```
In [1]:
            import pandas as pd
            import matplotlib.pyplot as plt
            import warnings
             import matplotlib.style as style
   In [2]:
            exchange_rates = pd.read_csv("euro-daily-hist_1999_2022.csv")
   In [3]:
            exchange_rates
                                                                                    [Chinese
   Out[3]:
                                                                            [Swiss
                                                                                                       [Czech
                                [Australian [Bulgarian [Brazilian [Canadian
                                                                                             [Cypriot
                                                                                       yuan
                   Period\Unit:
                                                                             franc
                                                                                                       koruna
                                   dollar]
                                                 lev ]
                                                          real ]
                                                                   dollar ]
                                                                                   renminbi
                                                                                             pound ]
                                                                                 1
                                                                                                            1
                                                                                           ]
                0
                    2023-12-15
                                    1.6324
                                               1.9558
                                                         5.4085
                                                                    1.4653 0.9488
                                                                                      7.7812
                                                                                                       24.477
                                                                                                 NaN
                    2023-12-14
                                    1.6288
                                               1.9558
                                                         5.3349
                                                                             0.949
                                                                                      7.7866
                                                                    1.4677
                                                                                                 NaN
                                                                                                       24.408
                    2023-12-13
                2
                                    1.6452
                                               1.9558
                                                         5.3609
                                                                    1.4644 0.9452
                                                                                      7.7426
                                                                                                       24.476
                                                                                                 NaN
                3
                    2023-12-12
                                    1.6398
                                               1.9558
                                                         5.3327
                                                                    1.4656 0.9443
                                                                                      7.7447
                                                                                                 NaN
                                                                                                        24.42
                4
                    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
            6452
                    1999-01-07
                                    1.8474
                                                 NaN
                                                           NaN
                                                                    1.7602 1.6165
                                                                                              0.58187
                                                                                                       34.886
                                                                                        NaN
            6453
                    1999-01-06
                                                                                              0.58200
                                    1.8820
                                                 NaN
                                                           NaN
                                                                    1.7711 1.6116
                                                                                        NaN
                                                                                                       34.850
            6454
                                                                                              0.58230
                    1999-01-05
                                    1.8944
                                                 NaN
                                                           NaN
                                                                    1.7965 1.6123
                                                                                        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	7.456
	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(

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_14108\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()
```

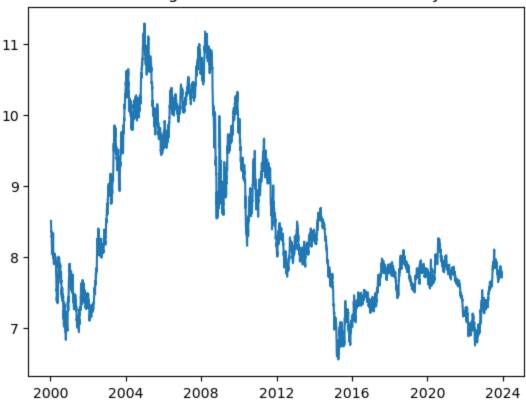
```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 6395 entries, 0 to 6394
Data columns (total 41 columns):
```

```
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
5
                            6395 non-null
     [Swiss franc ]
                                             object
6
     Chinese yuan renminbi 6127 non-null
                                             float64
7
     [Cypriot pound ]
                            2305 non-null
                                             object
8
     [Czech koruna ]
                            6395 non-null
                                             object
9
                            6395 non-null
     [Danish krone ]
                                             object
     [Estonian kroon ]
10
                            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
     [Croatian kuna ]
                            5880 non-null
                                             object
15
     [Hungarian forint ]
                            6395 non-null
                                             object
16
     [Indonesian rupiah ]
                            6395 non-null
                                             object
17
     [Israeli shekel ]
                            6127 non-null
                                             object
    [Indian rupee ]
                            6127 non-null
                                             object
19
    [Iceland krona ]
                            4049 non-null
                                             float64
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
                            6395 non-null
25
     [Mexican peso ]
                                             object
     [Malaysian ringgit ]
                            6395 non-null
                                             object
27
     [Norwegian krone ]
                            6395 non-null
                                             object
     [New Zealand dollar ]
                            6395 non-null
                                             object
29
     [Philippine peso ]
                            6395 non-null
                                             object
30
    [Polish zloty ]
                            6395 non-null
                                             object
    [Romanian leu ]
                            6394 non-null
                                             float64
32
    [Russian rouble ]
                            5933 non-null
                                             object
    [Swedish krona ]
                            6395 non-null
                                             object
33
    [Singapore dollar ]
                            6395 non-null
                                             object
35
    [Slovenian tolar ]
                            2050 non-null
                                             object
36
    [Slovak koruna ]
                            2561 non-null
                                             object
37
     [Thai baht ]
                            6395 non-null
                                             object
38
    [Turkish lira ]
                            6394 non-null
                                             float64
     [US dollar ]
                            6395 non-null
                                             object
    [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()
```

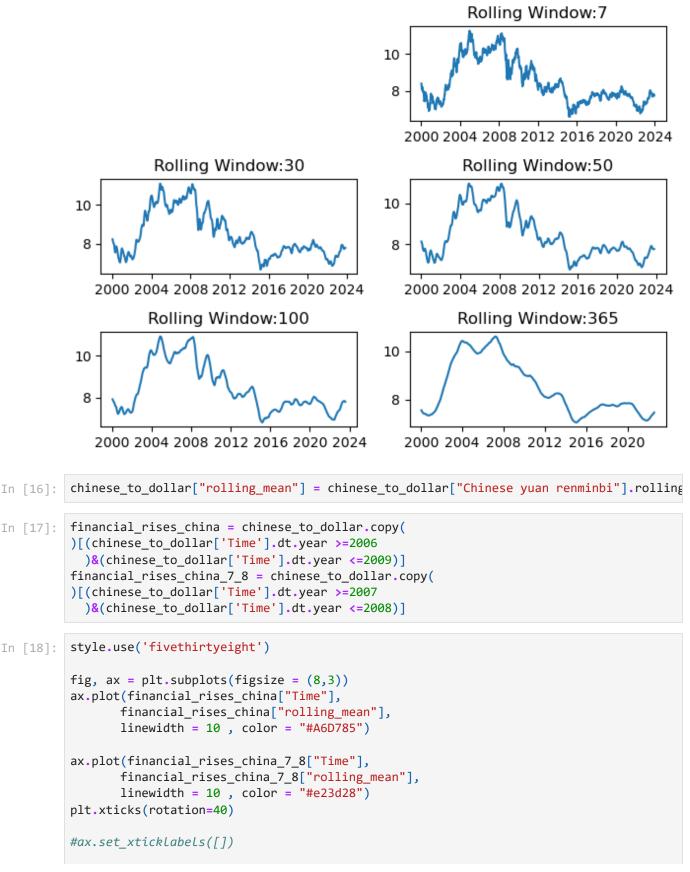
```
<class 'pandas.core.frame.DataFrame'>
         RangeIndex: 6395 entries, 0 to 6394
         Data columns (total 2 columns):
              Column
                                     Non-Null Count Dtype
              Time
                                     6395 non-null
                                                     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_14108\1713599069.py:6: MatplotlibDepre
cationWarning: Auto-removal of overlapping axes is deprecated since 3.6 and will be r
emoved 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 CORONA ONTO CHINA ECONOMY AND ON THEIR MONEY

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

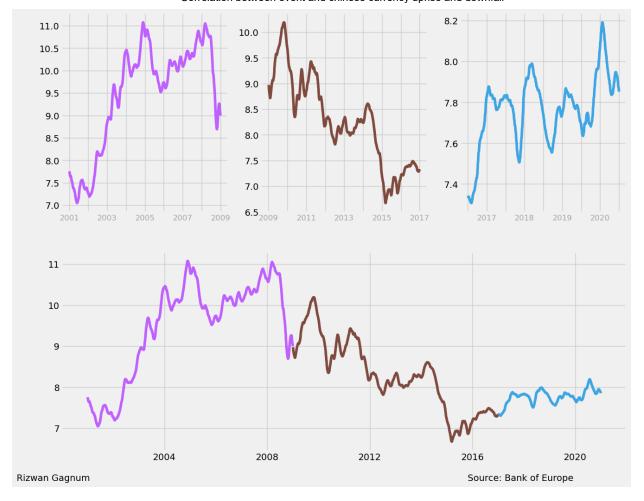
Bush obama and trump presedency effect on chinese currency

```
In [21]:
         bush obama trump = chinese to dollar.copy(
         )[(chinese_to_dollar['Time'].dt.year >=2001
           )&(chinese_to_dollar['Time'].dt.year <2021)]</pre>
         bush = bush_obama_trump.copy(
         )[bush_obama_trump['Time'].dt.year < 2009]
         obama = bush_obama_trump.copy(
         )[(bush_obama_trump['Time'].dt.year >=2009
           )&(bush_obama_trump['Time'].dt.year <2017)]</pre>
         trump = bush_obama_trump.copy(
         )[(bush_obama_trump['Time'].dt.year >=2017
           )&(bush_obama_trump['Time'].dt.year < 2021)]</pre>
In [31]: 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.plot(bush['Time'] , bush['rolling_mean'],
                  color = '#BF5FFF')
         ax1.set_xticklabels(['', '2001', '', '2003', '', '2005', '', '2007', '', '2009
```

alpha = 0.3, size = 12)

```
ax1.text(0.11 , 0.25 , 'BUSH' , 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(obama['Time'] , obama['rolling_mean'],
        color = '#7F4A3E')
ax2.set_xticklabels(['', '2009', '', '2011', '', '2013', '', '2015', '', '2017
                   alpha = 0.3, size = 12)
ax2.text(0.11 , 0.25 , 'OBAMA' , 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(trump['Time'] , trump['rolling_mean'],
        color = '#3DA5E4')
ax3 set_xticklabels(['' , '2017','', '2018','' , '2019' , '' , '2020','' , '2021',''],
                   alpha = 0.3, size = 12)
ax3.text(0.11 , 0.25 , 'Trump' , 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(bush['Time'] , bush['rolling_mean'],
        color = '#BF5FFF')
ax4.plot(obama['Time'] , obama['rolling_mean'],
        color = '#7F4A3E')
ax4.plot(trump['Time'] , trump['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 14108\2557110016.py:12: UserWarning: F
ixedFormatter should only be used together with FixedLocator
  ax1.set_xticklabels(['', '2001', '', '2003', '', '2005', '', '2007', '', '2
009'],
C:\Users\iqra com\AppData\Local\Temp\ipykernel_14108\2557110016.py:22: UserWarning: F
ixedFormatter should only be used together with FixedLocator
  ax2.set_xticklabels(['' , '2009' , '' , '2011' , '' , '2013' , '' , '2015' , '' ,'2
017'],
C:\Users\igra com\AppData\Local\Temp\ipykernel 14108\2557110016.py:32: UserWarning: F
ixedFormatter should only be used together with FixedLocator
 ax3.set_xticklabels(['' , '2017','', '2018','' , '2019' , '' , '2020','' , '202
1',''],
C:\Users\iqra com\AppData\Local\Temp\ipykernel_14108\2557110016.py:49: UserWarning: T
ight layout not applied, tight layout cannot make axes width small enough to accommod
ate all axes decorations
```

plt.tight_layout()



Conclusion

In conclusion, the analysis conducted on the Chinese currency's exchange rates, particularly during different presidential administrations, suggests a potential correlation between political events and fluctuations in the currency. The examination of the Bush, Obama, and Trump presidencies reveals periods of both rise and downfall in the Chinese currency. However, it is crucial to note that correlation does not imply causation, and there could be various other factors influencing currency movements, including economic policies, global events, and market dynamics.

While the analysis hints at a possible connection between political changes and currency variations, it is essential to approach such conclusions with caution. The financial markets are complex, and multiple variables contribute to currency movements. Therefore, attributing these fluctuations solely to changes in the presidency may oversimplify the situation.

Further research and a more in-depth analysis, considering additional economic indicators and global factors, would be necessary to draw more definitive conclusions about the relationship between political events and the Chinese currency. The observed patterns could be coincidental, and other external factors may play a more significant role in influencing currency trends.

In []: