**Acknowledgements**

The goal of this study is to develop a real-time Python dashboard application for visualising, analysing, and forecasting global suicide rates. Python, PHP, and MySQL were primarily used to create this project from scratch. The main goals of this study are to forecast how many fatalities will occur in the next ten years as well as to better understand the underlying issues with past suicides. An initial analysis is done manually as usual practice on Jupyter notebook using two CSV files received from Kuggle.

Many studies on death analysis have been conducted in the past, but this research is particularly focused on providing a fully functional dashboard and making it accessible everybody for free of cost. The main goal of this project is to create an application that can dynamically anticipate upcoming deaths in various nations every year. To administer the data from the backend, a modern Admin Panel built with the Laravel PHP Framework.

Applications for the front and back-end are both available online for anyone with internet access to view.  Only the suicide dash app admins may add new records to database through the admin panel section. A feedback collection page is integrated with the Laravel application as part of the project so that site users can email their feedback via the contact us section. After receiving feedback, two emails will be sent: one with details to the admin email for the Dash app, and the other with thanks message to the user. Admins will also be able to see the feedback from admin panel and delete them if they need. CRUD (create, read, update, delete) operations are allowed for maintaining the suicide dataset from backend.

Python dash contains mainly three parts, visualisation, forecast and Data Integrity check. All the data will be visualized using plotly package. Using dropdown or range selector users can filter different information according to their wish. In the forecast section three models predict the future suicides in all the available countries. SARIMA, FBProphet and Custom AR models are used to predict the values. RMSE score along with line chart is also displayed on each model selection. Finally, the Data Integrity part is used for checking outliers. Periodic Outlier detection is used for finding outliers from the dataset.

As you know handling large amounts of data is hard in modelling. Since our model must make predictions from the new data added from the backend cron jobs are used from Linux centos SSH server from Mochahost. The modelling and data maker are the two python scripts run weekly which would dynamically clean the dataset and make predictions from the updated dataset loaded from Maria MySQL DB. Root VPS access has been obtained from the host service provider to configure pip packages preparing the server to be ready for deploying python dash application.

Eventually, the whole research ended up being a success. The most difficult part of this project is to deploy the application to SSH terminal. My experience and expertise on data architecture helped me to figure out the issues with the server which I have purchased from Mochahost.

**Declaration**

“I hereby declare that the work described in this project is, except where otherwise stated,

entirely my own work and has not been submitted as part of any degree at this or any other

Institute/University”

Signature : \_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_

Name : \_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_

Date : \_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_

**Dedication**

**“Thank God!”**

To my family,

Without their love, support and encouragement I would not be here.

To my teachers,

I am thankful to Dr. Jack and Dr. Abhishek from DKIT for motivating me throughout my journey of this thesis paper. Thereby, I dedicate this to them”

To my friends,

My friends around the world for all their love, support and criticism for my personal and professional well-being.

**Abstract**

This paper explains the process of data analysis on suicide data that have occurred in various countries around the globe. Data was chosen from the Kuggle website. The dataset on suicides includes a variety of suicide-related statistics. No personal information was used in the research process and all data used for this project was fully impersonal. The main goal of this project is to create a versatile Python Dash app that can generate real-time predictions and visualizations from data provided from the backend. Initially, Jupyter Notebook was used to clean data that was obtained from the Kuggle website. The dataset was cleaned in Jupyter notebook manually, using imputation technique missing data has been handled, also a machine learning technique called periodic outlier detection (POD) is used for treating outliers. An unsupervised Machine learning Algorithm called DBSCAN (Density-Based Spatial Clustering of Applications with Noise) is used for POD process. Exported data has been stored into the GitHub repository for further action. Data storage is done to a Relational Database Management System (RDBMS) called Mysql. XAMPP has been used to configure it in Windows local machine.

Dash app prediction used mainly three models. Sarimax model showed best results from the analysis. Followed by FB Prophet model also gave good results. Finally custom Autoregression has also seems to show good results with more RMSE score. My models were making live predictions each week. Due to the server capacity and size of the data I was unable to run the model every day. Mochahost centos terminal connected with SSH via putty to prepare the server. CRON jobs are configured on the server using terminal as well inorder for making the model automatically every week at 10:00 AM server time. Code used for the project is available at following github link: <https://github.com/sujilkumarkm/suicide_dash_app_2022.git>. Dashboard was deployed to the following 2 links: <https://suicide-dash-app.herokuapp.com/> and <http://204.93.172.126:8000/>. Admin Panel login section is used to do CRUD(create, read, update and delete) operations. The link for admin panel is given here: <https://www.dkit.ie.narayam.net/admin/login>, here page administrator can login with their email and password to manage the whole website which also contain a feedback section.

Link to feedback section is <https://www.dkit.ie.narayam.net/contact> where page visitors can drop their feedback which will be stored into the database, and it will be available for the admins of the page to manage from the backend. This app will also send email to user and admin separate. Mailtrap configuration is used for setting up email service. A .env file is configured on Laravel app with all required environmental variables to facilitate the app to properly work on the live server. Throughout the app maintenance, no coding has been done on the production server, instead all the coding were done on the repository which was eventually pulled to the production. Thus, this project makes full use of git version control technology. Dash app also display the data integrity which is also updated every week to make sure page visitors can see the outliers. A final page with few important variable’s information has been shared as tables where users could search for the specific record using filtering option.

**Abbreviations**

**WHO** World Health Organization

**POD** – Periodic Outlier Detection

**ML** – Machine Learning

**RDBMS** - Relational Database Management System

**DBSCAN** - Density-Based Spatial Clustering of Applications with Noise

**RMSE –** Root Mean Squared Error

**CRUD** – Create, Read, Update, Delete

**Contents**

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| Acknowledgements  Declaration  Dedication  Abstract  Abbreviations  Contents  List of Tables  List of Figures  1 Introduction  2 Literature Review  3 Exploration of data  4 Design and Implementation  5 Evaluation and Testing  6. Results  7. Conclusion  8. Deployment  Appendix A  Appendix B  Bibliography |  |

**List of Tables**

**List of Figures**

**Chapter1**

**Introduction**

This iteration is a deep dive into the suicide dataset to learn much more about the reasons for the thousands of suicides that occur each year around the world. Even though various studies on suicide have already been done previously, such as John et al. (2018), this study aimed to produce new insights that can help government bodies better grasp the problems that lie beneath them. This research could also benefit them in developing new strategies to minimize mortality rates over time. This research will look at a variety of suicide attributes and predict how many more fatalities will occur in various countries in the next years.

The goal of this research is to figure out why people commit suicide in each country. Every year, 800,000 individuals commit suicide, according to Wikipedia (2012). Suicide, for example, is becoming a more prevalent and serious problem in India, according to the World Health Organization (WHO). To address these issues, we must examine various patterns and clusters in the data and determine what circumstances cause someone to consider suicide. In addition, a web-based system will be developed that may offer dynamically illuminating visualizations of the suicide dataset, as well as opportunities for page administrators to submit new suicides to the dataset. This initiative will have a huge influence on society by allowing the government to identify and assist those who are in need, hence reducing the number of suicides each year in each nation. The government will not only save lives but also make the globe a better and safer place for people to live by implementing suitable steps based on the findings of this study.

Different social, economic, and cultural contexts exist in different countries. Russia and Ukraine, for example, are two of the most mentioned countries recently. The world is aware that the two countries are involved in a major dispute. When you see that kind of observation and data insights in Explanatory data analysis (EDA), it's always suspicious (there is a presumption that there is a relationship between the conflict and suicides in two countries); the two countries, among others, have high suicide rates.

|  |  |
| --- | --- |
| **Research questions** | **Project Goals** |
| 1. How to create a multipurpose web app for prediction. 2. How to handle mass data on the database server? 3. How to make dB schema? 4. Get feedbacks from page visitors. 5. How suicide rates change in different countries. 6. How to re-run the models without affecting the speed of server or output? 7. How much data is having outliers dynamically? 8. Check for any trend in suicides in different countries over the years. 9. Check reasons why suicides in each country changed over the years? | 1. Check relation between GDPs Per Capita and suicide rate 2. Which country is affected by the highest number of suicides with respect to population? 3. Which age groups is more likely to suicide? 4. Predict number of suicides going to happen in each continent in next 5 years 5. Predict top 5 countries with least number suicides in coming 5 years. 6. Find out the age group of people who are more likely to suicide? 7. Check the performance of each model by looking the error and accuracy. 8. Make the app work according to the update from back-end. |

In 2011, 554 people in Ireland committed suicide, according to the CSO statistical release (2011). In terms of the country's population, this is a large figure. Each suicide will have its own set of motives. Have you given any thought to the various reasons of these figures? You won't know the answers to these questions unless you start studying and researching suicides, like Zetzsche et al's (2007) research when they sought to figure out why people commit suicide in Western and Central Europe and came up with a few extremely interesting findings. moreover, all the suicides might be due to several factors that we are not aware of. To put it differently, collecting all that information is challenging, but there are some elements that make individuals think of committing suicide in every country. Some of these common factors are included in the suicide dataset as features, which can be used to dig deeper into the data and compare the trend in data from multiple countries.

Database design is the first step in data upload to the database. This is done using Laravel one of the most powerful and advanced PHP Framework is used. Database schema is designed using Artisan Eloquent Models. DB Table structure is defined in migration files in the Laravel ‘database’ directory.

|  |
| --- |
| **Core technologies**  • Programming Language:  – Python,  – PHP  – IDE: Visual Studio Code and Jupyter Notebook  – Libraries:  \* pandas  \* numpy  \* requests  \* sklearn  \* MySQL  \* dash\_bootstrap\_components  \* dash  \* matplotlib and plotly  \* seaborn  \* sklearn  \* prophet  \* statsmodels   * Frame Works   \* Bootstrap CSS,  \* PHP Laravel,  \* Python Dash   * Server   \* Heroku  \* Mochahost |

Python is the most used language for data analysis, according to studies from Data Camp and Bootcamp, and it is the most widely used programming language globally. Python is one of the most potent and popular programming languages in the world, so it should come as no surprise that it continues to dominate the data analysis sector. So, Python has been chosen for this study project on suicide analysis. It had taken about three months to finish this project.

• **Data Collection**: Data has been collected from the open source Kuggle Platform. Data firstly manually downloaded to the local machine then uploaded to the Jupyter notebook for further processing. A second dataset has been added to the master file which is received from a public machine learning repository. Also, in order to combine continents another csv was collected from internet where each country name is having continent names to classify which continent that country belong to.

• **Data Preparation:** In this stage, Data is prepared by correcting datatypes and each row will be checked for missing or Null values and looked for outliers. These missing values and outliers are carefully treated without loss of information and keeping the accuracy and balance of the features. An output CSV is finally generated after normalizing and cleaning and that will be loaded to the MySQL database.

• **Data Exploration:** After careful preparation of data, meaningful visualisations were made using plotly, seaborn and matplotlib packages. During visualisation stage sub dataset like russia.csv was made to look deeper and understand socio-economic backgrounds and suicides relationships in few countries separately.

• **Labeling**: In order to make visualisations about the risk and non-risk groups, a new dataset with risk and non-risk columns were required. A machine learning algorithm called as DT-classifier was used to classify the suicides into risk and non-risk groups. Since final dataset contained more than hundred thousand records, doing more research about the classifier modelling was necessary to understand how this classification works.

• **Modeling**: Since the study conducted was based on sequential data, time series models were used to make predictions. SARIMA, Custom Auto Regression and FB Prophet models were created and compared before making the final app. Another machine learning algorithm was also included with modelling to make data integrity check using POD.

• **Evaluation:** Evaluation of the models is done by examining the RMSE scores. Evaluation of model is done initially on the Jupyter notebook for initial analysis as well as dynamically on the suicide dash app. These scores are used to compute goodness of the fit. It is found by taking the correlation coefficient between true values and predicted values.

• **Deployment**: The most important step in publishing a web-based application is deployment. The python dash app needs to publish on live servers like Heroku or any other popular Host service provider. So that any user could visit the app anytime. It was planned to show line plots on the number of suicides per year. MySQL database from the same service provide has been loaded and maintained with suicide dataset.

Diagram

Description automatically generated

***Fig x.x.x*** *Lifecycle of the project*

The report's framework begins with Chapter 2, which reviews the literature on the subjects most closely connected to the research questions listed in Table 1.1. Chapter 3 (Exploration of the Data) then details all procedures used to gather and clean the suicide dataset to produce the final dataset for analysis. Labeling is explained in Chapter 4 (Design and Implementation). The based model's construction and design approach for testing and training with worldwide suicides. Chapter In Chapter 5, "Parameter Evaluation, and Testing," the procedure for evaluating the based model. The number of deaths happened in Russia over time is covered in Chapter 6 (Results) in general. Chapter 7 (Conclusion) explores the question of Future work, and the research questions were addressed. Chapter 8 concludes (Deployment)

**Chapter 2**

**Literature Review**

**2.1 Related work**

Several apps have been made like suicide analysis dashboard, Poverty and Equity Dashboard (2020) was one of the main such inspirations for creating this application. Even though there were several such works, there were no app that is dynamically updated based on the new records added from the admin panel end. Using Migration and Population Density data from World Bank. Dashboardom (2017) has created a dashboard. Their dashboard was designed in a simple way without having lots of CSS styles, but easily understandable for any non-technical person. They found that countries like Bahrain and Maldives have high migrant rate compared to other countries from the available world bank data. Another simple visualization project made by real python for avocados sales in US dashboard (2018) shows the simplest way from coding to deploying of python dash applications.

Another study seen in time series forecast was about the Accidental Deaths in India: Patowary et al. (2018) used ARIMA Model and they used dataset of natural and unnatural accidents in India between 1967 to 2015. Their study shown that ARIMA (2,2,1) model is suitable for the prediction of that dataset. In the study of Kumar Jha and Pande (2021) found that Facebook Prophet model outperformed other models in terms of accuracy. They have used Addictive Model and ARIMA model along with Prophet model for the forecasting. The Prophet model have shown them better fit, less error, better prediction compared to the other two. Auto Regressive AR is issued in forecast of wind speed by Huang and Chalabi (1995) was a good example of how good AR Models on time series data. In this study the time varying parameters of AR model were modelled by smoothed, integrated random walk process. The whole data was utilized for the visualizations in this project. I solely utilized data from the "Russian Federation" for the time series modeling and forecasting phase. Working with time series forecasting is an important part of this dissertation. I have several different targets in this dissertation including dashboard visualization, forecast modelling, database management etc. I have been looking for ways to predict the number of suicides in upcoming years. This interest in time series and ML made me dive deep into sophisticated time series models like SARIMA and VAR to make models on the suicide data and forecast future suicides in different countries. The ARIMA model is a combination of multiple models, including the Autoregressive model, the Moving average model, and the Autoregressive Moving Average model. The form of the ARIMA model is represented by ARIMA (p, d, q), where p is the autoregressive order, d is the number of differences, and q is the moving average order.

Vector Auto Regressive Model is mostly used in finance and econometrics because they offer a framework for achieving important modelling goals, including data description, Forecasting, Structural Inference, and Policy Analysis. VAR Model is a workhouse time series multivariate model that relates current observations of a variable with past observations of itself and past observations of other variables in the system.

Thirdly, we need a database server for data to be stored on the server. I will be using PSQL or ThisSQL servers for data storage and management. I want the data in this DB to be updated from time to time and this model must be updated based on the new data injected in each time. The reason for choosing these DB’s is the flexibility of usage and its syntax matching with Structured Query Language (SQL) minute differences.

**2.2 Suicide Analysis**

“Data is a precious thing and will last longer than the systems themselves.”, this is a quote from Mr. Tim Berners-Lee, founder of the World Wide Web. Data is the new energy source to fuel the tech industry, despite the problems that data presents, it appears that the right use of data is bringing a lot of useful application for society. Data scientists were able to develop successful applications that could potentially aid society in understanding underlying issues that we have yet to discover with the use of effective machine learning algorithms.

What is Time Series Analysis?

A particular method of examining a set of data points gathered over a period is called as "time series analysis" (Tableau). Organizations can effectively comprehend systemic patterns across time by using time series analysis. Business users can study seasonal trends and learn more about its causes using data visualizations.

There are three types of forecasts are there(Chatfield, 2000).

(a) Judgmental forecasts: using subjective judgment, intuition, "inside" commercial knowledge, and any additional pertinent data.

(b) Univariate forecasts: which may be supplemented by a function of time like a linear trend, only consider the current and historical values of the single series being projected.

(c) Multivariate forecasts: in which predictions of a given variable are at least somewhat influenced by the values of one or more additional time series variables, sometimes known as predictor or explanatory variables. If the variables are mutually dependent, multivariate forecasts may rely on a multivariate model with several equations.

There are a wide variety of problems that time series algorithms may resolve. Companies with large sales teams employ time series forecasting to help them make better business decisions. For example, a few of these concrete examples of these potential uses are

* estimating a stock's daily closing price.
* estimating a store's daily unit sales of a product.
* predicting a state's unemployment rate each quarter.
* estimating the daily average price of gasoline.

**2.3 Limitations on Suicide Analysis**

Each suicide that occurs in the world has a different set of causes since there are thousands, if not millions, of reasons why people commit suicide. A dataset that has already been aggregated won't produce empirical findings. Personal data must also be gathered and examined in order to draw more exact conclusions about the causes of those fatalities.

1. Country Names Missing: The analysis would have been more insightful if there had been data available from every country in the world as there are many nations missing from the dataset.
2. Server Requirements: Handling such large dataset requires more server capacity run the model time to time. Current server is only capable of running models with a smaller number of records, so the project is made in such a way that I can be run only once in every week.

**Time Series Forecasting with ARIMA, SARIMA and SARIMAX**

Time series forecasting is a challenging issue with no simple solution. The optimal statistical model is never quite evident because there are so many models that each claim to perform better than the others. Because of this, ARMA-based models (Brendan Artley) are frequently a suitable place to start. They are suitable as a baseline model in any time series problem and can produce respectable results on most time-series problems. The abbreviation for the ARIMA model is "Auto-Regressive Integrated Moving Average," and for the purposes of easy understanding, we shall separate it into AR, I, and MA.

AR stands for autoregressive component (p). The number of lagged series we employ is determined by the p parameter, which represents the autoregressive part of the ARIMA model as AR(p).

Diagram, text, schematic

Description automatically generated

Formula of AR

AR(0) - White Noise

If the p parameter is set to zero (AR(0)), there are no autoregressive terms. White noise is all that this time series is. Each data point is taken as a sample from a distribution with mean 0, variance 2, and standard deviation 0. This generates an unpredictable series of random numbers. This is quite helpful as a null hypothesis since it prevents our analysis from accepting false-positive patterns.

Oscillations and Random Walks in AR (1)

With the p parameter set to 1, we are multiplying the prior timestamp by a factor, then including white noise. A random walk results from a multiplier of 1, while white noise is produced by a multiplier of 0.

The time series will display mean reversion if the multiplier is in the range of 0 < α₁ < 1. This indicates that after regressing from the mean, the values tend to float around 0 and return to it.

Higher-order terms, AR(p)

The only way to increase the p parameter further is to add more timestamps that have been multiplied on their own. Although we can go back as long as we wish, it is increasingly likely that we should employ more factors, like the moving average (MA(q)), as we go further back.

ARMA and ARIMA Models

The AR (Autoregressive) and MA (Moving Average) components alone make up the ARMA and ARIMA architectures.

ARMA: The ARMA model consists of a constant plus the sum of the AR and MA lags and their multipliers, plus white noise. This equation serves as the foundation for all subsequent models and as a framework for numerous forecasting models in various fields.

Text

Description automatically generated with medium confidence

The ARIMA model is an ARMA model, however the model we describe using I includes a preprocessing step (d). The difference order, I(d), indicates how many transformations are required to make the data steady. An ARMA model on the differenced time series is what an ARIMA model is thus.

Models SARIMA, ARIMAX, and SARIMAX: Although the ARIMA model is excellent, adding seasonality and exogenous variables can have a significant impact. We must employ a different model because the ARIMA model presumes that the time series is stationary.

Diagram

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Except for the additional set of autoregressive and moving average components, this model and the ARIMA model are quite similar. The frequency of seasonality (ex. 12 monthly, 24 hourly) cancels out the extra lags. SARIMA models allow for both seasonal frequency and non-seasonal frequency differences in data. Automatic parameter search frameworks like pmdarima (Documentation) can make it simpler to determine which parameters are ideal.

A picture containing diagram

Description automatically generated

The SARIMAX model is seen in the fig x.x.x above. Exogenous variables are used in this model, or in other words, external data is used in our forecast. Examples of exogenous variables in the real world include the price of gold, oil, the outside temperature, and the exchange rate. It's interesting to consider that all external variables are still de facto implicitly modeled in the forecast from the historical model. However, if we incorporate external data, the model will react to its impact much more quickly than if we rely just on the influence of lagging components.

**FB Prophet**

An open-source library called Facebook Prophet forecasts time series data. It assists both individuals and companies in analyzing market values and forecasting the future. It puts into practice a method for predicting time series data that is based on an additive model where non-linear trends are fit with yearly, monthly, and daily seasonality, as well as holiday impacts. It functions best with historical data from multiple seasons and time series with seasonal impacts. It decomposes time series data into following components. Prophet is a method for predicting time series data that uses an additive model to fit non-linear trends with seasonality that occurs annually, monthly, daily, and on weekends as well as during holidays. Strongly seasonal time series and multiple seasons of historical data are ideal for it. Prophet typically manages outliers well and is robust to missing data and changes in the trend.

**Trend**: A pattern can be seen in the data. The time series data's non-periodic variations are modeled. The dataset's long-term movement is depicted by a trend. A trend may be constant, uphill (uptrend), or downhill (downtrend) (horizontal). Trends typically emerge for a while before disappearing.



#### **Seasonality:** This happens due to the daily, weekly, and yearly changes in the data.

A picture containing chart

Description automatically generated

#### **Holiday effect:** In a time-series dataset, these are the recurring days and events. It is concerned with the recurrence of well-known holidays like Christmas and others.

### **Benefits of using Facebook Prophet**

The advantages of use Facebook Prophet for time series modeling are as follows.

Prophet is utilized in numerous Facebook applications to generate accurate forecasts for planning and goal-setting. In most instances, it is discovered that it outperforms every other strategy. So that you can receive forecasts in only a few seconds, using the fitted models in Stan (Documentation).

Get an accurate forecast from muddled data without any manual work. Prophet can withstand outliers, missing data, and significant time series changes. Additionally, data points that differ from the main dataset observations are removed. It can manage the impacts of seasonality and holidays. It manages the spikes in the dataset and takes them into account while training the model.

Users of the Prophet method have a lot of options for modifying and adjusting forecasts. By incorporating your subject knowledge, you can employ human-interpretable features to enhance your forecast.

Both R and Python have a prophet technique, but they both use the identical Stan fitting code. To obtain forecasts, speak in a language that you are familiar with.

**Custom AR**

Custom AR is used because it is more flexible in terms of customization.

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Where Xi is value in ith previous year and Wi is weight or coefficient of regression. Custom AR is a manually written and it is the foundation on which other models are created. Custom Auto regression model is created because this study needed a transparent working mechanism to see how these forecasting algos work. Linear Regression Algorithm has been used in fitting the model.

**Chapter 3**

**Exploration of Data**

**3.1 Data Collection**

No data scrapping was involved in this research. Dataset has been directly downloaded from the Kuggle website. As you can see from the fig x.x.x data has been directly downloaded and stored into an encrypted HP laptop hard drive. Data has been loaded to Jupyter Notebook inside anaconda environment for initial analysis(EDA). The main objective of this research is to bring underlying issues with suicides happening around the world, neither data scrapping was required nor important to the main objectives of the study. The best part of using Kuggle Dataset was that it did not require any extra effort to write scripts on how to bring the dataset from APIs like twitter API or YouTube API. On the other hand, this could affect negatively by not getting into these results less experience with data scraping. Moreover, this research is more of visualisation and prediction oriented where web scrapping or API data collection are less important.

MySQL

Local Machine

CSV from Kuggle

**3.2 Data Preparation**

Data imported to the Jupyter notebook file has been analyzed using EDA and statistical analysis techniques. The first stage is to check for correct data types. Data in the form of data frame will be analyzed with panda’s package. Describe function gives an overview to the data. After fixing the outliers, duplicate rows have been checked and treated with care. If the entire row having most of the information repeated from the previous row, then that have been removed. Outliers can be checked using different methods such as box

plots or statistical methods. I have used some statistical approach to pull out the extreme values. DB scan has been applied to handle the outliers.

Decompose the Data (2020): In case of a broad upward trend without any discernible seasonal or cyclical patterns. The data must then be broken down to see more of the complexity that lies behind the linear visualization. We may divide the data into four parts with the help of the seasonal decompose Python function from the statsmodels package.

Any time series data should be examined for trends that could have an impact on the outcomes and could guide the selection of the forecasting model. Several typical time series data patterns include:

1. Level: The average value in the series
2. Trend: Increases, decreases, or stays the same over time
3. Seasonal or Periodic Pattern: Pattern repeats periodically over time
4. Cyclical Pattern: Pattern that increases and decreases but usually related to non-seasonal activity, like business cycles
5. Random or Irregular Variations: Increases and decreases that don’t have any apparent pattern

**Python Dashboard in Plotly**

The dashboard about suicide was inspired by Johns Hopkins COVID-19 Dashboard (Johns Hopkins). This dashboard displays real-time Covid-19 data that is regularly updated, with corresponding graphical representations created automatically. The python dashboard, which visualizes the data and insights for the public, will be the most appealing aspect of this project. Python dash and streamlit have emerged as the most capable frameworks for web-based visualisation projects in recent years. This project provides both static and dynamic visualisations. Before the real web dashboard app, individual static graphs are produced to obtain insight from the data. A final dashboard app with dynamic visualisations will be constructed when the initial static models are completed in Jupyter Notebook., I want to talk about python dashboard. It’s always wonderful to see how we can make models and interpret them. But it is also important to note, recently there are number of concerns about how well we can make modifications to the existing model and maintain them. So, our model must work dynamically and make prediction based on the available data. In recent years programmers used use VueJS or web-based languages for making dashboards, we now have most advanced packaged like has made these process easier and more efficient. I am going to use some of the python packages like plotly to make interactive dashboard and make models that can make great predictions.

**Suicide per hundred thousand around the world**



***Fig. 4.3:*** *Suicide per hundred thousand around the world - Timeline in plotly*

Plotly is used to depict the global suicide rate in fig 4.3. Visitors can see information based on the year on an animation frame page. The rate of suicides per 100,000 is shown by colored zones.

**Suicide per hundred thousand in Male and Female**

Chart, bar chart

Description automatically generated

***Fig. 4.4:*** *Suicide per hundred k among males and females in different age groups in different countries*

As per fig 4.4, Most suicides are happening between the age of 35 and 54. And out of the majority are Males. In all the age groups females are less affected groups. Also, we can see from the age of five to fourteen children are less likely to commit suicide.

**Suicide per hundred thousand Vs GDP Per capita**

A screenshot of a computer

Description automatically generated

***Fig. 4.5:*** *Suicide per hundred thousand around the world in different countries*

As seen in the fig 4.5, nations such as Russia, Ukraine, and Hungary have some form of relationship in terms of the number of suicides per 100,000 people. According to a BBC News article, the causes of suicide in Ukraine are the consequences of Russia-Ukraine hostilities and the ongoing war. Ukraine is undoubtedly one of the most afflicted countries in terms of suicide, according to our statistics.

**Top ten countries with the highest suicide averages**

Chart

Description automatically generated

***Fig. 4.6:*** *Top ten countries with the highest suicide rates*

The extent of suicides in various nations is seen in Figure 4.6. It is apparent that the Russian Federation has the highest suicide rate of all the countries studied. According to earlier research by Bellman and Namdev (2022), Russia has a considerable problem with suicidal behavior among men, and their drinking habits have a substantial impact on their decision to commit suicide when compared to other nations.

**Population, Suicide, Suicide in Hundred Thousand Vs Gender and Age**

Graphical user interface, chart, histogram

Description automatically generated

***Fig. 4.7:*** *Suicide, Suicide per hundred thousand, Population in different genders and age in all countries*

As seen in fig. 4.7, Females outnumber men in terms of population. In relation to total suicides among men and women, the majority occurred between the ages of 25 and 34. It is undeniable that the male population has a higher suicide rate per 100,000. Furthermore, the majority of those who died were above the age of 75.

**Population, Suicide, Suicide in Hundred Thousand Vs Gender and Age**

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***Fig. 4.8:*** *Total suicides in the world in each year distribution*

Fig 4.8 shows that there is a quick rise in the suicide rate from the year 1990. After that, it continued increasing until the next ten years. From 2000 it started to decline.

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**Fig. 4.9:** Global Distribution of different features

According to the fig 4.9 above, the number of suicides per 100,000 climbed from 1985 to 1995, then rapidly fell. At the same time, global GDP per capita climbed gradually from 1985 to 1995, remained stable until 2003, then increased abruptly until 2014, before falling precipitously in 2015. (World Economic Situation UN). Life expectancy and unemployment have been influenced by missing data and imputation, making the lines appear irrelevant.

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***Fig. 4.10:*** *correlation matrix of suicide dataset*

Normally, checking for features in the correlation matrix is necessary to see whether there are any features with strong correlation. We usually eliminate such variables from the dataset since they might cause the model to overfit. As seen in fig. 4.10, there is a significant correlation between suicide and suicide per hundred thousand. As a result, before modeling, the suicides feature was deleted from the Data frame.

**Chapter 4**

**Design and Implementation**

**4.1 Database design and Migration**

For running python packages and preparing the server ready new VPS package has been purchased from Mochahost. pip dash packages have been installed on the server. Initially small bits of code have been written on the local machine inside Jupyter notebook. XAMPP has been also installed on the machine to facilitate Maria DB. Using command prompt csv files have been uploaded to the dkit database. Using SQL ‘LOAD DATA LOCAL INFILE’ commands command has been run in order to get he data which is produced by the data\_maker.py file inside the suicide\_dash\_app folder.

A screenshot of a computer

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***Fig x.x.x*** *Cpanel Home Page*

**4.2 SSH Access and Putty Configuration**

In order to use the SSH access time to time, putty configuration was required. Data architecture skills need to be used to configure or prepare the server. cPanel’s official documentation (2021) can be checked to get details on preparation of the server.

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***Fig x.x.x*** *SSH access to the server and files in the dash app*

**4.3 DB Schema and table creation**

After creating the database, database design is done from Laravel artisan console. The required columns and table names were created as migration files in the Laravel database directory. Appropriate datatype has to be defined for each specific variable. Tables required for both python dash app as well as Laravel backend for admin panel has been defined in one go. "php artisan migrate” command has been used to run the migration. Database section also has database seeders used to inject predefined data like admin names and password for super admin access. “php artisan db:seed has been used to run the seed.



**4.5 Contribution of PHP Framework**

In earlier days, people used tell that PHP is slow and unsafe framework, as new versions of PHP have evolved, new frameworks like Code Igniter, Laravel were came into action with more sophisticated technical capabilities keeping the application faster, smarter, efficient and easy to use. Laravel uses MVC Architecture(Model, View, and Controller). It is the most useful any developer can easily rely on in this era (2022). Laravel uses power DB management system called eloquent ORM(Object relational Mapper). Also, Artisan console provides easy access to it’s boilerplate codes by just typing few commands.

**4.6 Modeling and Forecasting**

This project has various objectives, including creating a dashboard for visualizations, forecasting using Machine Learning Models, and creating an Admin Panel Portal for updating new suicides etc. Working with a time series model and projecting future values would be the most intriguing and challenging aspect of this endeavor. A decision tree classifier was employed to categorize the risk and non-risk groups while dealing with numerical and categorical information. Brunello et al . seem to have effectively utilized them in their research, which prompted me to extend the concept to suicide analysis. Checking seasonality, trends, and stationarity, as well as assessing prediction tests for identifying the optimal model for prediction, such as AIC and BIC, are all part of this research.

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***Fig. 4.11:*** *ARIMA Forecast and observed value graph on Russian data*

AIC and BIC graphs were made as shown in fig 4.11 for checking the order of the ARIMA model. In this research I am trying to work on predictions so, I will be looking at the AIC. A lower AIC score means a better predicting model. If the order is set too high, it could result in a high AIC value, this stops us from overfitting the training data. BIC is similar to AIC; lower BIC indicates a better model. BIC likes to choose a simple model with the lower order. AIC is better at predictive models, but BIC is choosing a good explanatory model.

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***Fig. 4.12:*** *AIC and BIC Score in ascending order on Russian data*

Here Fig 4.12, I am looking at a better predicting model so, I will be choosing AIC with the least score. This is an ARMA(2,0) Model

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***Fig. 4.13:*** *ACF and PACF to choose the model in ARIMA on Russian data*

From fig 4.13, In the above ACF and PACF, we can see ACF tails off and PACF cuts off since we have a MA(q) model. So this is an AR(1) Model.

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***Fig. 4.13:*** *Sarimax Model Results with order (1,0,0) on Russian data*

**4.4.1 grid search ARIMA parameters for time series**

We can automate the process of training and evaluating the ARIMA Models on different combinations of model hyperparameters. In machine learning, this is called grid search.

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***Fig. 4.14:*** *grid search result from ARIMA Model on Russian data*

In fig 4.14, we can see I have implemented the grid search and evaluated the ARIMA Model. Also, I have evaluated a set of different parameters.

**4.4.2 Prediction using Vector Auto Regression Models (VAR Model)**

Another Model used for the time series data is the VAR model (Vector Auto Regression). The reason behind using this model is that it helps in forecasting models based on multiple variables in time series. Usually, we use single variable and sequential time for time series analysis. But here I was able to include multiple variables in the model as you can see in the figure. Vector Autoregressive Models are one of the best models we could use to choose for time series.

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***Fig. 4.15:*** *predicting future values in VAR Model on Russian data*

You can see in fig 4.15, that we have predicted the number of suicides for the year 2016 using VAR Model on the time series sequential data.

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***Fig. 4.16:*** *Regression result summary from VAR Models on Russian data*

The coefficient scores for multiple variables in the time series analysis are shown in Fig 4.16. Each variable's standard error is also included, and coefficient values reflect the correlation of variables utilized to create the model's statistical equation.

I was able to create predictions on multi-variate time series data using VAR Models, as I described previously. The actual value of and the anticipated value distribution in the VAR Model are shown in Figure 4.17.

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***Fig. 4.17*** *Plot of Forecast vs Actuals from VAR Models on Russian data*

From fig 4.17 you can see how the model forecasted different variables in the Russian suicide dataset concerning actual values.

**4.4.3 Prediction using Auto Regression Models (AR Model)**

The next model I have created is the Auto regression model, Train and Test were split into seventy and 30 per cent. Seventy per cent of the data was used for training the model and the rest thirty per cent was used for testing. I have got an 11.792 Root mean squared error. Also, I could save different models to the local and I was able to load the models later and update them accordingly.

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***Fig. 4.18*** *Plot of Forecast vs Actuals from VAR Models on Russian data*

As you can see in the diagram above, the suicides per hundred thousand are distributed throughout the year in ‘Russian Federation’ is shown.

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***Fig. 4.19*** *Plot of Forecast vs Actuals from AR Models on Russian data*

In this fig 4.18, the blue line is the test data and the red line is the predicted values. I have AR 1 Model with one window.

**4.4.4 Decision Tree Classifier**

I have added a new column called risk where I split the data into two classes, class 1 stands for high risk and class 0 for low risk. Using the decision tree model, I have made a classification.

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***Fig. 4.20*** *Plot of Forecast vs Actuals from AR Models on Russian data*

Fig 4.20 shows the result from the Decision Tree classification algorithm. Lee and Oh (1996) have done studies on Neural networks using a Decision Tree classifier to distinguish between complex features. Here this goals are to separate the risk group and non-risk groups based on the feature called risk. I got 99.62 training accuracy and a hundred per cent testing accuracy.

**Chapter 5**

**Evaluation and Testing**

Evaluation of the model is as important as making the model. I have created 3 models in ARMA, Auto regression and Vector Auto regression. Using mean squared error and R-squared error I took the error rates of different models. Accuracy is also calculated to understand how efficient and precise this model is. Despite the fact that the model below was constructed using a Russian dataset. The live dashboard and modeling for the final product will be based on the whole country's master dataset.

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***Fig. 4.21*** *ARIMA Diagnostic plots on Russian data*

After finalizing this model, I would be able to add more evaluation techniques. Above are the four diagnostic plots I have created after running ARIMA Model. Looking at the plots I can say there is no pattern in the standardized model. Looking at the Histogram there is no Gaussian distribution (green and red lines should be almost the same for gaussian). The QQ Plot seems to be not normally distributed, if it is normally distributed all the blue dots will be aligned over the line (except for some values in either end).

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***Fig. 4.22*** *Plot of Forecast vs Actuals from VAR Models on Russian data*

As you can see in fig 4.22, we have calculated the accuracy of each variable. This helps us understand how well our model is performing.

**4.5 Data Storage**

Thirdly, we need a database server for data to be stored on the server. I will be using PSQL or ThisSQL servers for data storage and management. I want the data in this DB to be updated from time to time and this model must be updated based on the new data injected in each time. The reason for choosing these DB’s is the flexibility of usage and its syntax matching with Structured Query Language (SQL) minute differences. Initially, data was stored in the CSV format in different files. Later I uploaded them into Mochahost Psql server.

**4.6 Web Server and Hosting**

We know there are thousands of hosting companies providing hosting services, I have chosen Mochahost as one of the best service providers for small business websites. This goal is to make a highly dynamic web application on the server. I have purchased a VPS service which allows running PIP Packages on their server making the server IDE more suitable for the Dash App. Mochahost cPanel will be connected to the Github repository where this application is updated from time to time. Using git technology for the hosting makes the process more sophisticated and professional in terms of version control.

**4.7 Data Security**

Data Security has become an important concern in this era. Even though the suicide dataset is publicly available on the internet, I have followed the best practices in data security to ensure there is no data leakage. I have used encrypted windows drive to store the data. Whole project codes are updated from time to time to GitHub private repository. Any information related to this study has been considered for data security and ethical practices before actually using them. No personal information is used in this study. For web applications, files are kept in a private repository and used that repo to pull changes to the live mocha host server.

**4.8 Applications and software**

The most popular IDE for coding is Microsoft Visual Studio. I've been using Github for version control. In addition to Jupyter Notebook, Spyder, and Atom, I've used various coding tools such as Jupyter Notebook, Spyder, and Atom. All the testing is done in the Anaconda environment on the local machine. For the whole study, Python version 3.8.8 was employed. The PIP package is needed to set up the IDE. Minor csv CSV file checks are also performed using Microsoft Excel. The built-in git version control facilities in Visual Studio are occasionally used to manage the repository's branches. Microsoft Visual Studio is the main IDE used for coding. For version control I have used Github. I also have used other tools like Jupyter Notebook, Spyder, Atom for coding purposes. All the testing are done with the local Anaconda environment. Python version 3.8.8 is used for the whole research. PIP package is used for configuring the IDE. Microsoft excel is also used for minor csv file inspections. Visual Studio’s inbuilt git version control features are used time to time for managing the branches in the repository. MS Word and notepad are used for reporting and notes. PowerPoint is a presentation software that allows you to create slides. PDF files are managed with Adobe Acrobat DC. !pip and git commands are run via the built-in terminal in Visual Studio, Anaconda Prompt, and Windows Terminal. The entire project is run on the Windows operating system. The browsers utilized in this experiment are Google Chrome and Mozilla Firefox.

**5.0 Ethical Considerations**

* + 1. To begin, consider some recent unethical events. When a researcher's action has a negative impact on participants or society, it is considered harm. There are a number of reasons why researchers' actions cause so much harm for society or people. Likewise, one of the most obvious examples of such accidents was Human Radiation Experiments (2017) was one of the biggest examples of such incidents. In 1994 US President Clinton created an advisory team to research human radiation that has been conducted over the years. In this study, doctors injected Plutonium into the body of many patients and many of them did not consent to be part of this study. Also, there was a company called Quaker Oats (2020) which is also part of this study included radioactive components in oatmeal and were unknowingly fed to the children.
    2. In this study, no such experiment is done on humans in the process of data collection or analysis. An aggregated suicide dataset only provides information about the country's general population and related detail as features is used throughout the research. No prior experiment is conducted to gather data for this research. No harm is made to any subject in this regard. There are several benefits related to the data. Data provides an overview of how many suicides are happening from time to time. Talking about the societal impact of this research is enormous. For example, Study of Benefits of Electric Cars (2016) has created a significant impact on how this research has benefited society to help understand the carbon footprint reduction and cost-saving. In suicide analysis, I am trying to make use of data to leverage suicide attempts by helping the government to take measures or policies from the outcome of this study it’s going to help create plans to tackle such acts in coming years.
    3. This study of suicide analysis was based on a dataset that is open source and available for download on Kaggle. In terms of data storage and security, I wouldn't call it a particularly sensitive dataset since, for starters, this information is not private on the internet, and the creator has left public access open. Second, this suicide dataset does not contain any personally identifiable information; rather, it is a summary dataset that provides broad statistics on the country's mortality rates. Also, information on who is more prone to commit suicide, such as age group, internet users, human development index, and so on.

**5.0.1 SWOT Analysis**

1. The suicide dataset, as previously indicated, did not contain any personal information. I strongly believe that in future research, we should include more humans in the trial in order to get data from individuals in real time. The most critical step is to obtain written consent from everyone who wishes to participate in the study. Consider what would happen if these human individuals who are participating in the study/experiment had not given their consent. They may eventually take us to court is one of the real-life examples where in 1942 prisoners were asked to undergo dangerous experiments to understand the survival chance of soldiers sometimes even leading to deaths. Understanding personal, social, and business impacts of data practice.
2. In addition, even sharing information of individual sharing with any other colleagues or third party would be through proper procedure and getting signs on consent forms.
3. **Strength: -** In this study, I am trying to see suicide rates in different countries from time to time. This research strength is its dynamic nature. Similar weather forecast of google or Microsoft, this model will be run from time to time based on the latest data. This research aims at tackling suicide tendencies in every country’s population. This research is going to predict how many people are going to commit suicide in the next 5 years in different countries or continents. When working with a socially responsible research project, it is going to stand out in the world of the internet. Similar to the websites showcasing covid trends live, this website is also going to show the same impact of suicide numbers and create respective visualizations for any general audience to easily understand what the trend in data would be.
4. **Weakness: -** For forecasting, the data is aggregated, and no unique information about individuals is provided. As a result, I believe the data must have more precise traits that may be used to create reliable suicide predictions. However, if additional characteristics could have been added to the dataset, the model would have been more accurate. Things like topic diagnostic information, population happiness index, education index, and each country's happiness index. As a result, this forecast is more of a broad grasp of data trends.
5. **Opportunity: -** It's difficult to put into words how much can be learned through suicide data analysis. The government is working to figure out what causes suicides and how to reduce the number of suicides each year. We can build creative strategies to lessen the effect of suicides by studying and interpreting current statistics. Machine Learning models might be used to develop smart apps that help mobile users based on their activity data. Suicide analysis ushers in a new era of artificial intelligence in which we can track who is on the verge of dying.
6. Now let’s look at this data and its opportunities. Have you ever thought of having a suicide prediction model for each country? The wide range of opportunities using AI and the Time Series model on big data is possible using current technologies. Internet of things, cloud computing and Machine Learning are the best examples of state-of-the-art technologies. The suicide prediction model and live dashboard visualization is a great analysis model which any growing business can take inspiration from. Just imagine a burger selling vendor creating a live predicting model of a specific kind of burger that is sold at a particular season of a year? or maybe checking bestselling milkshakes each month? Wouldn’t these analyses make them grow? or even predict how many products are going to be sold in the coming months so they can prepare their store for the coming period to avoid lack of materials. Thus, this model is ultimately showing what kind of predictions or analysis our business and health industry need today to go smarter and do smarter businesses.
7. **Threat**: - Data may be utilized in a variety of ways. Some individuals utilized it for good causes, while others exploited it in a different way. It's possible that the suicide data will be abused in some way. However, from this perspective, they are less likely to occur if we do not provide detailed information about people. In this scenario, I'd argue that if new features are added to the model in the future, I'll have to change the model statically and then make it dynamic using cron tasks. Furthermore, additional storage space may be soon necessary when it comes to keeping individual information, and this model may function badly due to server needs. Even if we have alternative possibilities for purchasing cloud storage space, it will still be more expensive; still, I will have to find ways to enhance the needs. When it comes to the examination of prior years' suicides, the differences in counts over different political administration eras might have a political influence.

**6.0 Conclusion**

Models such as ARIMA, VAR, and AR were utilized to investigate and forecast the effect of suicide in different nations in this work. All of the models were built using the 'Russian Federation' suicide dataset as a starting point. When compared against other algorithms, the ARIMA model outperformed the others. One of the models, called SVM, has been found as a good fit for working on the suicide dataset, which must be done alongside other models that have already been completed. In comparison to other models, ARIMA and Decision tree classifiers provided me with greater accuracy. From this point forward, I'll be working on the primary dashboard, which is an online tool that makes real-time forecasts based on data input.

**7.0 Proposed Future Analysis**

The preliminary analysis of this research reveals that additional specific data is necessary for the project. In order to produce more accurate and meaningful forecasts in the future, daily suicide data with more precise personal information will need to be obtained. Future development will involve expanding the backend options to include a form with more fields particular to the person who died so that individual data may be collected. For instance, the person's blood group, health information, alcohol use, and so on. More models, such as SVM, will be used in future study to assist us understand the best approaches for forecasting and achieving optimal results.

**8.0 Deployment**

Python Dashboard has been deployed to Mochahost server initially. Getting server ready with required python package is very important step in the deployment stage. Several blogs regarding python dash app (2020) have been referenced in order figure out what is the procedure for hosting the python dash app.

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The first step is to run the data\_maker.py file which takes the data from the assets/data folder. Raw data is stored inside this folder where the data\_maker initially prepare the data and merge the dataset and export to output.csv file making it ready for further processing.

**Diagram

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*Fig x.x.x*

Second file called data\_modelling run the models using output from first file mentioned above to create models and make forecasted and normal data based on each country name from the data available. These country wise files then are loaded into the dashboard pages using normal python dashboard programming and forecasting is made accordingly.

On the other hand, all the other graphs will be made with real time data loaded from the mysql database to make live visualisations in the frontend. Initially python dash app is pushed into the GitHub and Heroku repositories to maintain the version control technology. Then the repo has been cloned into the Mochahost server in order to run the app. The process is shown in the fig x.x.x

Admin panel backend section done in PHP Laravel the second hosted app. It shares the same MySQL database to keep the records UpToDate. So, as we move on, new suicide records around the world will be recorded in the admin panel suicide records section. One of the books helped in the development of admin panel is called ‘Mastering Laravel’ (Pecoraro et al., 2015). Proper research concerning the security features in Laravel framework (Aborujilah et al., 2022) compared with other popular ones has been referenced before choosing Laravel for safely data handling from the backend.

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