```
In [1]:
            import pandas as pd
            import matplotlib.pyplot as plt
            import warnings
            import matplotlib.style as style
            exchange_rates = pd.read_csv("euro-daily-hist_1999_2022.csv")
  In [2]:
  In [3]:
            exchange_rates
  Out[3]:
                                                                                   [Chinese
                                                                           [Swiss
                                                                                                      [Czech
                               [Australian [Bulgarian [Brazilian [Canadian
                                                                                      yuan
                                                                                             [Cypriot
                   Period\Unit:
                                                                            franc
                                                                                                      koruna
                                   dollar]
                                                                   dollar]
                                                lev ]
                                                          real ]
                                                                                   renminbi
                                                                                             pound ]
                                                                                1
                                                                                                           1
                                                                                          ]
                    2023-12-15
                0
                                    1.6324
                                               1.9558
                                                         5.4085
                                                                    1.4653 0.9488
                                                                                     7.7812
                                                                                                      24.477
                                                                                                NaN
                    2023-12-14
                                    1.6288
                                               1.9558
                                                         5.3349
                                                                    1.4677
                                                                            0.949
                                                                                     7.7866
                                                                                                NaN
                                                                                                      24.408
                    2023-12-13
                                    1.6452
                                               1.9558
                                                                    1.4644 0.9452
                                                                                     7.7426
                                                                                                      24.476
                                                         5.3609
                                                                                                NaN
                3
                    2023-12-12
                                    1.6398
                                                                    1.4656 0.9443
                                                                                     7.7447
                                               1.9558
                                                         5.3327
                                                                                                NaN
                                                                                                       24.42
                    2023-12-11
                                    1.642
                                               1.9558
                                                         5.3169
                                                                    1.4609 0.9478
                                                                                     7.7206
                                                                                                NaN
                                                                                                      24.367
            6451
                    1999-01-08
                                    1.8406
                                                NaN
                                                           NaN
                                                                    1.7643 1.6138
                                                                                       NaN
                                                                                             0.58187
                                                                                                      34.938
                    1999-01-07
                                                                                             0.58187
            6452
                                    1.8474
                                                NaN
                                                           NaN
                                                                    1.7602 1.6165
                                                                                       NaN
                                                                                                      34.886
            6453
                    1999-01-06
                                    1.8820
                                                NaN
                                                           NaN
                                                                                             0.58200
                                                                                                      34.850
                                                                    1.7711 1.6116
                                                                                       NaN
            6454
                    1999-01-05
                                    1.8944
                                                NaN
                                                                    1.7965 1.6123
                                                                                             0.58230
                                                           NaN
                                                                                       NaN
                                                                                                      34.917
            6455
                    1999-01-04
                                    1.9100
                                                NaN
                                                           NaN
                                                                    1.8004 1.6168
                                                                                       NaN
                                                                                             0.58231
                                                                                                      35.107
           6456 rows × 41 columns
4
  In [4]:
            #As we already know that periods = time but fro individual perspective it is difficult
            exchange_rates.rename(columns = {'[Chinese yuan renminbi]' : 'Chinese yuan renminbi',
  In [5]:
            exchange_rates['Time'] = pd.to_datetime(exchange_rates['Time'])
  In [6]:
```

exchange_rates

In [7]:

Out[7]:

| | | Time | [Australian dollar] | [Bulgarian lev] | [Brazilian real] | [Canadian dollar] | [Swiss franc] | Chinese yuan renminbi | [Cypriot pound] | [Czech koruna] | [Danis krone |
|---|------|----------------|------------------------|--------------------|----------------------|----------------------|----------------------|-----------------------------|-----------------|-----------------------|-----------------|
| - | 0 | 2023- 12-15 | 1.6324 | 1.9558 | 5.4085 | 1.4653 | 0.9488 | 7.7812 | NaN | 24.477 | 7.45! |
| | 1 | 2023- 12-14 | 1.6288 | 1.9558 | 5.3349 | 1.4677 | 0.949 | 7.7866 | NaN | 24.408 | 7.456 |
| | 2 | 2023- 12-13 | 1.6452 | 1.9558 | 5.3609 | 1.4644 | 0.9452 | 7.7426 | NaN | 24.476 | 7.456 |
| | 3 | 2023- 12-12 | 1.6398 | 1.9558 | 5.3327 | 1.4656 | 0.9443 | 7.7447 | NaN | 24.42 | 7.456 |
| | 4 | 2023- 12-11 | 1.642 | 1.9558 | 5.3169 | 1.4609 | 0.9478 | 7.7206 | NaN | 24.367 | 7.45(|
| | ••• | | | | | | | | | | |
| | 6451 | 1999- 01-08 | 1.8406 | NaN | NaN | 1.7643 | 1.6138 | NaN | 0.58187 | 34.938 | 7.443 |
| | 6452 | 1999- 01-07 | 1.8474 | NaN | NaN | 1.7602 | 1.6165 | NaN | 0.58187 | 34.886 | 7.443 |
| | 6453 | 1999- 01-06 | 1.8820 | NaN | NaN | 1.7711 | 1.6116 | NaN | 0.58200 | 34.850 | 7.44! |
| | 6454 | 1999- 01-05 | 1.8944 | NaN | NaN | 1.7965 | 1.6123 | NaN | 0.58230 | 34.917 | 7.449 |
| | 6455 | 1999- 01-04 | 1.9100 | NaN | NaN | 1.8004 | 1.6168 | NaN | 0.58231 | 35.107 | 7.45(|

6456 rows × 41 columns

In [8]: exchange_rates = exchange_rates[exchange_rates['Chinese yuan renminbi'] != '-']
 exchange_rates['Chinese yuan renminbi'] = exchange_rates['Chinese yuan renminbi'].asty
 exchange_rates

C:\Users\iqra com\AppData\Local\Temp\ipykernel_22452\1107561610.py:2: SettingWithCopy
Warning:

A value is trying to be set on a copy of a slice from a DataFrame. Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy

exchange_rates['Chinese yuan renminbi'] = exchange_rates['Chinese yuan renminbi'].a
stype(float)

Out[8]:

| | Time | [Australian dollar] | [Bulgarian lev] | [Brazilian real] | [Canadian dollar] | [Swiss franc] | Chinese yuan renminbi | [Cypriot pound] | [Czech koruna] | [Danis krone |
|------|----------------|------------------------|--------------------|----------------------|----------------------|----------------------|-----------------------------|---------------------|-----------------------|-----------------|
| 0 | 2023- 12-15 | 1.6324 | 1.9558 | 5.4085 | 1.4653 | 0.9488 | 7.7812 | NaN | 24.477 | 7.45! |
| 1 | 2023- 12-14 | 1.6288 | 1.9558 | 5.3349 | 1.4677 | 0.949 | 7.7866 | NaN | 24.408 | 7.456 |
| 2 | 2023- 12-13 | 1.6452 | 1.9558 | 5.3609 | 1.4644 | 0.9452 | 7.7426 | NaN | 24.476 | 7.456 |
| 3 | 2023- 12-12 | 1.6398 | 1.9558 | 5.3327 | 1.4656 | 0.9443 | 7.7447 | NaN | 24.42 | 7.456 |
| 4 | 2023- 12-11 | 1.642 | 1.9558 | 5.3169 | 1.4609 | 0.9478 | 7.7206 | NaN | 24.367 | 7.456 |
| | | | | | | | | | | |
| 6451 | 1999- 01-08 | 1.8406 | NaN | NaN | 1.7643 | 1.6138 | NaN | 0.58187 | 34.938 | 7.443 |
| 6452 | 1999- 01-07 | 1.8474 | NaN | NaN | 1.7602 | 1.6165 | NaN | 0.58187 | 34.886 | 7.443 |
| 6453 | 1999- 01-06 | 1.8820 | NaN | NaN | 1.7711 | 1.6116 | NaN | 0.58200 | 34.850 | 7.44! |
| 6454 | 1999- 01-05 | 1.8944 | NaN | NaN | 1.7965 | 1.6123 | NaN | 0.58230 | 34.917 | 7.449 |
| 6455 | 1999- 01-04 | 1.9100 | NaN | NaN | 1.8004 | 1.6168 | NaN | 0.58231 | 35.107 | 7.45(|

6395 rows × 41 columns

```
In []:
In [9]: exchange_rates.reset_index(drop=True, inplace=True)
In [10]: chinese_to_dollar = exchange_rates[['Time' , 'Chinese yuan renminbi']].copy()
In [11]: exchange_rates.info()
```

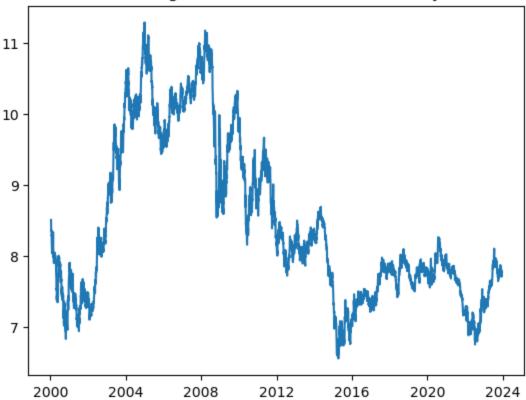
```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 6395 entries, 0 to 6394
Data columns (total 41 columns):
```

```
#
     Column
                            Non-Null Count Dtype
     -----
                            -----
0
     Time
                            6395 non-null
                                            datetime64[ns]
1
     [Australian dollar ]
                            6395 non-null
                                            object
2
     [Bulgarian lev ]
                            5996 non-null
                                            object
3
     [Brazilian real ]
                            6127 non-null
                                            object
4
     [Canadian dollar ]
                            6395 non-null
                                            object
     [Swiss franc ]
                            6395 non-null
                                            object
6
    Chinese yuan renminbi 6127 non-null
                                            float64
7
     [Cypriot pound ]
                            2305 non-null
                                            obiect
8
                            6395 non-null
     [Czech koruna ]
                                            object
9
     [Danish krone ]
                            6395 non-null
                                            object
10
     [Estonian kroon ]
                            3075 non-null
                                            object
    [UK pound sterling ]
                            6395 non-null
                                            object
12
    [Greek drachma ]
                            515 non-null
                                            object
     [Hong Kong dollar ]
                            6395 non-null
                                            object
14
     [Croatian kuna ]
                            5880 non-null
                                            object
15
     [Hungarian forint ]
                            6395 non-null
                                            object
16
     [Indonesian rupiah ]
                            6395 non-null
                                            object
    [Israeli shekel ]
17
                            6127 non-null
                                            object
    [Indian rupee ]
                            6127 non-null
                                            object
                            4049 non-null
                                            float64
19
    [Iceland krona ]
20
     [Japanese yen ]
                            6395 non-null
                                            object
    [Korean won ]
                            6395 non-null
                                            object
22
    [Lithuanian litas ]
                            4098 non-null
                                            object
23
     [Latvian lats ]
                            3843 non-null
                                            object
24
    [Maltese lira ]
                            2305 non-null
                                            object
    [Mexican peso ]
                            6395 non-null
                                            object
25
     [Malaysian ringgit ]
                            6395 non-null
                                            object
27
                            6395 non-null
     [Norwegian krone ]
                                            object
     [New Zealand dollar ]
                            6395 non-null
                                            object
     [Philippine peso ]
29
                            6395 non-null
                                            object
30
    [Polish zloty ]
                            6395 non-null
                                            object
    [Romanian leu ]
                            6394 non-null
                                            float64
    [Russian rouble ]
                            5933 non-null
                                            object
32
33
    [Swedish krona ]
                            6395 non-null
                                            object
    [Singapore dollar ]
                            6395 non-null
                                            object
35
    [Slovenian tolar ]
                            2050 non-null
                                            object
36
    [Slovak koruna ]
                            2561 non-null
                                            object
    [Thai baht ]
37
                            6395 non-null
                                            object
38
    [Turkish lira ]
                            6394 non-null
                                            float64
39
     [US dollar ]
                            6395 non-null
                                            object
40 [South African rand ] 6395 non-null
                                            object
dtypes: datetime64[ns](1), float64(4), object(36)
memory usage: 2.0+ MB
```

```
In []:
In [12]: #chinese_to_dollar = chinese_to_dollar[chinese_to_dollar['Chinese yuan renminbi'] != '
#chinese_to_dollar['Chinese yuan renminbi'] = chinese_to_dollar['Chinese yuan renminbi
#chinese_to_dollar
In [13]: chinese_to_dollar.info()
```

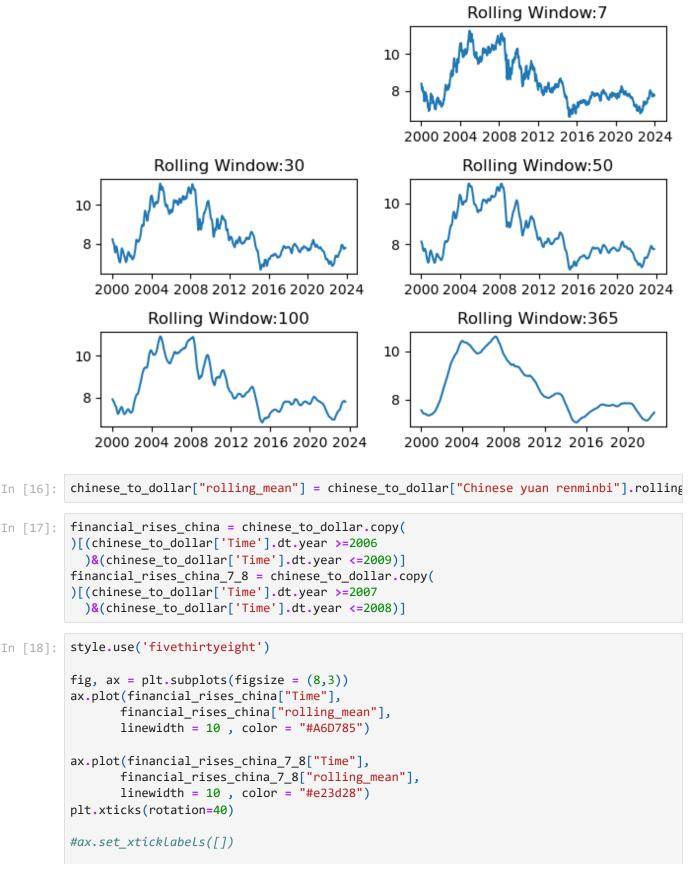
```
<class 'pandas.core.frame.DataFrame'>
         RangeIndex: 6395 entries, 0 to 6394
         Data columns (total 2 columns):
              Column
                                     Non-Null Count Dtype
              Time
                                     6395 non-null
                                                     datetime64[ns]
              Chinese yuan renminbi 6127 non-null
                                                     float64
         dtypes: datetime64[ns](1), float64(1)
         memory usage: 100.1 KB
         plt.title('This diagram also contain all the holidays')
In [14]:
         plt.plot(chinese_to_dollar['Time'], chinese_to_dollar['Chinese yuan renminbi'])
         plt.show()
```

This diagram also contain all the holidays



Now we make analysis on rolling meaning we find what insight we can get if we convert on week, days, month and year how much fluctuation are there

C:\Users\iqra com\AppData\Local\Temp\ipykernel_22452\1713599069.py:6: MatplotlibDepre cationWarning: Auto-removal of overlapping axes is deprecated since 3.6 and will be removed two minor releases later; explicitly call ax.remove() as needed. plt.subplot(3,2,i)



```
#x = 0.02
         #for year in ["2006", '2007', "2008", "2009", "2010"]:
               ax.text(x, -0.02, aplha=0.5, fontsize=30, transform=plt.gca().transAxes)
               x += 0.22888
         #ax.set_xticklabels([])
         (array([13149., 13330., 13514., 13695., 13879., 14061., 14245., 14426.,
Out[18]:
                  14610.]),
          [Text(13149.0, 0, '2006-01'),
           Text(13330.0, 0, '2006-07'),
           Text(13514.0, 0, '2007-01'),
           Text(13695.0, 0, '2007-07'),
           Text(13879.0, 0, '2008-01'),
           Text(14061.0, 0, '2008-07'),
           Text(14245.0, 0, '2009-01'),
           Text(14426.0, 0, '2009-07'),
           Text(14610.0, 0, '2010-01')])
          11.0
          10.5
          10.0
           9.5
           9.0
```

EFFECT OF COVID 19 ONTO CHINA ECONOMY

```
Out[20]: (array([16436., 16801., 17167., 17532., 17897., 18262., 18628.]),
        [Text(16436.0, 0, '2015'),
        Text(16801.0, 0, '2016'),
        Text(17532.0, 0, '2018'),
        Text(17532.0, 0, '2019'),
        Text(18262.0, 0, '2020'),
        Text(18628.0, 0, '2021')])

8.2

8.0

7.8

7.6

7.4

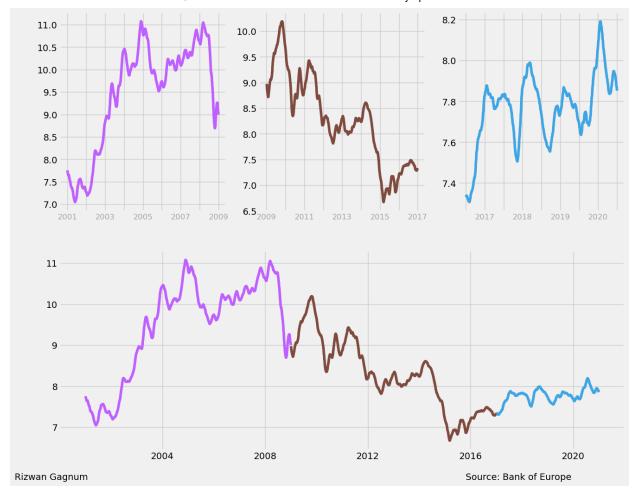
7.2
```

As you can see from above grapgh covid have siginifican effect ON chinese monetary and economy.

Major event that happen in china that effect china economy as you can see in the graph

```
In [21]: bush_obama_trump = chinese_to_dollar.copy(
         )[(chinese_to_dollar['Time'].dt.year >=2001
           )&(chinese_to_dollar['Time'].dt.year <2021)]</pre>
          china1 = bush_obama_trump.copy(
         )[bush_obama_trump['Time'].dt.year < 2009]
          china2 = bush_obama_trump.copy(
          )[(bush_obama_trump['Time'].dt.year >=2009
            )&(bush_obama_trump['Time'].dt.year <2017)]</pre>
         china3 = bush_obama_trump.copy(
          )[(bush_obama_trump['Time'].dt.year >=2017
            )&(bush_obama_trump['Time'].dt.year < 2021)]</pre>
In [24]: style.use('fivethirtyeight')
         plt.figure(figsize = (14,18))
         \#x = plt.subplot(3,3,1)
         ax1 = plt.subplot(3,3,1)
         ax2 = plt.subplot(3,3,2)
         ax3 = plt.subplot(3,3,3)
         ax4 = plt.subplot(3,1,2)
         #AX1
          ax1.plot(china1['Time'] , china1['rolling_mean'],
                  color = '#BF5FFF')
```

```
ax1.set_xticklabels(['', '2001', '', '2003', '', '2005', '', '2007', '', '2009
                   alpha = 0.3, size = 12)
ax1.text(0.11 , 0.25 , 'Exchange rate' , fontsize = 12 , weight = 'bold',
        color = '#BF5FFF' , transform = plt.gca().transAxes)
ax1.text(0.093, 0.25, '(2001-2009)', weight = 'bold',
         alpha = 0.3, transform = plt.gca().transAxes)
##AX2
ax2.plot(china2['Time'] , china2['rolling_mean'],
        color = '#7F4A3E')
ax2.set_xticklabels(['', '2009', '', '2011', '', '2013', '', '2015', '', '2017
                   alpha = 0.3 , size = 12)
ax2.text(0.11 , 0.25 , 'Exchange rate' , fontsize = 12 , weight = 'bold',
        color = '#7F4A3E', transform = plt.gca().transAxes)
ax2.text(0.093, 0.25, '(2009-2017)', weight = 'bold',
        alpha = 0.3, transform = plt.gca().transAxes)
##AX3
ax3.plot(china3['Time'] , china3['rolling_mean'],
        color = '#3DA5E4')
ax3.set_xticklabels(['', '2017','', '2018','', '2019', '', '2020','', '2021',''],
                   alpha = 0.3, size = 12)
ax3.text(0.11 , 0.25 , 'Exchange rate' , fontsize = 12 , weight = 'bold',
        color = '#3DA5E4' , transform = plt.gca().transAxes)
ax3.text(0.093, 0.25, '(2017-2021)', weight = 'bold',
         alpha = 0.3, transform = plt.gca().transAxes)
##AX4
ax4.plot(china1['Time'] , china1['rolling_mean'],
        color = '#BF5FFF')
ax4.plot(china2['Time'] , china2['rolling_mean'],
        color = '#7F4A3E')
ax4.plot(china3['Time'] , china3['rolling_mean'],
        color = '#3DA5E4')
### Signature
ax4.text(-0.08 , -0.15, 'Rizwan Gagnum' + ' '*133 + 'Source: Bank of Europe',
         size =14 , transform = plt.gca().transAxes)
plt.tight_layout()
plt.show()
C:\Users\iqra com\AppData\Local\Temp\ipykernel_22452\392685953.py:12: UserWarning: Fi
xedFormatter should only be used together with FixedLocator
  ax1.set_xticklabels(['', '2001', '', '2003', '', '2005', '', '2007', '', '2
009'],
C:\Users\igra com\AppData\Local\Temp\ipykernel 22452\392685953.py:22: UserWarning: Fi
xedFormatter should only be used together with FixedLocator
  ax2.set_xticklabels(['', '2009', '', '2011', '', '2013', '', '2015', '', '2
C:\Users\iqra com\AppData\Local\Temp\ipykernel_22452\392685953.py:32: UserWarning: Fi
xedFormatter should only be used together with FixedLocator
  ax3.set_xticklabels(['', '2017','', '2018','', '2019', '', '2020','', '202
1',''],
C:\Users\iqra com\AppData\Local\Temp\ipykernel 22452\392685953.py:49: UserWarning: Ti
ght layout not applied. tight_layout cannot make axes width small enough to accommoda
te all axes decorations
plt.tight_layout()
```



Conclusion

In conclusion, the exchange rates of China's currency, the Yuan Renminbi, have been significantly influenced by major events over the years. For instance, the global financial crisis of 2008 had a profound impact, leading to a decrease in the value of the Chinese Yuan as economic uncertainties prevailed.

On the contrary, strategic economic initiatives like the Belt and Road Initiative and periods of robust economic growth have contributed to an increase in the value of the Yuan. Additionally, shifts in exchange rate policies by the People's Bank of China, trade tensions with other nations, and geopolitical events have played crucial roles in shaping the fluctuations in China's currency exchange rates.

These major events highlight the sensitivity and responsiveness of the Chinese Yuan to changes in the economic and geopolitical landscape. A nuanced understanding of these events is essential for comprehending the dynamics behind the fluctuations, as they underscore the intricate relationship between global events and the valuation of China's currency in the international market.

In []: