# The research of prediction of future violent crime rate

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**Abstract.** Crime is a destabilizing factor in our society. However, if the government could understand how the violent crime rate would change, they could enact proper policy to cope with it. In this case, this research will predict the future violent crime rate based on the data from 1991 to 2020 since 1991 is an important turning point of American crime rate. With R-Studio, this research analyzes the effectiveness of the average method, naïve method, drift method, linear model, and autoregressive moving average model (Arima model) in the prediction and selects the best one with the smallest residuals and negligible autocorrelation. Based on the analysis, the Arima (0, 2, 0), which is one type of Arima model, does the greatest job in the prediction. According to the model this research eventually uses, the violent crime rate will start to increase from 2021 to 2025, showing that the violent crime condition may get worse. As a result, the government should pay more attention to violent crime to maintain the social security.

**Keywords:** Prediction, violent crime rate, autoregressive moving average model.

# 1. Introduction

Crime is an important factor that would severely threaten the society. In order to maintain the social security, proper policy should be designed and enforced by the government. In this process, if characteristics of crime, such as reasons, trends, and geographical distribution, could be clearly learned through research based on statistical tool, better policy is more likely to be made and help undermine the crime rate, which could greatly benefit the society.

Several studies have researched the trend of crime rate in the history. According to Tseloini et al, five major types of crime including residential burglary, car theft, theft from cars, personal theft, and assault, declined between 1988 and 2004 in 26 countries based on the data from the International Crime Victim Survey. However, the survey focused on the European and North American countries, which could not show the global trend [1]. Besides the international trend, several studies concentrated on American crime rate. According to Saridakis, from 1960 to 2000, population in the prison and merely social condition had little effect on the violent crime rate; besides, violent crime rate was related to income inequality and racial composition [2]. In the study conducted by Zimring, there was a great decline of American crime rate from 1991 to 2000, taking 1990 as a cut-off point. Specifically, each type of crime decreased by at least 23 percent [3]. Another research tried to study this phenomenon. Gramlich concluded that although the number of property and violent crimes fell a lot, the public believed condition of crimes got worse. In addition, a proportion of crimes were not reported by the police, which might contribute to the great decline [4].

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Crime rate is strongly associated with other factors. Greenberg discussed the Cantor-Land approach, which focused on the relationship between crime rate and Unemployment rate. Although unemployed people were motivated to commit crimes for earning, they would also stay at home to protect themselves [5]. Chintrakarn explored the relationship between income inequality and crime rate. Based on the panel analysis, as income inequality rised, people needed more protection, and thus the crime rate would decrease [6]. Light et.al compared the crime rate among illegal immigrants, formal immigrants, and natives. According to the research, undocumented immigrants had lower crime rate compared to the other two categories since they were afraid of being detected as well as hoped for economic success. Therefore, the punishment on illegal immigrants might not affect crime rate. However, the crime rate of the descendants of illegal immigrants was similar to the locals probably due to assimilation theory [7]. Significant global events would also influence the American national crime rate. According to Boman, Covid 19 epidemic caused the overall decline in the crime rate after the government imposed strict restriction to resist the epidemic because lockdown policy prevented group of criminals from committing crime together. However, individual crimes were not affected by Covid 19 [8].

The studies above analyzed the historical trend and the factors contributing to the crime rate. At the same time, forecasting the crime rate is also a very promising field. In this way, this study will analyze different time series models in predicting violent crime rate and find out the most efficient one. Finally, the predicted result could be obtained.

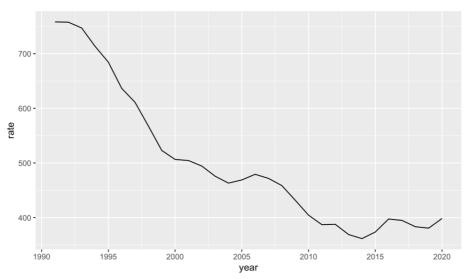
# 2. Methods

#### 2.1. Data source

In this research, the data on violent crime rate each year is from the Crime Data Explorer on the website of Federal Bureau of Investigation (FBI). The crime rate data in this website is collected from the Summary Reporting System (SRS) and the National Incident-Based Reporting System (NIBRS).

# 2.2. Data selection

In this research, violent crime rate data from 1991 to 2020 is used to predict the future crime rate. According to the studies of Zimring, 1991 was an important turning point when the crime rate started showing a sharp decreasing trend [9]. In this way, it could be a desirable starting point. At the same time, the crime rate in 2020 is the latest data FBI reports. As a result, this research will use the data before 2020 to predict the violent crime rate from 2021 to 2025 (Fig 1 and Table 1).



**Figure 1.** Violent crime rate in each year.

| Year | Rate  | year | rate  |
|------|-------|------|-------|
| 1991 | 758.2 | 2006 | 479.3 |
| 1992 | 757.7 | 2007 | 471.8 |
| 1993 | 747.1 | 2008 | 458.6 |
| 1994 | 713.6 | 2009 | 431.9 |
| 1995 | 684.5 | 2010 | 404.5 |
| 1996 | 636.6 | 2011 | 387.1 |
| 1997 | 611   | 2012 | 387.8 |
| 1998 | 567.6 | 2013 | 369.1 |
| 1999 | 523   | 2014 | 361.6 |
| 2000 | 506.5 | 2015 | 373.7 |
| 2001 | 504.5 | 2016 | 397.5 |
| 2002 | 494.4 | 2017 | 394.9 |
| 2003 | 475.8 | 2018 | 383.4 |
| 2004 | 463.2 | 2019 | 380.8 |
| 2005 | 469   | 2020 | 398.5 |

**Table 1.** Violent crime rate in each year [9].

# 2.3. Research protocol

This study will test five different approaches to figure out the best one, and they are the average method, naïve method, drift method, linear model, and autoregressive integrated moving average model (Arima model). This research will program in R-Studio to construct the models and check their accuracy, including checking the range and autocorrelation of residuals [10]. Following, the research would use the best model to obtain the future prediction.

2.3.1. The average method. Prediction is equal to the average value of the data set, and the equation is:

$$\hat{y}_{T+h|T} = \bar{y} = (y_1 + \dots + y_T)/T \tag{1}$$

2.3.2. The naïve method. Prediction is equal to the latest value in the data set, and the equation is:

$$\hat{y}_{T+h|T} = y_{T+h-m(k+1)} \tag{2}$$

2.3.3. The drift method. Prediction is based on the average change in historical data, and the equation is:

$$\hat{y}_{T+h|T} = y_T + \frac{h}{T-1} \sum_{t=2}^{T} (y_t - y_{t-1}) = y_T + h \left( \frac{y_T - y_1}{T-1} \right)$$
(3)

2.3.4. The linear model. Least-squared regression line would be used for prediction when there is a linear relationship between single predictor x and forecast variable y, and the equation is:

$$y_t = \beta_0 + \beta_1 x_t + \varepsilon_t \tag{4}$$

2.3.5. The Arima model. It is the combination of autoregression and moving average model [10]. In this research, most of the operations would be automatically performed in the R-Studio. Its full model could be written as:

$$y_t' = c + \phi_1 y_{t-1}' + \dots + \phi_p y_{t-p}' + \theta_1 \varepsilon_{t-1} + \dots + \theta_q \varepsilon_{t-q} + \varepsilon_t$$
 (5)

<sup>\*</sup>Year is the time for the annual violent crime rate data next to it

<sup>\*\*</sup> Rate is the number of people who commit violent crime per 100,000 people

### 3. Results and discussion

# 3.1. Testing of each model

3.1.1. Testing of the average method. According to figure 2, the residuals from the average method varied from -135 to 261, which is very large. Apart from this, the residuals show an evident decreasing trend, which means the model is not effective. According to table 2, by using the Ljung-Box test, the Q\*value is extremely high, which means the residuals have a very strong autocorrelation.

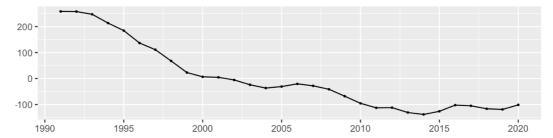
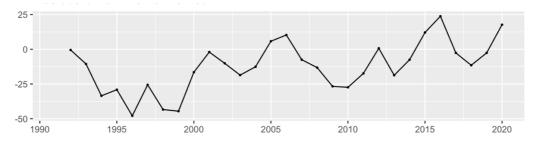


Figure 2. Residuals from the average method.

**Table 2.** Autocorrelation testing of the average method.

|                | Q*     | degree of freedom |
|----------------|--------|-------------------|
| Ljung-Box test | 83.262 | 6                 |

3.1.2. Testing of the naïve method. According to figure 3, the residuals are between -50 and 25, which is not too big. However, the residuals are not randomly distributed below or above the, which means the residuals do not look like white noise. In table 3, when the residuals are checked in the Ljung-Box test, there is a small degree of autocorrelation. When the ACF graph of the residuals is observed in figure 3, although most of the values are lower than the significance level, the value of the first lag is too big, which means there is strong autocorrelation in the first lag.



**Figure 3.** Residuals from the naïve method.

**Table 3.** Autocorrelation testing of the naïve method.

|                | Q*     | degree of freedom |
|----------------|--------|-------------------|
| Ljung-Box test | 17.168 | 6                 |

3.1.3. Testing of the drift method. According to figure 4, the residuals are between -50 and 25, which is not too big. However, a larger proportion of residuals are below 0, which means the residuals do not look like white noise. In addition, based on the Ljung-box test result in table 4, there is a small degree of autocorrelation, and the autocorrelation exists in the first lag.

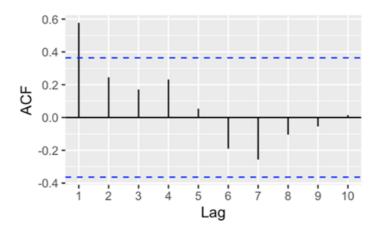
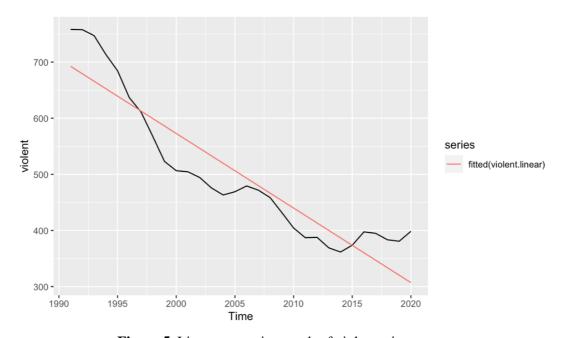


Figure 4. ACF of residuals from the drift method.

**Table 4.** Autocorrelation testing of the drift method.

|                | Q*     | degree of freedom |
|----------------|--------|-------------------|
| Ljung-Box test | 17.168 | 6                 |

3.1.4. Testing of the linear model. Firstly, this research would use R-Studio to find the least-squares regression line between the violent crime rate and the time. The regression line in figure 5 represents the trend of the data. The coefficient of the year corresponding to the violent crime rate is -13.288, which means on average the violent crime rate decreases by 13.288 as one year has passed. Since the P value is lower than 0, the trend is highly significant, which means there is a linear relationship between the year and the violent crime rate. After that, the residuals of the linear model of crime rate will be checked (table 5).

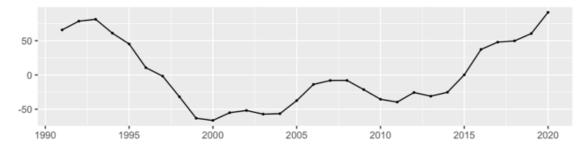


**Figure 5.** Linear regression result of violent crime rate.

**Table 5.** Linear regression result of violent crime rate.

|           | Estimate | Std.Error | Pr       |
|-----------|----------|-----------|----------|
| Intercept | 705.732  | 18.721    | <2e-16   |
| Trend     | -13.288  | 1.055     | 3.67e-13 |
| R-squared |          | 0.8501    |          |
| P value   |          | 4.67e-13  |          |

In figure 6, the residuals are between -68 and 83, which are relatively large. According to the Breusch-Godfrey test result in table 6, the residuals from the linear models of the violent crime rate are autocorrelated. In addition, the first three lags of the ACF graph are higher than the critical value, showing that within the first lag, autocorrelation is strong.

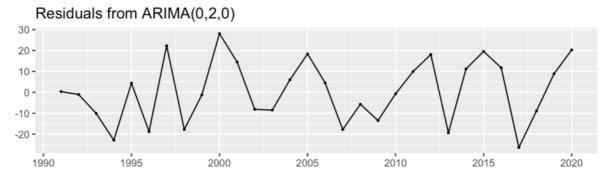


**Figure 6.** Residuals from the linear method.

**Table 6.** Autocorrelation testing of the linear model.

|                | Q*     | degree of freedom |
|----------------|--------|-------------------|
| Ljung-Box test | 26.338 | 6                 |

3.1.5. Testing of the Arima model. This research would use "auto.arima" in the tool package of R-Studio to find the Arima model with the least residuals, which is Arima (0, 2, 0). When we check the residuals in figure 7, they are nearly randomly distributed above or below 0 between -26 and 28. In this case, it can be shown that the residuals are relatively small and look like white noise. According to the Ljung-Box test in table 7, there is a very small degree of autocorrelation that can be omitted. Apart from this, in figure 8, all the ACF value is lower than the significance level, and the value of the first lag is extremely low (table 7).



**Figure 7.** Residuals from the Arima model.

**Table 7.** Autocorrelation testing of the linear model.

|                | Q*    | degree of freedom |
|----------------|-------|-------------------|
| Ljung-Box test | 7.539 | 6                 |

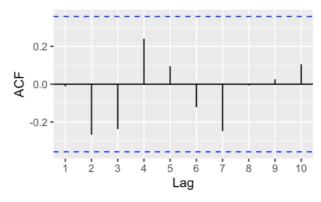


Figure 8. ACF of residuals from the Arima model.

# 3.2. Selection of models

Based on table 8, the Arima model is the most effective in predicting the American violent crime rate in the future. It has the smallest residuals, and the autocorrelation is negligible, which means it possesses the merits of accurate models. Therefore, this research would use Arima (0, 2, 0) to perform the prediction in the following session.

Range of residuals 297 (-136~261)

**Table 8.** Comparison of different models.

#### autocorrelation Average method very strong Naive method 75 (-50~25) middle Drift method 75 (-50~25) middle Linear model 151 (-68~83) relatively strong negligible Arima model 54 (-26~28)

# 3.3. Prediction results and discussion

Based on table 9, the violent crime rate will increase by 17.7 per 100,000 people, and the prediction is 216.2 in 2021, 233.9 in 2022, 251.6 in 2023, 469.3 in 2024, and 487 in 2025. Since this research is conducted in 2023, the most valuable part should be the prediction in 2024 and 2025.

In figure 9, Arima (0, 2, 0) gives a forecast of increasing violent crime rate from 2021 to 2025, which is different from the long-term trend. In this way, it could be explained that after a sharp decrease of violent crime rate, the rate of the violent crime rate decline decreased gradually but will eventually increase again. In this process, the derivative of the violent crime rate continues to increase so that the model shows a U-shape. Another possibility is that this model may not be fully accurate. Based on the figure, the violent crime rate remains relatively stable from 2011 to 2020, so it may continue to be unchanged rather than start to increase. However, Arima (0, 2, 0) considers the decreasing trend from 1991 to 2010, which will dramatically change the result.

Regardless of the possible errors, this research predicts that the violent crime rate will increase in the following years, which means the condition of violent crime in following years is not as optimistic as the condition in the past thirty years. Therefore, the government should pay more attention to the violent crime. For example, the government should increase its expenditure on social security or call on the community to prevent violence from happening, which could effectively undermine criminals.

 Year
 Predicted value

 2021
 416.2

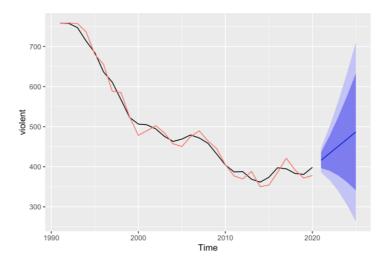
 2022
 433.9

 2023
 451.6

 2024
 469.3

 2025
 487

**Table 9.** Predicted value of violent crime rate.



**Figure 9.** The prediction of Arima(1,2,0).

# 4. Conclusion

Previous studies explored the historical trend of crime rate: the number of all types of crimes declined dramatically after 1990. In addition, some studies showed that the crime rate is related to certain factors such as unemployment rate, income inequality, immigrants, and influential events. However, this research study focuses on one type of crime, violent crime, predicting its future change based on time series analysis. Within the five candidate models, Arima (0, 2, 0) shows the highest accuracy. It has negligible autocorrelation and the smallest residuals, so it could be used in the prediction in this research.

The prediction of the violent crime rate shows an increasing trend in the following models, which contradicts the long-term trend after 1990. In this case, this research recommends the government to take violent crime seriously, taking action to prevent the happening of violent crime.

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