

# Analysing Economic Data from FRED in Python

Automatically Pulling and Analyzing Data from the FRED Databased in Python

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## Background

In this analysis, I use data from the Federal Reserve bank of St Louis Database (FRED) to illustrate data analysis using Python and Pandas. As a starting point, I load the required packages- Pandas, Numpy, Matplotlib, Seaborn, and Plotly. The package `fredapi` does not come into Python by default. I hence install it and then load it.

## Objectives

## Summary of Results

The broad objective of the analysis is to showcase working with data from the FRED api in python.

The secondary objective of the analysis are as follows:

## Data

I start by installing the packages `fredapi` and `plotly`.

```
```{python}
# !pip install fredapi
# !pip install plotly
```
```

Next, I load the required modules: matplotlib, numpy, seaborn, pandas, plotly, and fredapi.

```
```{python}
## Import the required modules
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import plotly.express as px
from fredapi import Fred

## Set options
plt.style.use("fivethirtyeight")
pd.set_option("max_columns", 20)
```
```

Next, we shift our attention to our data source: FRED. FRED database is owned and maintained by the federal reserve bank of St. Louis in the United States. The database contains a vast array of economic variables from countries across the globe. The [link](https://fred.stlouisfed.org/) to the database is <https://fred.stlouisfed.org/>. The database consists of 822,000 variables from 114 sources across the globe.

I load the data from the FRED api. To start downloading data, you have to request for an api key from FRED by creating an account. From your account, you can then get an api key. You will then create write code as follows:

```
fred_api = "your_fred_api_key_here"
```

I have hidden my API key for obvious reasons. Please generate your own api key from FRED.

Next, we create a Fred object to allow us to access the data.

```
```{python}
fred = Fred(api_key=fred_api)
```
```

From here, we can search or pull data from FRED.

```
```{python}
fred.search("Kenya")
```
```

| series id         | id                | realtime_start | realtime_end | title      |
|-------------------|-------------------|----------------|--------------|------------|
| PTEAUSDM          | PTEAUSDM          | 2023-08-21     | 2023-08-21   | Global pr  |
| PTEAUSDA          | PTEAUSDA          | 2023-08-21     | 2023-08-21   | Global pr  |
| PTEAUSDQ          | PTEAUSDQ          | 2023-08-21     | 2023-08-21   | Global pr  |
| FXRATEKEA618NUPN  | FXRATEKEA618NUPN  | 2023-08-21     | 2023-08-21   | Exchange   |
| DDEI06KEA156NWDB  | DDEI06KEA156NWDB  | 2023-08-21     | 2023-08-21   | Bank's Re  |
| DDSI03KEA156NWDB  | DDSI03KEA156NWDB  | 2023-08-21     | 2023-08-21   | Bank Cap   |
| HCIYISKEA066NRUG  | HCIYISKEA066NRUG  | 2023-08-21     | 2023-08-21   | Index of I |
| SLUEM1524ZSKEN    | SLUEM1524ZSKEN    | 2023-08-21     | 2023-08-21   | Youth Un   |
| KENNGDPRPCPCPPPT  | KENNGDPRPCPCPPPT  | 2023-08-21     | 2023-08-21   | Real GDE   |
| PPPTTLKEA618NUPN  | PPPTTLKEA618NUPN  | 2023-08-21     | 2023-08-21   | Purchasin  |
| KENNGDPXORPCPPPT  | KENNGDPXORPCPPPT  | 2023-08-21     | 2023-08-21   | Real Non-  |
| PLOINVKEA624NUPN  | PLOINVKEA624NUPN  | 2023-08-21     | 2023-08-21   | Price Lev  |
| PLICPPKEA670NRUG  | PLICPPKEA670NRUG  | 2023-08-21     | 2023-08-21   | Price Lev  |
| PATENT4NKEUTILITY | PATENT4NKEUTILITY | 2023-08-21     | 2023-08-21   | U.S. Gran  |
| CIPPPGKEA156NUPN  | CIPPPGKEA156NUPN  | 2023-08-21     | 2023-08-21   | Investmen  |
| PLGDPOKEA670NRUG  | PLGDPOKEA670NRUG  | 2023-08-21     | 2023-08-21   | Price leve |
| RGDPLPKEA625NUPN  | RGDPLPKEA625NUPN  | 2023-08-21     | 2023-08-21   | Purchasin  |
| DDAI01KEA642NWDB  | DDAI01KEA642NWDB  | 2023-08-21     | 2023-08-21   | Number o   |
| DDEI05KEA156NWDB  | DDEI05KEA156NWDB  | 2023-08-21     | 2023-08-21   | Bank's Re  |
| KENEREERIX        | KENEREERIX        | 2023-08-21     | 2023-08-21   | Real Effec |
| DDSI01KEA645NWDB  | DDSI01KEA645NWDB  | 2023-08-21     | 2023-08-21   | Bank Z-S   |
| SLEMPOTOTLSPZSKEN | SLEMPOTOTLSPZSKEN | 2023-08-21     | 2023-08-21   | Employment |
| KENNGDPRPCPPPT    | KENNGDPRPCPPPT    | 2023-08-21     | 2023-08-21   | Real Gros  |
| MKTGNIKEA646NWDB  | MKTGNIKEA646NWDB  | 2023-08-21     | 2023-08-21   | Gross Nat  |
| DDSI06KEA156NWDB  | DDSI06KEA156NWDB  | 2023-08-21     | 2023-08-21   | Liquid As  |
| FPCPITOTLZGKEN    | FPCPITOTLZGKEN    | 2023-08-21     | 2023-08-21   | Inflation, |
| MKTGDPKEA646NWDB  | MKTGDPKEA646NWDB  | 2023-08-21     | 2023-08-21   | Gross Dor  |
| WUIKEN            | WUIKEN            | 2023-08-21     | 2023-08-21   | World Un   |
| SIPOVGINIKEN      | SIPOVGINIKEN      | 2023-08-21     | 2023-08-21   | GINI Inde  |
| RTFPNAKEA632NRUG  | RTFPNAKEA632NRUG  | 2023-08-21     | 2023-08-21   | Total Fac  |
| KENGXGXG01GDPPT   | KENGXGXG01GDPPT   | 2023-08-21     | 2023-08-21   | Total Exp  |
| KENFCNODMFNUM     | KENFCNODMFNUM     | 2023-08-21     | 2023-08-21   | Use of Fir |
| DDDI03KEA156NWDB  | DDDI03KEA156NWDB  | 2023-08-21     | 2023-08-21   | Non-Bank   |
| DDSM01KEA066NWDB  | DDSM01KEA066NWDB  | 2023-08-21     | 2023-08-21   | Volatility |
| DDOE02KEA086NWDB  | DDOE02KEA086NWDB  | 2023-08-21     | 2023-08-21   | Consumer   |
| DDOE01KEA086NWDB  | DDOE01KEA086NWDB  | 2023-08-21     | 2023-08-21   | Consumer   |
| DDDM01KEA156NWDB  | DDDM01KEA156NWDB  | 2023-08-21     | 2023-08-21   | Stock Ma   |
| NYGDPPCAPKDKEN    | NYGDPPCAPKDKEN    | 2023-08-21     | 2023-08-21   | Constant   |
| RKNANPKEA666NRUG  | RKNANPKEA666NRUG  | 2023-08-21     | 2023-08-21   | Capital S  |
| PCAGDPKEA646NWDB  | PCAGDPKEA646NWDB  | 2023-08-21     | 2023-08-21   | Gross Dor  |
| DDSI02KEA156NWDB  | DDSI02KEA156NWDB  | 2023-08-21     | 2023-08-21   | Bank Non   |
| SEADTLITRZSKEN    | SEADTLITRZSKEN    | 2023-08-21     | 2023-08-21   | Literacy I |
| RGDPNAKEA666NRUG  | RGDPNAKEA666NRUG  | 2023-08-21     | 2023-08-21   | Real GDE   |
| DDOI02KEA156NWDB  | DDOI02KEA156NWDB  | 2023-08-21     | 2023-08-21   | Bank Dep   |
| KENENEERIX        | KENENEERIX        | 2023-08-21     | 2023-08-21   | Nominal I  |
| KGPPPGKEA156NUPN  | KGPPPGKEA156NUPN  | 2023-08-21     | 2023-08-21   | Governme   |
| DDSI04KEA156NWDB  | DDSI04KEA156NWDB  | 2023-08-21     | 2023-08-21   | Bank Cre   |
| KENFCBODCKNUM     | KENFCBODCKNUM     | 2023-08-21     | 2023-08-21   | Geograph   |
| MNKENA052SCEN     | MNKENA052SCEN     | 2023-08-21     | 2023-08-21   | Value of I |
| KENFCSODMFXDC     | KENFCSODMFXDC     | 2023-08-21     | 2023-08-21   | Use of Fir |
| KENDGCDRPT        | KENDGCDRPT        | 2023-08-21     | 2023-08-21   | Official F |

We see that we have 365 variables (the rows in the search output) available for Kenya. It is now up to us to choose what we desire to analyse.

Now let us get the data on the net issues of international debt securities by the government of Kenya. This is the last item in our search list. In downloading the data, we provide python with the search id which is IDSOFAMRINIKE in this case.

```
```{python}
kenya_debt = fred.get_series(series_id="IDSOFAMRINIKE")

kenya_debt.shape
```
```

(112,)

Let us peek into the first few rows of the series.

```
```{python}
kenya_debt.head()
```
```

|            | 0   |
|------------|-----|
| 1987-01-01 | NaN |
| 1987-04-01 | NaN |
| 1987-07-01 | NaN |
| 1987-10-01 | NaN |
| 1988-01-01 | NaN |

Let us also see the last few rows of the series.

```
```{python}
kenya_debt.tail()
```
```

|            | 0   |
|------------|-----|
| 2013-10-01 | NaN |
| 2014-01-01 | NaN |
| 2014-04-01 | NaN |
| 2014-07-01 | NaN |
| 2014-10-01 | NaN |

Let us see the null values in the data.

```

```{python}
kenya_debt.isna().sum()
```

```

109

There appears to be 109 null values in the data out of the possible 112 values. We are out of luck here.

As an alternative, let us look at the Kenya shilling-US Dollar exchange rate.

```

```{python}
exchange = fred.get_series('FXRATEKEA618NUPN')
```

```

We examine the first and last rows of the series. It appears we have some reasonable data here from 1950 to 2010.

```

```{python}
exchange.head()
```

```

|            | 0        |
|------------|----------|
| 1950-01-01 | 7.140861 |
| 1951-01-01 | 7.140861 |
| 1952-01-01 | 7.140861 |
| 1953-01-01 | 7.140861 |
| 1954-01-01 | 7.140861 |

```

```{python}
exchange.tail()
```

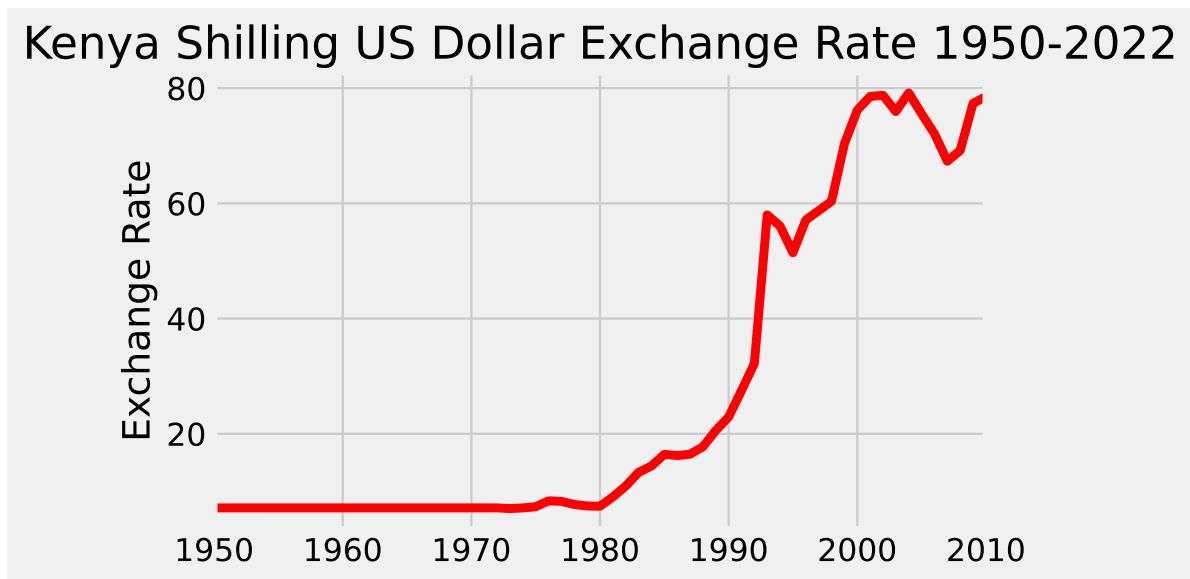
```

|            | 0         |
|------------|-----------|
| 2006-01-01 | 72.100835 |
| 2007-01-01 | 67.317638 |
| 2008-01-01 | 69.175320 |
| 2009-01-01 | 77.352012 |
| 2010-01-01 | 78.539167 |

## Analysis

```
{python}  
exchange.plot(title = "Kenya Shilling US Dollar Exchange Rate 1950-2022", ylabel = "Exchange Rate")  
}
```

<AxesSubplot:title={'center': 'Kenya Shilling US Dollar Exchange Rate 1950-2022'}, ylabel='Exchange Rate'>



## Exploration

## In-depth analysis

## Conclusion

## References