as the best model. The Ljung-Box statistics among others were used in assessing the quality of the model. ARIMA (0,1,0) was the best model for the Zimbabwe annual Traffic Accident data. Forecasting retained the value at the forecast origin. The implications of these findings are that based on the annual road traffic accident data for the period under consideration, it is difficult to make reasonable forecasts of the number of road traffic accidents for the years ahead of 2013. This is due to the fact that the values at different times of a white noise process are statistically independent.

Keywords Zimbabwe, Traffic accidents, Stationarity, ARIMA, Forecast, White noise

1. Introduction

The state of roads in Zimbabwe is appalling. The Zimbabwe National Roads Authority (ZINARA) is failing to cope with the much needed rehabilitation and repairs of roads. The problem has been compounded by harsh economic conditions prevailing in the country. Potholes are littered around all major roads causing accidents to motorists and pedestrians alike. The total number of accidents in Zimbabwe went up by 61.2% from 2009 (17,388) to 2010 (28,037) and the number of deaths increased in the same period by 7.3% and by 12.8% from 2010 to 2011. [1] About 50 million people and 1.2 million people in the world are killed annually due to road traffic accidents (RTA). [2]

The Zimbabwe vehicle population (VP) has been increasing annually from 2009 to date due to the dollarization of the economy which made it easy for individuals to acquire second hand vehicles mainly from Japan. Such an increase in VP was not followed by road expansion resulting in congestion in urban areas and minor accidents. Fatal accidents have mainly been confined to the highways. Road accidents have been increasing in developing countries and Africa in particular. [3], [4], [5] Significant progress towards prevention and control of RTA

has been limited to high income and/or highly industrialised countries ([6], [7]) in [8].

There are many causes of road traffic accidents, among them human error, vehicle conditions, road environment, over-speeding, road users and many more [9]. [10] in [5] suggest use of alcohol as one of the causes of accidents while [11] gives too much travelling in summer as a major cause of accidents. Zimbabwe has tightened laws governing the driving of public motor vehicles (PMV) by putting the minimum age of a PMV driver at 25 years with a minimum of 5 years driving experience and introducing re-tests every 5 years among other measures. Enforcement of such measures on the PMV drivers has been compromised by both corruption in the police force and low un-prohibitive fines of up to a maximum of twenty American dollars per count.

The Box-Jenkins model building strategy (1976) has been used to analyse accident data by several authors [4], [12], and [13] and found to be effective. Road traffic accident data in Zimbabwe was analysed using generalised linear models [14]. The Box- Jenkins model will be used in this paper.

2. Methodology

2.1. Source of Data

The data was obtained from Zimbabwe Parliamentary report of the Portfolio Committee on Transport and Infrastructural Development on The Causes of Road Carnage, May 2014 report covering the period 1997 to 2013.

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The data is based on reported traffic accidents. R software was used in the analysis.

2.2. The Box-Jenkins Model

The Box-Jenkins time series model was employed. Unit root tests were used to test for stationarity of the data and the non-stationary series (Z_t) was differenced d times to obtain a stationary series. Z_t is thus an ARIMA (p, d, q) model that conforms to the relationship:

$$\phi(B)(1-B)^d Z_t = \theta(B)a_t \quad (1)$$

Where:

$$\phi(B) = 1 - \phi_1 B - \phi_2 B^2 - \dots - \phi_p B^p \quad (2)$$

$$\theta(B) = 1 - \theta_1 B - \theta_2 B^2 - \dots - \theta_q B^q \quad (3)$$

B is the backshift operator, p is the number of autoregressive parameters in the model while q represents the number of moving average parameters in the same model, and d is the differencing parameter signifying how many times the series has been differenced in order to achieve stationarity. Details of ARIMA (p, d, q) models can be found in [15] (pp 40-43).

The autocorrelation function (ACF) and partial autocorrelation function (PACF) were used to determine the tentative time series models (model formulation). R software was used to estimate the parameters. The tentative model(s) were then subjected to rigorous diagnostic checks using residuals: ACF and PACF plots of residuals. The Akaike Information Criteria (AIC) and Bayesian

Information Criteria (BIC) were used to select the best model. The final model was then used to make forecasts for 2014, 2015, 2016, 2017 and 2018.

3. Results and Discussion

3.1. Data Analysis and Tests for Stationarity: Total Reported Accidents

The Augmented Dickey Fuller test for stationarity was used to test for stationarity. [16], [17]

The TS-plot of reported annual traffic accidents (ATA) (figure 1) is clearly non stationary. The augmented Dickey Fuller test on the data showed a p-value of 0.4075 suggesting further that the ATA data is non stationary. The mean is not constant and seems to be decreasing with time. After differencing once, the data became stationary with $p=0.01$ in the ADF test. [11], [12], [4], [13]. The ACF and PACF plots of the differenced series were obtained and appear in the following graphs (figure 2 and figure 3).

3.2. Model Selection and Diagnostics

The ACF and PACF plots show no significant spike at any given lag suggesting that the differenced series could be a white noise process. Three different models were tried and these are ARIMA(0,1,0), ARIMA(1,1,0) and ARIMA(1,1,1). Table 1 below shows the results of the analysis for the three models.

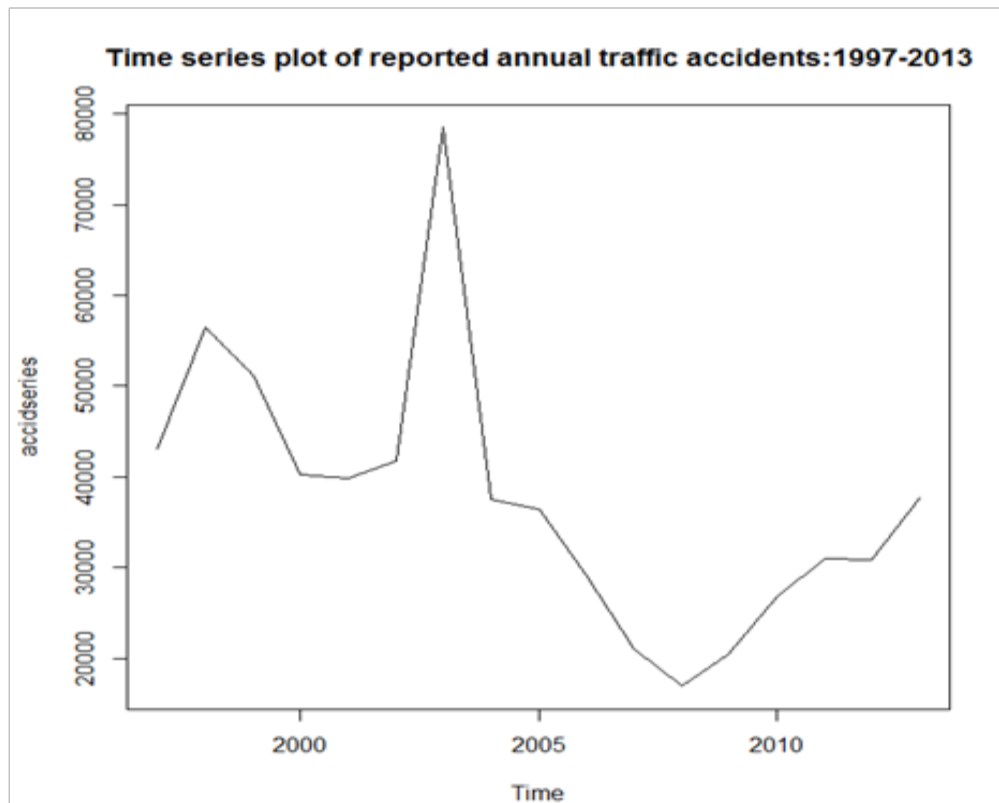


Figure 1. Time series plot of reported annual traffic accidents

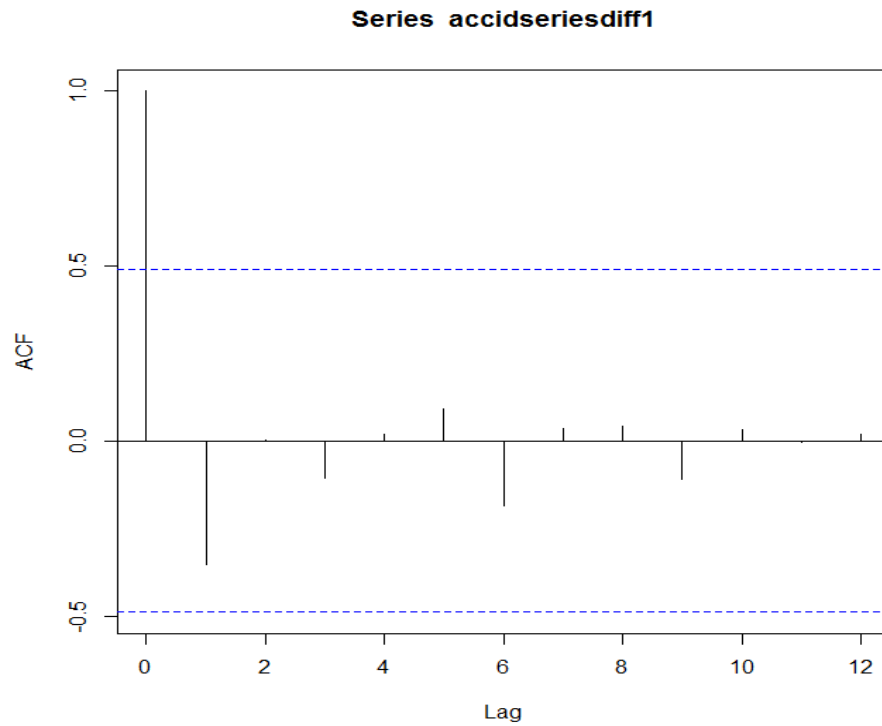


Figure 2. ACF plot of reported annual traffic accidents

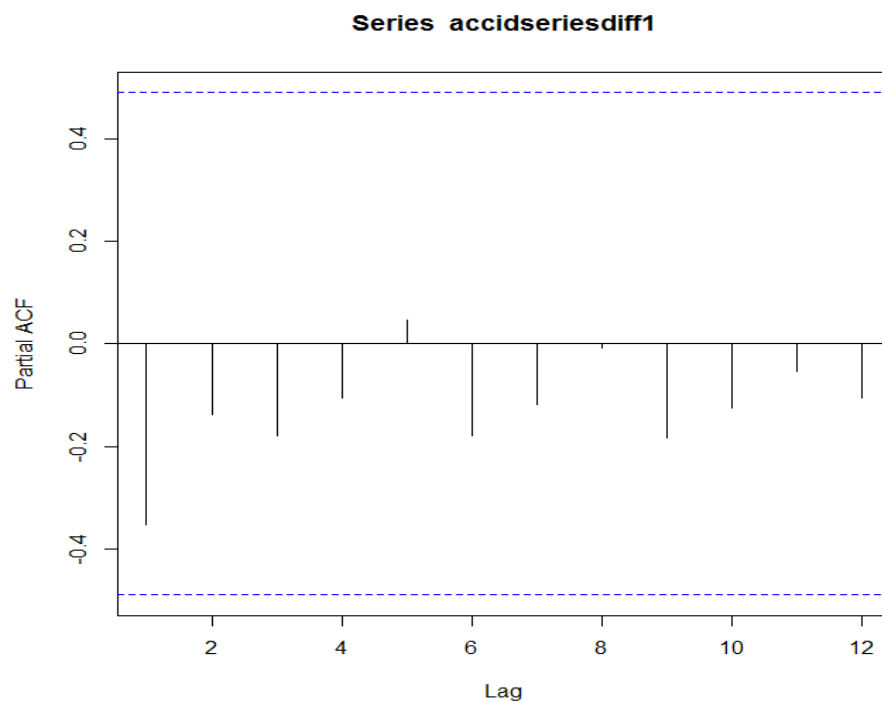


Figure 3. PACF plot of reported annual traffic accidents

Table 1. Information criteria and Ljung-box p-values for the three models

Model	AIC	AICc	BIC	Ljung-Box P-value
ARIMA(0,1,0)	355.11	355.4	355.8	0.9942
ARIMA(1,1,0)	355.03	355.95	356.57	0.9998
ARIMA(1,1,1)	356.08	358.08	358.4	0.9997

Based on the corrected Akaike Information criteria (AICc) and the Bayesian Information criteria (BIC), an ARIMA(0,1,0) has the lowest AICc=355.4 and BIC=355.8 and was thus chosen as the tentative model for the ATA data [11], [12]. The Ljung Box test for the residuals of ARIMA(0,1,0) shows no evidence of non-zero correlations in the residuals with $p=0.9942$. An ADF test with $p=0.02253$

suggests that the residuals are stationary. The ARIMA(0,1,0) model was thus confirmed as the final model for the data. An `auto.arima()` function in R also yielded an ARIMA(0,1,0)

model for the ATA data.

An ACF plot (figure 4) of the residuals for an ARIMA(0,1,0) showing no significant spike.

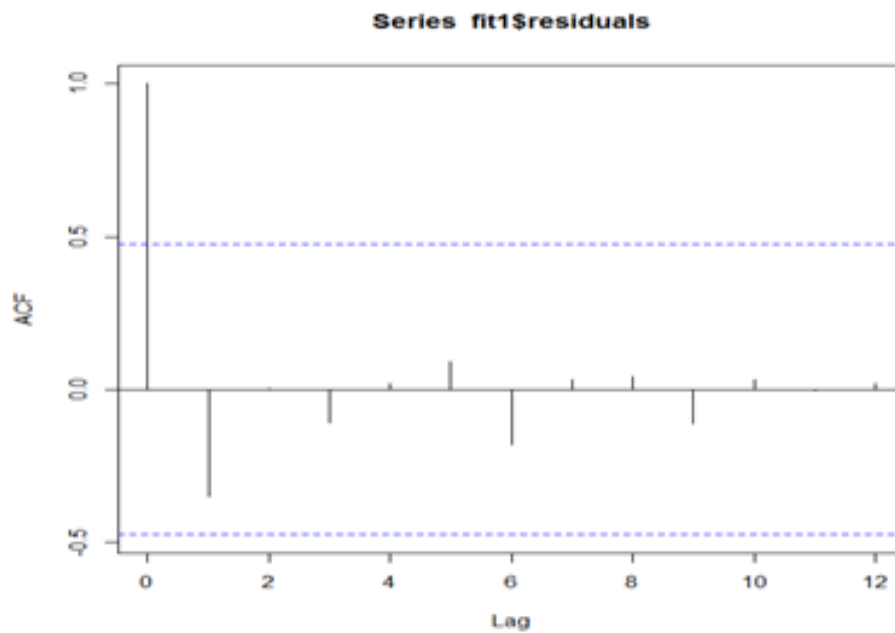


Figure 4. ACF plot of residuals for ARIMA(0,1,0) model

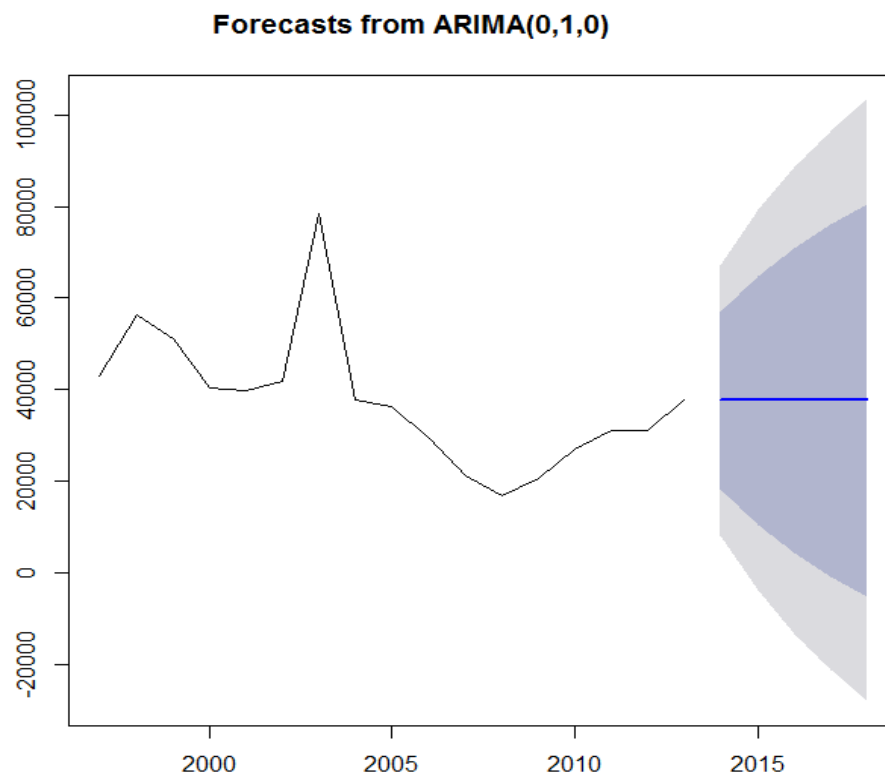


Figure 5. The 80 % (dark shaded band region) and 95% (light shaded band region) confidence bands for the forecast of Zimbabwe Traffic accidents

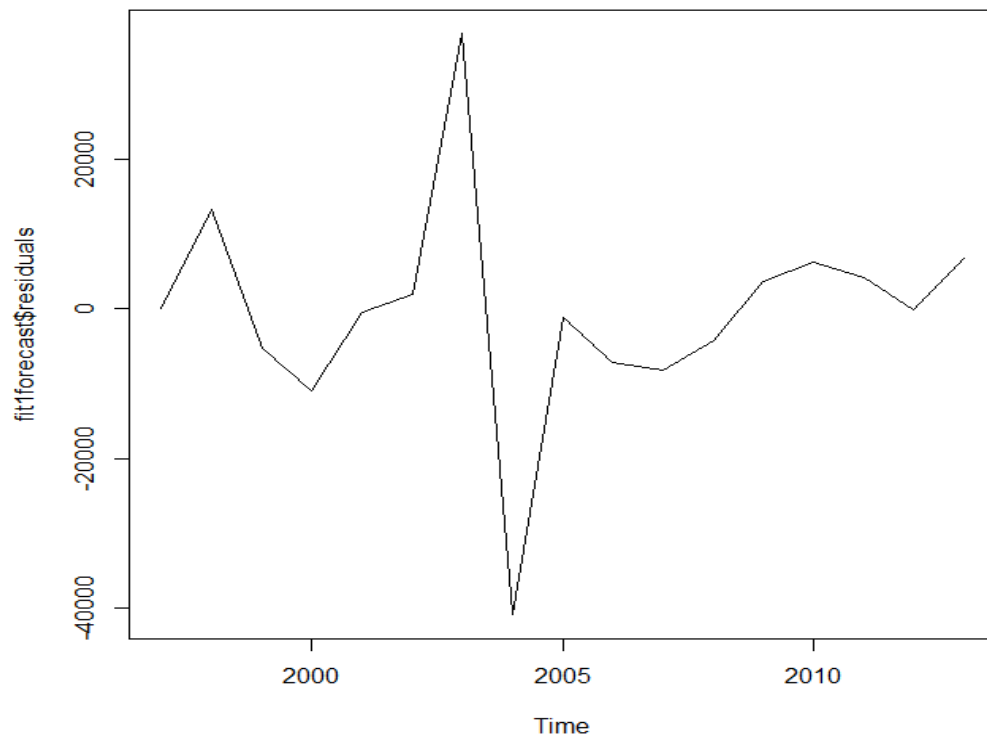


Figure 6. Time series plot of forecast residuals

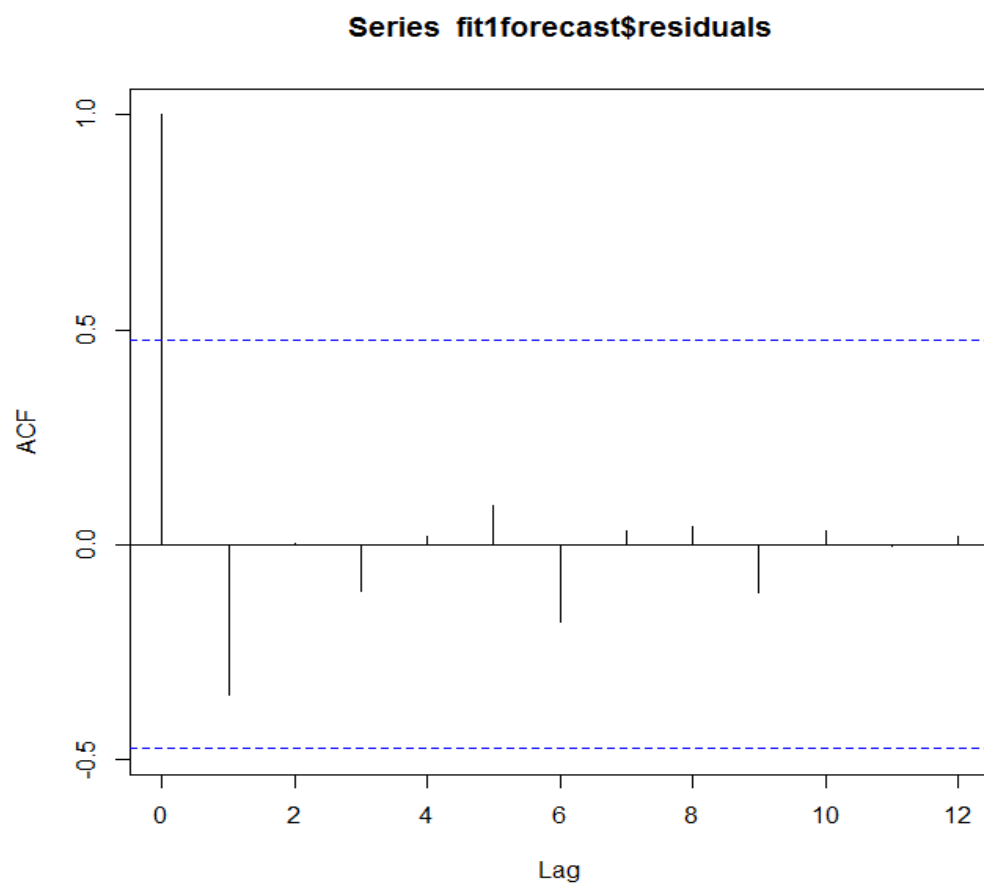


Figure 7. ACF plot of forecast residuals

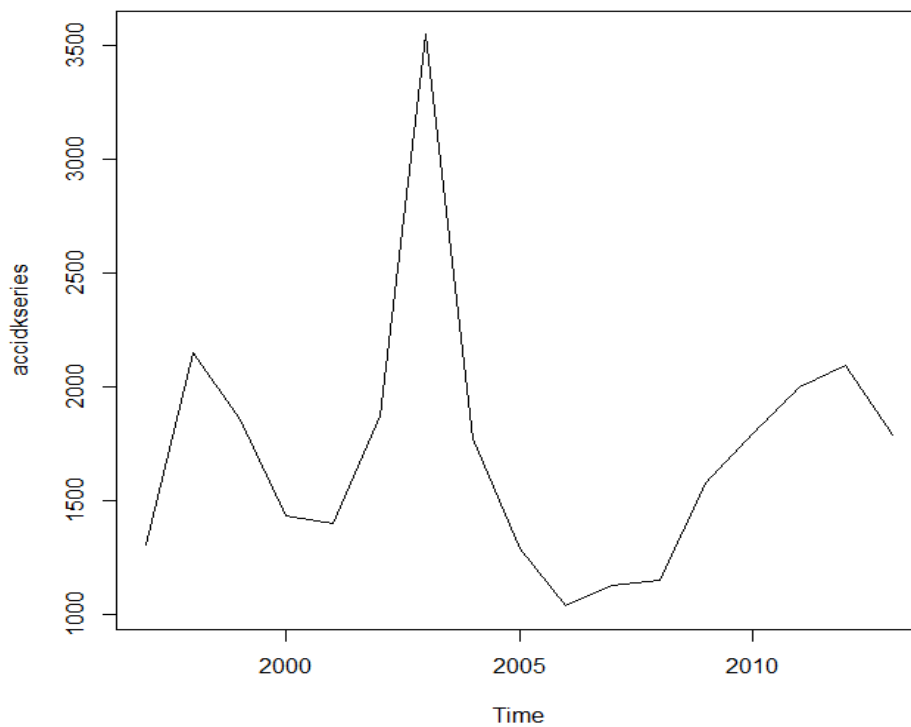


Figure 8. Showing Number of persons killed in road accidents

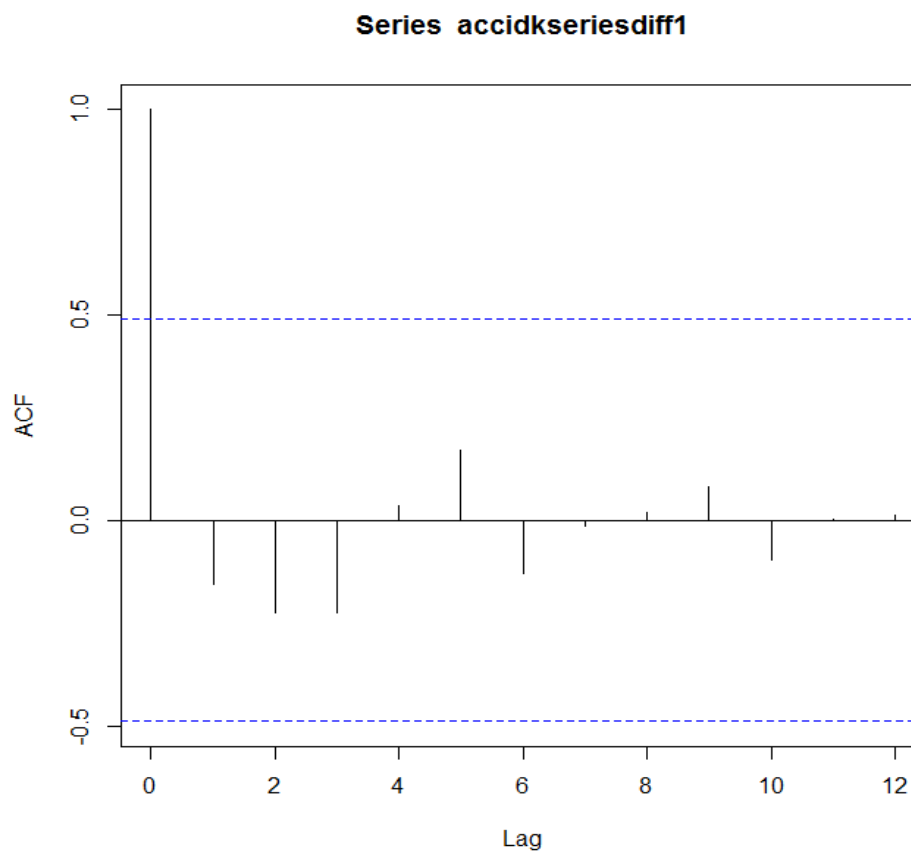


Figure 9. ACF plot of differenced series of persons killed in accidents

3.3. Forecasting

As was expected, forecasting Zimbabwe Accident data yielded a value at the forecast origin, 37 619 accidents which corresponds to the number of accidents in 2013. Suppose Z_t is the accident data series, then differencing once gives $Z_t - Z_{t-1}$, which in this case was found to be white noise implying that $Z_t - Z_{t-1} = a_t$. The following table (table 2) shows forecasts for 2014-2018 and confirms the above theory.

The in-sample forecast errors in figure 6 above seem to have a constant variance over time.

Figure 7 shows ACF plot of the forecast errors with no significant spike suggesting lack of serious correlations among them. It also means the model may not be improved any further. The Box-Ljung statistic for the forecast errors is 4.6044 with a p-value of 0.9699 implying that there is no evidence of non-zero autocorrelations in the in-sample forecast errors.

Table 2. Forecasts and their corresponding 80% and 95% confidence limits

Year	Point forecast	Lo80	Hi80	Lo95	Hi95
2014	37619	18395.486	56842.51	8219.172	67018.83
2015	37619	10432.846	64805.15	-3958.635	79196.63
2016	37619	4322.897	70915.10	-13302.995	88541.00
2017	37619	-828.028	76066.03	-21180.655	96418.66
2018	37619	-5366.084	80604.08	-28121.013	103359.01

3.4. Data Analysis and Tests for Stationarity: Persons Killed in Accidents

In this section, analysis is focused on persons killed in road traffic accidents.

The shape of the plot is the same as that for the total number of reported accidents. The number of people who died in road accidents increased from 1399 in 2001 reaching a peak of 3549 in 2003 which also corresponds to the high number of reported road accidents of 78 481 in 2003. The series is clearly non-stationary since its mean appears to be changing with time. Testing for stationarity using the ADF test gave a $p = 0.07759$ thus failing to reject the null hypothesis that the number of persons killed series is unit root non-stationary. The series became stationary after differencing once. An examination of the ACF and PACF of the differenced series follows in figures 9 and 10.

Both the ACF and PACF show no significant spike. ACFs or PACFs are significant if they exceed $\frac{1.96}{\sqrt{k}} = 0.47537$ for $k=17$.

Since none of the values of the ACF in table 3 exceeds 0.47537, the ACFs are not significant. This suggests an ARIMA (0,1,0) model, the same model fitted to the reported accidents data. A comparison of the AICc for the three possible models described earlier is based on data in table 4.

Clearly the AICc suggest the ARIMA(0,1,0) as a tentative model for the data. The Ljung-Box statistics showed non-zero correlations in the residuals with $p=0.964$. The residuals for this were found to be unit root stationary using the ADF test with $p=0.01$. Forecasting for this model was not necessary since it was going to give the value at the forecast origin as the forecast values for 2014-2018.

Table 3. ACF values for the number of persons killed in traffic accidents in Zimbabwe

Lag	ACF	Lag	ACF
0	1	7	-0.015
1	-0.156	8	0.020
2	-0.225	9	0.083
3	-0.225	10	-0.097
4	0.037	11	0.004
5	0.170	12	0.012
6	-0.128		

Table 4. Information criteria for the proposed models

	ARIMA(0,1,0)	ARIMA(1,1,0)	ARIMA(1,1,1)
AICc	257.37	259.6	258.38

4. Conclusions

The research used annual data from the Zimbabwe Parliamentary report of the Portfolio Committee on Transport and Infrastructural Development on The Causes of Road Carnage, May 2014 report covering the period 1997 to 2013. The data was composed of three groups: total reported cases, persons killed in road accidents and persons injured. The data was unit root non-stationary based on results from the Augmented Dickey Fuller (ADF) test. First order differencing forced the data to become stationary. The Box-Jenkins model building strategy (1976) was applied on differenced series. The ACF and PACF of the differenced series gave a guideline on tentative models to be used. A spike in both plots (ACF and PACF) was considered significant if it was greater than 0.47537. There were no significant spikes in the ACF and PACF plots for the two data sets analysed. Three models were suggested: ARIMA(0,1,0), ARIMA(1,1,0) and ARIMA(1,1,1). The ARIMA model with the smallest AICc and BIC was chosen as the best model. Ljung-Box statistics were also used in assessing the quality of the model. ARIMA (0,1,0) emerged as the best model for Zimbabwe Total reported Traffic Accident cases as well as for the number of persons killed. Due to the nature of the model used which uses the value at the forecast origin as a forecast, forecasts for 2014-2018 yielded the same result. The implications of these findings are that making forecasts using a white noise process is difficult due to the fact that the values at different times are statistically independent.

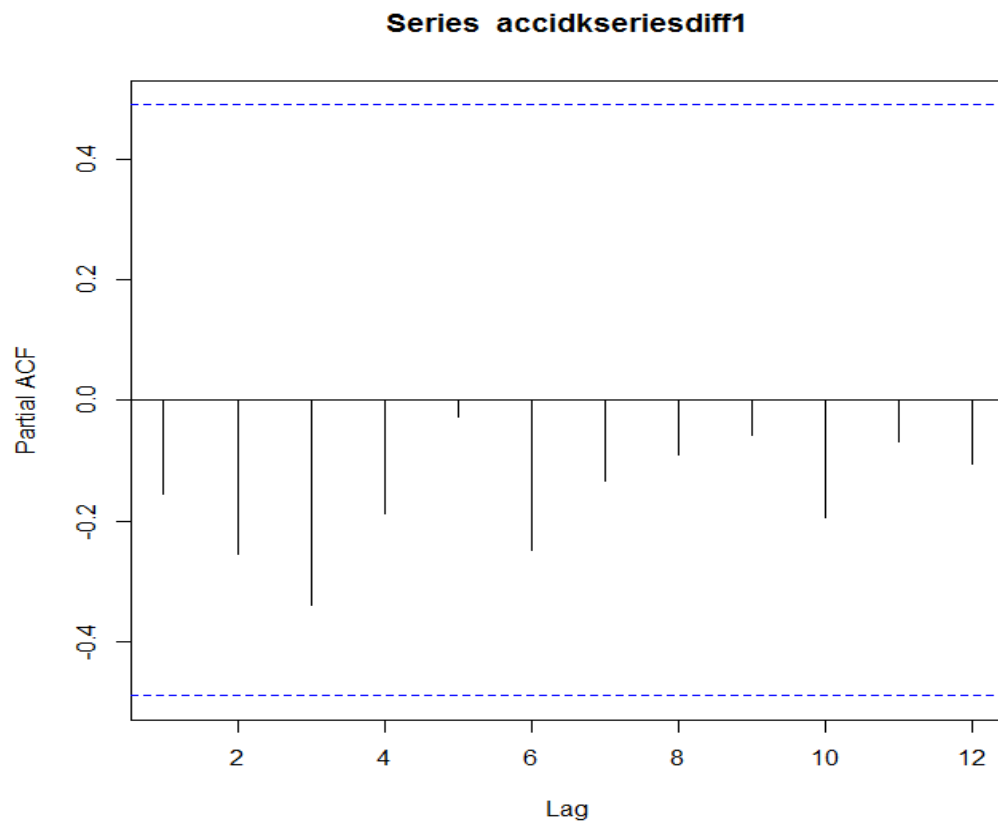


Figure 10. PACF plot of the differenced series of persons killed in accidents in Zimbabwe

Appendix 1

Zimbabwe Road Traffic Accident Data

Year	Total reported cases	Number of persons killed	Number of persons injured
1997	43089	1307	17906
1998	56433	2152	25984
1999	51219	1858	23722
2000	40316	1433	18105
2001	39841	1399	18153
2002	41753	1871	20419
2003	78481	3549	37144
2004	37596	1771	17577
2005	36390	1291	18531
2006	29250	1037	13819
2007	21092	1127	10378
2008	16904	1149	10427
2009	20553	1576	12354
2010	26841	1796	14336
2011	30985	2001	15305
2012	30911	2094	14965
2013	37619	1787	14131

Source: Zimbabwe Parliamentary report of the Portfolio Committee on Transport and Infrastructural Development on The Causes of Road Carnage, May 2014 report covering the period 1997 to 2013.

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