

# Real Estate Price Prediction

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# Outline

- Web Crawler
- Grey-Markov Model

# Open Government Data

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加入清單	行政區	土地位置或建物門牌	交易日期	交易總價 (萬元)註1	交易單價 (萬元/坪)註2	單價是否 含車位	建物移轉 面積(坪)	土地移轉 面積(坪)	建物型態註3	屋齡 註4	樓層別 /總樓層	交易種類	備註 事項	歷次 移轉	詳細 資訊
選取	松山區	八德路二段346-350號(雙號)	110/04/09	1,008	112.63	否	8.95	0.79	大樓(11層含以上有電梯)	12	010/014	房地	有	有+	More
選取	松山區	光復北路100巷16-20號(雙號)	110/04/06	2,620	65.11	否	40.24	14.22	公寓(5樓含以下無電梯)		004/004	房地	無	無	More
選取	松山區	八德路四段656-660號(雙號)	110/04/03	1,200	55.63	否	21.57	3.77	華廈(10層含以下有電梯)	43	005/008	房地	有	有+	More
選取	松山區	八德路四段781-785號(單號)	110/04/10	3,188	90.54	否	35.21	3.44	大樓(11層含以上有電梯)	14	005/015	房地	有	有+	More
選取	松山區	延吉街22巷6-10號(單號)	110/04/13	1,530	66.21	否	23.11	4.57	華廈(10層含以下有電梯)	38	003/007	房地	有	無	More
選取	松山區	南京東路四段186-190號(雙號)	110/04/09	2,303	72.58	否	31.73	2.7	大樓(11層含以上有電梯)	37	008/014	房地	有	有+	More
選取	松山區	民生東路五段260巷21-25號(單號)	110/04/13	3,300	58.82	否	56.1	11.02	華廈(10層含以下有電梯)	40	004/007	房地	有	無	More
選取	松山區	八德路四段396-400號(單號)	110/04/13	1,245	58.78	否	21.18	6.88	公寓(5樓含以下無電梯)		004/004	房地	無	無	More
選取	松山區	八德路三段12巷20弄11-15號(單號)	110/04/07	3,400	78.59	否	43.26	10.49	華廈(10層含以下有電梯)	48	002/006	房地	有	無	More
選取	松山區	三民路29巷6-10號(單號)	110/04/02	4,868	77.43	是	62.87	13.38	大樓(11層含以上有電梯)	16	013/013	房地車	無	無	More

https://cloud.land.gov.taipei/ImmPrice/TruePriceA.aspx

# Scrapy

```
67 def parse(self, response, **kwargs):
68     driver = self.driver
69     driver.get(response.url)
70
71     driver.execute_script("document.getElementById('tab1').style.display = 'none';")
72     driver.execute_script("document.getElementById('tab2').style.display = 'block';")
73
74     for year in range(self.start_year, self.end_year + 1): # every year
75         fr = self.start_month if year == self.start_year else 1
76         to = self.end_month if year == self.end_year else 12
77         for month in range(fr, to + 1): # every month
78             for district_no in range(self.start_district_no, self.end_district_no + 1): # every district
79                 driver.execute_script(f"document.getElementById('{choose_district_id}{district_no}').checked = true;")
80                 self._select_time_range(year, month)
81                 self.driver.execute_script('arguments[0].click();', self.wait.until(EC.element_to_be_clickable((By.ID, search_btn_id))))
82
83                 self.wait.until(EC.visibility_of_element_located((By.ID, table_id)))
84                 district_name = district_list[district_no] + '區'
85                 self.wait.until(EC.text_to_be_present_in_element((By.CSS_SELECTOR, f"#{table_id} > tbody > tr:nth-child(3) > td:nth-child(2)"), district_name))
86
87                 total_rows = int(driver.find_element(By.ID, total_rows_id).text)
88                 for page in range(1, math.ceil(total_rows / 10) + 1): # every page
89                     row_count = len(driver.find_elements(By.CSS_SELECTOR, f"#{table_id} > tbody > tr"))
90
91                     for row_no in range(3, row_count + 1): # every row in the table
92                         if (year, month, district_name, page, row_no - 2) not in ignore_list: # click more info might trigger error
93                             table_td_list = driver.find_elements(By.CSS_SELECTOR, f"#{table_id} > tbody > tr:nth-child({row_no}) td")
94                             yield self._list_to_item(table_td_list, district_name)
95
96                     if page != math.ceil(total_rows / 10):
97                         self._swap_table_page('...' if page % 10 == 0 else page + 1)
98
99                 if total_rows > 10: # swap to first page before starting new district
100                     self._swap_table_page('第一頁' if total_rows > 100 else 1)
101                 driver.execute_script(f"document.getElementById('{choose_district_id}{district_no}').checked = false;")
102
```

# Selenium

1 2 3 4 5 6 7 8 9 10 ... 最末頁															
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click

https://cloud.land.gov.taipei/ImmPrice/TruePriceA.aspx

# Distance to Nearest MRT Exit

data from Taipei MRT

```
75 @staticmethod
76 def _trans_addr_to_coord(addr: str) -> Tuple[float, float]:
77     chrome_options = webdriver.ChromeOptions()
78     chrome_options.add_argument('--headless')
79     chrome_options.add_argument('--disable-gpu')
80     driver = webdriver.Chrome(chrome_options=chrome_options)
81     url = requote_uri(f"https://map.tgos.tw/TGOSimpleViewer/Web/Map/TGOSimpleViewer_Map.aspx?addr={addr}")
82     driver.get(url)
83
84     try:
85         WebDriverWait(driver, 30, 0.01).until(lambda url_change: driver.current_url != url)
86         info = driver.current_url
87         building_lon = float(info[info.index('CX=') + 3:info.index('CY=') - 1])
88         building_lat = float(info[info.index('CY=') + 3:info.index('L=') - 1])
89     except (TimeoutException, ValueError):
90         driver.close()
91         return -1.0, -1.0
92
93     driver.close()
94     return building_lon, building_lat
95
96 def _get_dist_to_mrt(self, lon: float, lat: float) -> Tuple[float, float]:
97     shortest_distance = float('inf')
98     shortest_exit_id = -1
99     for row in self.mrt_exit_coordinate:
100         lon1, lat1, lon2, lat2 = map(radians, [lon, lat, row[1], row[2]])
101         a = sin((lat2 - lat1) / 2) ** 2 + cos(lat1) * cos(lat2) * sin((lon2 - lon1) / 2) ** 2
102         c = 2 * asin(sqrt(a))
103         r = 6371
104         distance = c * r * 1000
105         if distance < shortest_distance:
106             shortest_exit_id = row[0]
107             shortest_distance = distance
108     return shortest_exit_id, shortest_distance
109
110
```

	exit_id	exit_name	exit_no	longitude	latitude
1	1	頂埔站出口1	1	121.418	24.9593
2	2	頂埔站出口2	2	121.419	24.9593
3	3	頂埔站出口3	3	121.42	24.9596
4	4	頂埔站出口4	4	121.42	24.9604
5	5	松山機場站出口1	1	121.552	25.0637
6	6	松山機場站出口2	2	121.551	25.0637
7	7	松山機場站出口3	3	121.552	25.0629
8	8	中山國中站出口	0	121.544	25.0609
9	9	忠孝復興站出口1	1	121.543	25.0418
10	10	忠孝復興站出口2	2	121.543	25.0414
11	11	忠孝復興站出口3	3	121.545	25.0415
12	12	忠孝復興站出口4	4	121.545	25.0418
13	13	忠孝復興站出口5	5	121.544	25.042
14	14	大安站出口1	1	121.542	25.0335
15	15	大安站出口2	2	121.542	25.0335
16	16	大安站出口3	3	121.542	25.0332
17	17	大安站出口4	4	121.544	25.0331
18	18	大安站出口5	5	121.544	25.0329
19	19	大安站出口6	6	121.544	25.034
20	20	科技大樓站出口	0	121.544	25.0262
21	21	六張犁站出口	0	121.553	25.0239
22	22	麟光站出口	0	121.559	25.0186
23	23	辛亥站出口	0	121.557	25.0051
24	24	萬芳醫院站出口	0	121.558	24.9994
25	25	萬芳社區站出口	0	121.568	24.9986
26	26	木柵站出口	0	121.573	24.9982
27	27	動物園站出口1	1	121.58	24.9982
28	28	動物園站出口2	2	121.579	24.9979
29	29	大直站出口1	1	121.547	25.0801
30	30	大直站出口2	2	121.547	25.0796
31	31	大直站出口3	3	121.547	25.0793
32	32	劍南路站出口1	1	121.555	25.0852

# Data Preview

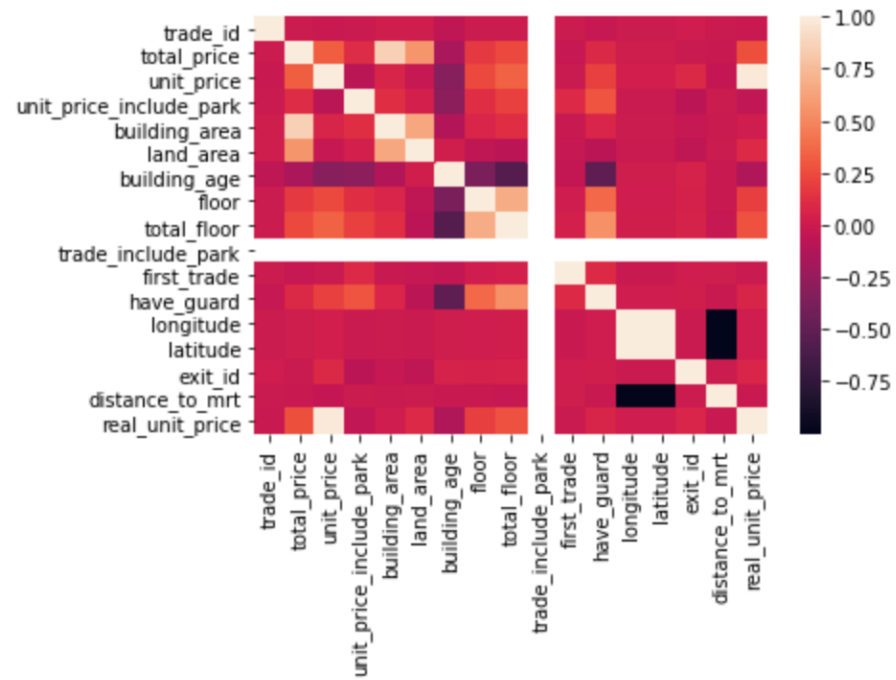
- 2018.01.01 – 2020.12.31
- 60368 rows

	51838	51839	51840	51841	51842	51843	51844	51845
🏠 trade_id	51874	51875	51876	51877	51878	51879	51880	51881
🏠 district	中山區	中山區	中山區	中山區	中山區	中山區	中山區	中山區
🏠 address	南京東路三段194巷1	遼寧街105巷3	新生北路三段88巷	林森北路415號	南京東路三段91號	新生北路一段106號	復興北路446號	松江路184巷26弄8
🏠 trade_date	2018-07-25	2018-07-25	2018-07-21	2018-07-30	2018-07-23	2018-07-30	2018-07-24	2018-07-23
🏠 total_price	79030000	68500000	10860000	6150000	24000000	8300000	10250000	23000000
🏠 unit_price	<null>	<null>	453100	371800	600900	537900	761500	677900
🏠 unit_price_include_park	0	0	0	0	0	0	1	1
🏠 building_area	37.54	34.92	23.97	16.54	39.94	15.43	13.46	33.93
🏠 land_area	42.09	30.55	8.92	1.55	4.35	2.14	2.27	4.6
🏠 building_type	透天厝	透天厝	公寓(5樓含以下無	大樓(11層含以上有	大樓(11層含以上有	大樓(11層含以上有	大樓(11層含以上有	華廈(10層含以下有
🏠 building_age	<null>	<null>	52	34	46	41	10	10
🏠 floor	<null>	<null>	2	9	4	11	3	7
🏠 total_floor	<null>	<null>	4	13	11	12	12	9
🏠 trade_include_park	0	0	0	0	0	0	0	0
🏠 remark	<null>	<null>	陽台外推	<null>	<null>	含增建或未登記建物	含增建或未登記...	含增建或未登記建物
🏠 first_trade	0	0	1	1	0	1	1	1
🏠 building_layout	6房2廳2衛 有隔間	6房2廳2衛 有隔間	2房1廳1衛 有隔間	1房1廳1衛 有隔間	0房0廳0衛 無隔間	1房1廳1衛 有隔間	1房1廳1衛 有隔間	1房1廳1衛 有隔間
🏠 have_guard	0	0	0	0	1	1	1	1
🏠 land_type	第三種住宅區。	第三種住宅區。	第肆種商業區(依...	第肆種商業區(依...	第三種住宅區。	第肆種商業區(依...	第三之一種住宅區，	第貳種商業區(依...
🏠 building_material	加強磚造	加強磚造	鋼筋混凝土造	鋼筋混凝土造	鋼筋混凝土造	鋼筋混凝土造	鋼筋混凝土造	鋼筋混凝土造
🏠 longitude	121.543	121.543	121.527	121.526	121.539	121.528	121.544	121.531
🏠 latitude	25.0513	25.0512	25.0678	25.0598	25.0522	25.0494	25.064	25.0566
🚶 exit_id	329	329	299	297	327	347	8	300
🏠 distance_to_mrt	40.3683	58.0452	557.765	304.493	309.624	452.804	342.174	304.735
🏠 create_time	2021-06-21 03...	2021-06-2...	2021-06-21 ...	2021-06-21 0...	2021-06-21 0...	2021-06-21 0...	2021-06-21 0...	2021-06-21 0...

# Data Preview

```
In [15]: import seaborn as sns
```

```
corr = df.corr()
sns.heatmap(corr)
plt.show()
```



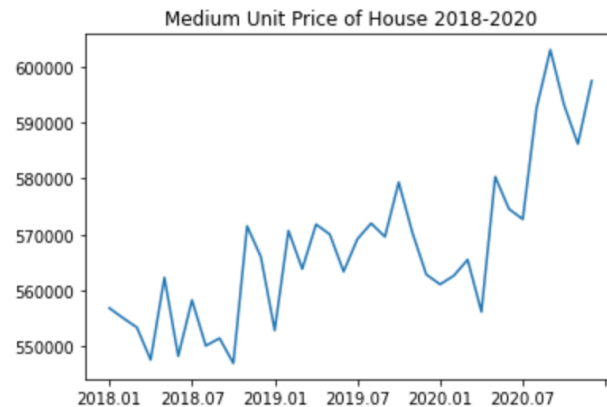


# Medium Price for Every Month

```
In [2]: db = MySQLdb.connect(host= , user= , passwd= , db= )
        cursor = db.cursor()
        start = '2018'
        end = '2020'
        df = pd.read_sql(f"SELECT * FROM trade_price_raw where trade_date between '{start}-01-01' and '{end}-12-31'", con=db)
        time_list = [dt.datetime.strptime(x, '%Y.%m') for x in pd.date_range(start=start+'0101', end=end+'1231', freq='MS')]
```

```
In [4]: df['trade_date'] = pd.to_datetime(df['trade_date'])
        df['real_unit_price'] = df['total_price'] / df['building_area']
        gb = df.groupby(df.trade_date.dt.to_period('M'))
        raw = list(gb['real_unit_price'].median())

        plt.plot(time_list, raw)
        plt.gca().xaxis.set_major_locator(mticker.MultipleLocator(6))
        plt.title(f'Medium Unit Price of House {start}-{end}')
        plt.show()
```



# Grey Model(GM) for time series prediction

Ex: Have a data series: [1, 2, 3]

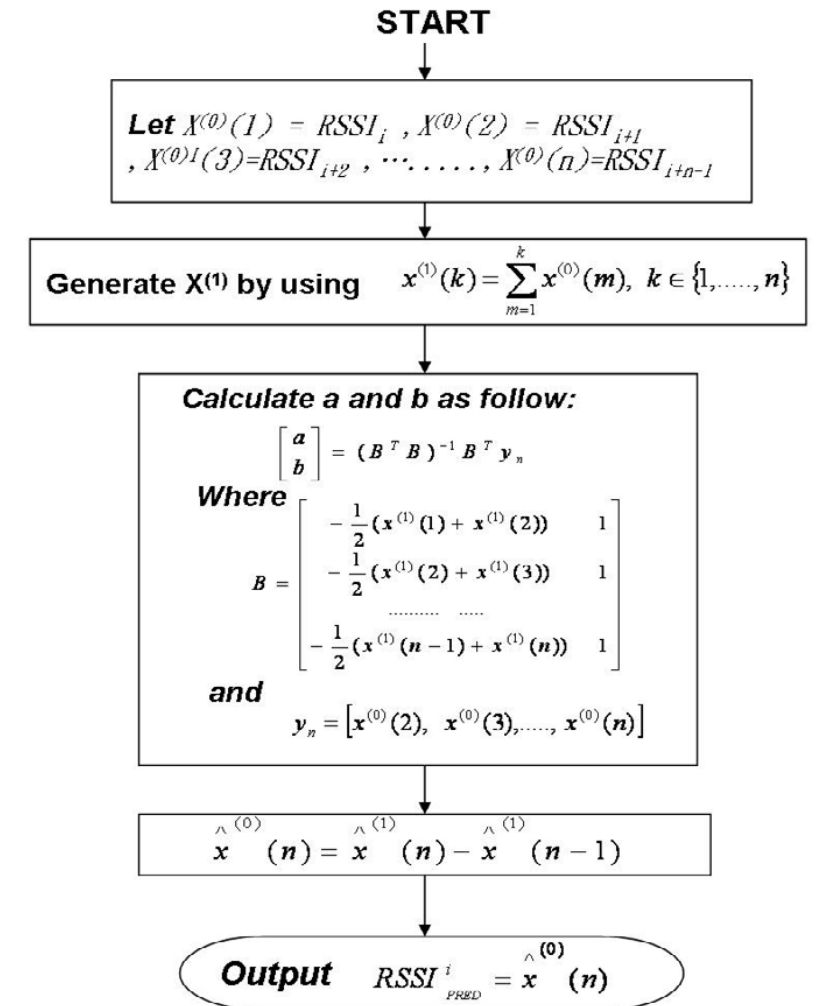
$$y = [1, 3, 6]^T$$

$$-\frac{1}{2}(1+3) \quad -\frac{1}{2}(3+6)$$

$$B = \begin{bmatrix} -2 & 1 \\ 4.5 & 1 \end{bmatrix}$$

$$\begin{bmatrix} a \\ b \end{bmatrix}^T = (B^T B)^{-1} B^T y$$

$$\text{Finally, get } x^{(1)}(k) = (x^{(0)}(1) - \frac{b}{a})e^{-a(k-1)} + \frac{b}{a}$$



# Grey Model(GM) for time series prediction

Ex: Have a data series: [1, 2, 3]

```
21 y_cum_sum = np.cumsum(raw_data)
```

$$y = [1, 3, 6]^T$$

```
23 for i in range(len(raw_data) - 1): # (-1/2)(X(1)(n-1) + X(1)(n))
24     b.append([-0.5 * (y_cum_sum[i] + y_cum_sum[i + 1]), 1])
```

$$B = \begin{bmatrix} -2 & 1 \\ 4.5 & 1 \end{bmatrix}$$

$$\begin{bmatrix} a \\ b \end{bmatrix}^T = (B^T B)^{-1} B^T y$$

```
26 # ((B^T)B)^(-1) * (B^T) * y
27 b = np.array(b)
28 b_t = np.transpose(b)
29 output = np.linalg.inv(np.dot(b_t, b))
30 output = np.dot(output, b_t)
31 y = np.transpose(raw_data[1:])
32 output = np.dot(output, y)
33 output = np.transpose(output)
34
35 return output[0], output[1]
```

Finally, get  $x^{(1)}(k) = (x^{(0)}(1) - \frac{b}{a})e^{-a(k-1)} + \frac{b}{a}$

```
6 def fit(raw_data) -> List[float]:
7     a, b = _get_coefficient(raw_data)
8
9     output_list = []
10    for i in range(len(raw_data)): # X(1)(k) = (X(0)(1) - b/a) * e^(-a(k-1)) + (b/a)
11        output_list.append((raw_data[0] - b / a) * np.exp(-a * i) + b / a)
12
13    output_list = [el - output_list[i - 1] if i > 0 else el for i, el in enumerate(output_list)] # inverse of cumulative sum
14    return output_list
```

# Grey Model(GM) for time series prediction

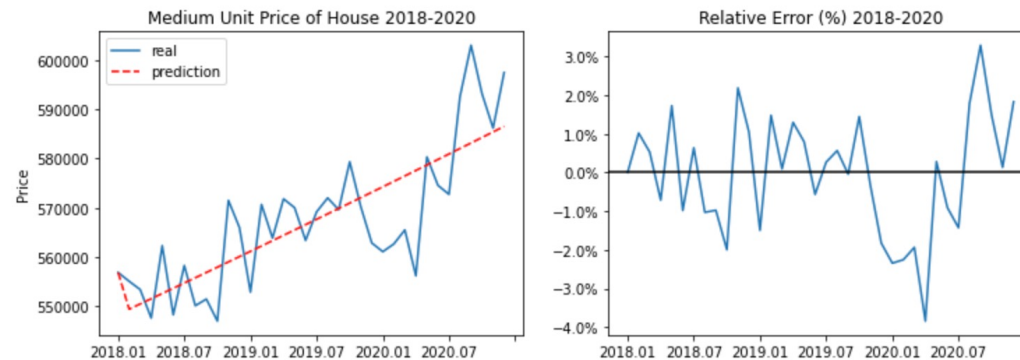
```
In [5]: pred = grey_model.fit(raw)
relative_err_list = [(i - j) / i for i, j in zip(raw, pred)]

plt.figure(figsize=(12, 4))

plt.subplot(1, 2, 1)
plt.plot(time_list, raw)
plt.plot(time_list, pred, color = 'red', linestyle = '--')
plt.ylabel('Price')
plt.gca().xaxis.set_major_locator(mticker.MultipleLocator(6))
plt.title(f'Medium Unit Price of House {start}-{end}')
plt.legend(['real', 'prediction'])

plt.subplot(1, 2, 2)
plt.plot(time_list, relative_err_list)
plt.axhline(y=0, ls='-', color='black')
plt.gca().xaxis.set_major_locator(mticker.MultipleLocator(6))
plt.gca().yaxis.set_major_formatter(mticker.PercentFormatter(xmax=1, decimals=1))
plt.title(f'Relative Error (%) {start}-{end}')

plt.show()
```



# Validate My Grey Model(GM)

```

38 def get_precision(raw_data: List[float], pred_data: List[float]) -> float:
39     raw_corr = 0
40     for i in range(1, len(raw_data) - 1):
41         raw_corr += raw_data[i] - raw_data[0]
42     raw_corr += (raw_data[-1] - raw_data[0]) / 2
43     raw_corr = np.abs(raw_corr)
44
45     pred_corr = 0
46     for i in range(1, len(pred_data) - 1, 1):
47         pred_corr += pred_data[i] - pred_data[0]
48     pred_corr += (pred_data[-1] - pred_data[0]) / 2
49     pred_corr = np.abs(pred_corr)
50
51     sub_corr = 0
52     for i in range(1, len(pred_data) - 1, 1):
53         sub_corr += (raw_data[i] - raw_data[0]) - (pred_data[i] - pred_data[0])
54     sub_corr += ((raw_data[-1] - raw_data[0]) - (pred_data[-1] - pred_data[0])) / 2
55     sub_corr = np.abs(sub_corr)
56
57     return (1 + raw_corr + pred_corr) / (1 + raw_corr + pred_corr + sub_corr)

```

$$|S| = \left| \sum_{k=2}^{n-1} (x(k) - x(1)) + \frac{1}{2} (x(n) - x(1)) \right|$$

$$|\hat{S}| = \left| \sum_{k=2}^{n-1} (\hat{x}(k) - \hat{x}(1)) + \frac{1}{2} (\hat{x}(n) - \hat{x}(1)) \right|$$

$$|\hat{S} - S| = \left| \sum_{k=2}^{n-1} [(x(k) - x(1)) - (\hat{x}(k) - \hat{x}(1))] + \frac{1}{2} [(x(n) - x(1)) - (\hat{x}(n) - \hat{x}(1))] \right|$$

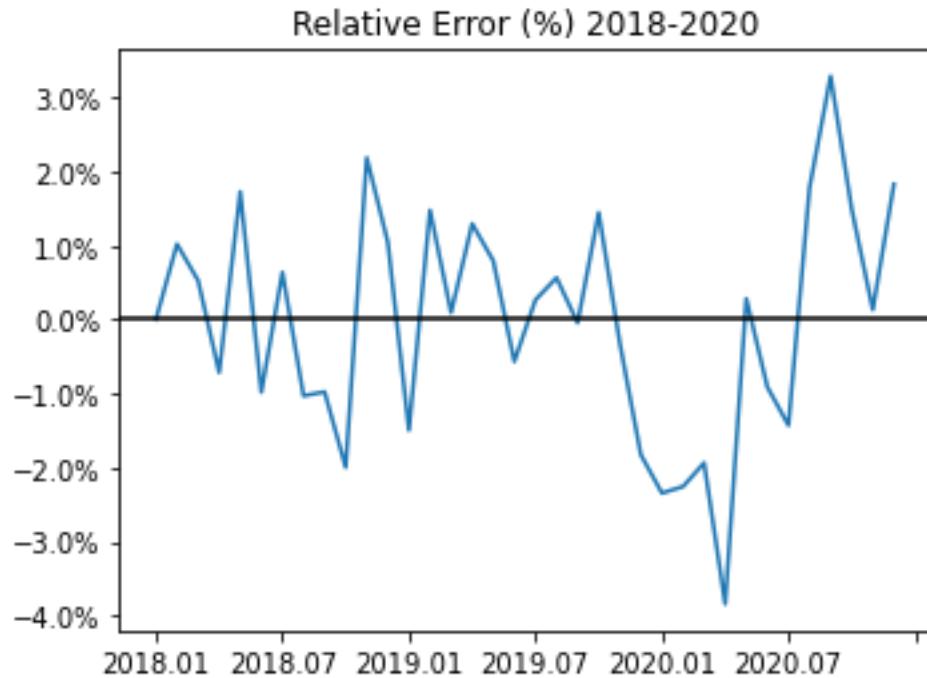
$$\varepsilon = \frac{1 + |S| + |\hat{S}|}{1 + |S| + |\hat{S}| + |\hat{S} - S|}$$

1(best)	> 0.9
2	> 0.8
3	> 0.7
4	> 0.6

In [5]: grey\_model.get\_precision(raw, pred)

Out[5]: 0.9926133493297309

# Markov Chain for Status Transform Prediction



relative\_err\_list

```
[4.411080748789029e-14,  
0.010189830765070271,  
0.005272155327967224,  
-0.007148685226892636,  
0.01726793203056345,  
-0.009827616273494664,  
0.0063637379045535965,  
-0.010312650533956894,  
-0.00978130742130919,  
-0.019991591186914154,  
0.021861328604199134,  
0.010439853727232697,  
-0.014971622996744918,  
0.014776632287169633,  
0.0009404103565779892,  
0.012943907832977405,  
0.007940105410239607,  
-0.005677805229021457,  
0.0026566345010391574,  
0.0056447302238749894,  
-0.00045300852457860107,  
0.014461054628911563,  
-0.0031504435686536556,  
-0.0182880057706567,  
-0.023462355530152627,  
-0.022582370697113965,  
-0.019384608076519334,  
-0.03845239113679737,  
0.002810579079834714,  
-0.009127560724893017,  
-0.014304775597038773,  
0.017824656826236452,  
0.032900256808440234,  
0.014914457774912094,  
0.0013213706187519137,  
0.0182795438506177]
```

# Markov Chain for Status Transform Prediction

relative\_err\_list

Min: -3.85%  
Max: 3.29%

```
[4.411080748789029e-14,  
0.010189830765070271,  
0.005272155327967224,  
-0.007148685226892636,  
0.01726793203056345,  
-0.009827616273494664,  
0.0063637379045535965,  
-0.010312650533956894,  
-0.00978130742130919,  
-0.019991591186914154,  
0.021861328604199134,  
0.010439853727232697,  
-0.014971622996744918,  
0.014776632287169633,  
0.0009404103565779892,  
0.012943907832977405,  
0.007940105410239607,  
-0.005677805229021457,  
0.0026566345010391574,  
0.0056447302238749894,  
-0.00045300852457860107,  
0.014461054628911563,  
-0.0031504435686536556,  
-0.0182880057706567,  
-0.023462355530152627,  
-0.022582370697113965,  
-0.019384608076519334,  
-0.03845239113679737,  
0.002810579079834714,  
-0.009127560724893017,  
-0.014304775597038773,  
0.017824656826236452,  
0.032900256808440234,  
0.014914457774912094,  
0.0013213706187519137,  
0.0182795438506177]
```

Slice into 7 intervals:

Status1: -3.85% ~ -2.98%

Status2: -2.98% ~ -2.12%

Status3: -2.12% ~ -1.26%

Status4: -1.26% ~ -0.04%

Status5: -0.04% ~ 0.05%

Status6: 0.05% ~ 1.32%

Status7: 1.32% ~ 2.19%

```
In [11]: model.interval_list
```

```
Out[11]: [Interval(-0.03845239113679737, -0.029836145459512152, lower_closed=True, upper_closed=True),  
Interval(-0.029836145459512152, -0.02121989978222694, lower_closed=True, upper_closed=True),  
Interval(-0.02121989978222694, -0.012603654104941728, lower_closed=True, upper_closed=True),  
Interval(-0.012603654104941728, -0.003987408427656512, lower_closed=True, upper_closed=True),  
Interval(-0.003987408427656512, 0.004628837249628703, lower_closed=True, upper_closed=True),  
Interval(0.004628837249628703, 0.013245082926913912, lower_closed=True, upper_closed=True),  
Interval(0.013245082926913912, 0.021861328604199134, lower_closed=True, upper_closed=True)]
```

# Markov Chain for Status Transform Prediction

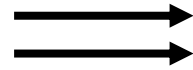
	Status1	Status2	Status3	Status4	Status5	Status6	Status7
Status1	0	0	0	0	0	0	0
Status2	0	0	0	0	0	0	0
Status3	0	0	0	0	0	0	0
Status4	0	0	0	0	0	0	0
Status5	0	0	0	0	0	0	0
Status6	0	0	0	0	0	0	0
Status7	0	0	0	0	0	0	0



# Markov Chain for Status Transform Prediction

relative\_err\_list

```
[4.411080748789029e-14,  
0.010189830765070271,  
0.005272155327967224,  
-0.007148685226892636,  
0.01726793203056345,  
-0.009827616273494664,  
0.0063637379045535965,  
-0.010312650533956894,  
-0.00978130742130919,  
-0.019991591186914154,  
0.021861328604199134,  
0.010439853727232697,  
-0.014971622996744918,  
0.014776632287169633,  
0.0009404103565779892,  
0.012943907832977405,  
0.007940105410239607,  
-0.005677805229021457,  
0.0026566345010391574,  
0.0056447302238749894,  
-0.00045300852457860107,  
0.014461054628911563,  
-0.0031504435686536556,  
-0.0182880057706567,  
-0.023462355530152627,  
-0.022582370697113965,  
-0.019384608076519334,  
-0.03845239113679737,  
0.002810579079834714,  
-0.009127560724893017,  
-0.014304775597038773,  
0.017824656826236452,  
0.032900256808440234,  
0.01491445774912094,  
0.0013213706187519137,  
0.0182795438506177]
```



Status6: vector = [0 0 0 0 0 1 0]

Status7: vector = [0 0 0 0 0 0 1]



Status6 to Status7

Status7 to ...

```
In [12]: model.trans_prob
```

```
Out[12]: array([[0.      , 0.      , 0.      , 0.      , 0.      ,  
0.      , 0.      ],  
[0.      , 0.5      , 0.5      , 0.      , 0.      ,  
0.      , 0.      ],  
[0.25   , 0.25   , 0.      , 0.      , 0.      ,  
0.      , 0.5      ],  
[0.      , 0.      , 0.2      , 0.2      , 0.2      ,  
0.2      , 0.2      ],  
[0.      , 0.      , 0.16666667, 0.16666667, 0.      ,  
0.5      , 0.16666667],  
[0.      , 0.      , 0.14285714, 0.42857143, 0.14285714,  
0.28571429, 0.      ],  
[0.      , 0.      , 0.      , 0.25   , 0.5      ,  
0.25   , 0.      ]])
```

Slice into 7 intervals:

Status1: -3.85% ~ -2.98%

Status2: -2.98% ~ -2.12%

Status3: -2.12% ~ -1.26%

Status4: -1.26% ~ -0.04%

Status5: -0.04% ~ 0.05%

Status6: 0.05% ~ 1.32%

Status7: 1.32% ~ 2.19%

# Markov Chain for Status Transform Prediction

2020.06	[0 0 0 1 0 0 0]	4 steps to 2020.10	$[0\ 0\ 0\ 1\ 0\ 0\ 0] * p(4)$
2020.07	[0 0 1 0 0 0 0]	3 steps to 2020.10	$[0\ 0\ 1\ 0\ 0\ 0\ 0] * p(3)$
2020.08	[0 0 0 0 0 0 1]	2 steps to 2020.10	$[0\ 0\ 0\ 0\ 0\ 0\ 1] * p(2)$
2020.09	[0 0 0 0 0 0 1]	1 steps to 2020.10	$[0\ 0\ 0\ 0\ 0\ 0\ 1] * p(1)$

P =

```
model.trans_prob
array([[0.      , 0.      , 0.      , 0.      , 0.      ,
        0.      , 0.      ],
       [0.      , 0.5     , 0.5     , 0.      , 0.      ,
        0.      , 0.      ],
       [0.25    , 0.25    , 0.      , 0.      , 0.      ,
        0.      , 0.5     ],
       [0.      , 0.      , 0.2     , 0.2     , 0.2     ,
        0.2     , 0.2     ],
       [0.      , 0.      , 0.16666667, 0.16666667, 0.      ,
        0.5     , 0.16666667],
       [0.      , 0.      , 0.14285714, 0.42857143, 0.14285714,
        0.28571429, 0.      ],
       [0.      , 0.      , 0.      , 0.25    , 0.5     ,
        0.25    , 0.      ]])
```

Plus these 4 vector and calculate the most possible status.

```
50 def validate_test_data(self):
51     prob_vector_list = np.zeros((1, self.count))
52     p = self.trans_prob
53     for i in range(4):
54         prob_dist = np.dot(self.__to_vector(self.relative_err_list[-i-1]), p)
55         prob_vector_list += prob_dist
56         p = np.dot(p, self.trans_prob)
57     return prob_vector_list / prob_vector_list.sum(axis=1) * 100
58
```

# Markov Chain for Status Transform Prediction

```
7 class GreyMarkovModel:
8     def __init__(self, relative_err_list: List[float], count: int):
9         self.relative_err_list = relative_err_list
10        self.max_relative_err = max(relative_err_list)
11        self.min_relative_err = min(relative_err_list)
12        self.count = count
13        self.interval_val = (self.max_relative_err - self.min_relative_err) / self.count
14        self.interval_list = []
15
16        self.expected_val_list = []
17        for i in range(count):
18            self.expected_val_list.append(min(relative_err_list) + (0.5 + i) * self.interval_val)
19
20        self.trans_prob = np.zeros((self.count, self.count))
21        self.__init_status_trans_prob()
22
23    def __init_status_trans_prob(self) -> None:
24        for i in range(self.count):
25            if i != self.count - 1:
26                self.interval_list.append(Interval(self.min_relative_err + i * self.interval_val, self.min_relative_err + (i + 1) * self.interval_val))
27            else:
28                self.interval_list.append(Interval(self.min_relative_err + i * self.interval_val, self.max_relative_err))
29
30        prev = None
31        for i, err_val in enumerate(self.relative_err_list):
32            matrix = self.__to_vector(err_val)
33            if prev:
34                self.trans_prob[prev] += matrix
35            prev = matrix.index(1)
36
37        sum_list = np.sum(self.trans_prob, axis=1)
38        for row_i, row in enumerate(self.trans_prob):
39            if sum_list[row_i] != 0:
40                for col_i, val in enumerate(row):
41                    self.trans_prob[row_i, col_i] /= sum_list[row_i]
```

```
In [8]: sum = 0
        for i in range(6):
            model = GreyMarkovModel(relative_err_list[:-i-1], 7)
            status_num = np.argmax(model.validate_test_data(), axis=1)[0]
            mc_pred = pred[-i-1] / (1 - model.expected_val_list[status_num])
            err = (raw[-i-1] - mc_pred) / raw[-i-1] * 100
            sum += err
            print(err)
        print('mean: ' + str(sum / 6) + '%')
```

0.06811352398320765  
-0.6141349057096572  
0.7553313717076561  
3.2589994478337347  
0.027630102950403166  
-0.5959804585459864  
mean: 0.4833265137032263%

Thank You