

EIOPA RISK-FREE CURVE APRIL-23 RECALCULATION

The risk-free curve is one of the principal inputs into an economic scenario generator. This notebook recalculates the risk-free curve using the parameters that are claimed to be used. The European Insurance and Occupational Pensions Authority (EIOPA) publishes their own yield curve prediction. To do this they use the Smith & Wilson algorithm.

Summary

The goal of this test is to replicate the EIOPA yield curve. This test will use the methodology that EIOPA claims it is using and the calibration vector that they publish. If the test is passed, the user can be more confident, that EIOPA risk free rate (RFR) curve was generated using the described methodology/calibration and that the process was implemented correctly.

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Note on Smith & Wilson algorithm

To replicate the calculations, this example uses a modified Smith&Wilson implementation (The original implementation is available on [GitHub](#):

- [Python](#)
- [Matlab](#)
- [JavaScript](#)

Limitations of the implementation

Current implementation only looks at a single currency and with/without Volatility Adjustment (VA). The day count convention assumes that each year has the same number of days.

Data requirements

This script contains the EIOPA risk-free rate publication for April 2023. The publication can be found on the [EIOPA RFR website](#).

The observed maturities `M_Obs` and the calibrated vector `Qb` can be found in the Excel sheet `EIOPA_RFR_20230430_Qb_SW.xlsx`.

The target maturities (`T_Obs`), the additional parameters (`UFR` and `alpha`), and the given curve can be found in the Excel `EIOPA_RFR_20230430_Term_Structures.xlsx`, sheet `RFR_spot_no_VA` if the test looks at the curve without the Volatility Adjustment and the sheet `RFR_spot_with_VA` if the test looks at the curve with the Volatility Adjustment.

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Success criteria

The following success criteria is defined:

- Maximum difference between the calculated curve and the one provided by EIOPA is less than 0.1 bps
- Average difference between the calculated curve and the one provided by EIOPA is less than 0.05 bps

In [63]:

```
test_statistics_max_diff_in_bps = 0.1
test_statistics_average_diff_in_bps = 0.05
```

The success function is called at the end of the test to confirm if the success criteria have been met.

```
In [64]:  
def SuccessTest(TestStatistics, threshold_max, threshold_mean):  
    out1 = False  
    out2 = False  
    if max(TestStatistics) < threshold_max:  
        print("Test passed")  
        out1 = True  
    else:  
        print("Test failed")  
  
    if np.mean(TestStatistics) < threshold_mean:  
        print("Test passed")  
        out2 = True  
    else:  
        print("Test failed")  
    return [out1, out2]
```

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External dependencies

This implementation uses three well established Python packages widely used in the financial industry. Pandas (<https://pandas.pydata.org/docs/>), Numpy (<https://numpy.org/doc/>), and Matplotlib (<https://matplotlib.org/stable/index.html>)

```
In [65]:  
import numpy as np  
import pandas as pd  
import matplotlib.pyplot as plt  
import matplotlib.ticker as mtick  
%matplotlib notebook  
pd.options.display.max_rows = 150
```

Importing data

```
In [66]:  
selected_param_file = 'Param_VA.csv'  
selected_curves_file = 'Curves_VA.csv'  
  
#selected_param_file = 'Param_no_VA.csv'  
#selected_curves_file = 'Curves_no_VA.csv'
```

```
In [67]:  
param_raw = pd.read_csv(selected_param_file, sep=',', index_col=0)
```

Parameter input

Parameters sheet

In [68]: `param_raw.head()`

Out[68]:

	Euro_Maturities	Euro_Values	Austria_Maturities	Austria_Values	Belgium_Maturities	Be
Country						
Coupon_freq	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000
LLP	20.000000	20.000000	20.000000	20.000000	20.000000	20.000000
Convergence	40.000000	40.000000	40.000000	40.000000	40.000000	40.000000
UFR	3.450000	3.450000	3.450000	3.450000	3.450000	3.450000
alpha	0.111906	0.111906	0.111906	0.111906	0.111906	0.111906

5 rows × 106 columns

The country selected is:

In [69]: `country = "Italy"`

In [70]:

```
maturities_country_raw = param_raw.loc[:,country+"_Maturities"].iloc[6:]
param_country_raw = param_raw.loc[:,country + "_Values"].iloc[6:]
extra_param = param_raw.loc[:,country + "_Values"].iloc[:6]
```

Extra parameters

Smith-Wilson calibration parameters

In [71]: `extra_param`

Out[71]:

Country	
Coupon_freq	1.000000
LLP	20.000000
Convergence	40.000000
UFR	3.450000
alpha	0.111906
CRA	10.000000

Name: Italy_Values, dtype: float64

In [72]: `relevant_positions = pd.notna(maturities_country_raw.values)`

```
In [73]: maturities_country = maturities_country_raw.iloc[relevant_positions]
```

Maturity vector

Vector of maturities used in the calibration

```
In [74]: maturities_country.head(15)
```

```
Out[74]: Country
1      1.0
2      2.0
3      3.0
4      4.0
5      5.0
6      6.0
7      7.0
8      8.0
9      9.0
10     10.0
11     11.0
12     12.0
13     13.0
14     14.0
15     15.0
Name: Italy_Maturities, dtype: float64
```

```
In [75]: Qb = param_country_raw.iloc[relevant_positions]
```

Calibration vector

Vector **Qb** provided as input

```
In [76]: Qb
```

```
Out[76]: Country
1      -8.945909
2       0.541504
3      1.925282
4      1.058779
5      -0.475350
6      -0.387083
7       0.712179
8      -0.383027
9      -0.393004
10     2.715405
11     -3.918373
12     2.701016
13     -0.034363
14     -0.031608
15     -1.692104
16      0.023894
```

```
17    0.051802
18   -0.063782
19    0.361738
20    0.440014
Name: Italv Values, dtype: float64
```

```
In [77]: curve_raw = pd.read_csv(selected_curves_file, sep=',', index_col=0)
```

```
In [78]: curve_country = curve_raw.loc[:,country]
```

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Calibration parameters and calibration vector provided by EIOPA

```
In [79]: # Maturity of observations:
M_Obs = np.transpose(np.array(maturities_country.values))

# Ultimate forward rate ufr represents the rate to which the rate curve will converge
ufr = extra_param.iloc[3]/100

# Convergence speed parameter alpha controls the speed at which the curve converges
alpha = extra_param.iloc[4]

# For which maturities do we want the SW algorithm to calculate the rates. In this case, we want rates for all maturities
M_Target = np.transpose(np.arange(1,151))

# Qb calibration vector published by EIOPA for the curve calibration:
Qb = np.transpose(np.array(Qb.values))
```

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Smith & Wilson calculation functions

In this step, the independent version of the Smith&Wilson algorithm is implemented. To do this, two functions are taken from the publicly available repository and modified to accept the product of Q^*b instead of the calibration vector b .

In [80]:

```

def SWExtrapolate(M_Target, M_Obs, Qb, ufr, alpha):
    # SWEXTRAPOLATE Interpolate or/and extrapolate rates for targeted maturities using a
    # out = SWExtrapolate(M_Target, M_Obs, Qb, ufr, alpha) calculates the rates for maturities
    #
    # Arguments:
    #     M_Target = k x 1 ndarray. Each element represents a bond maturity of interest.
    #     M_Obs = n x 1 ndarray. Observed bond maturities used for calibrating the curve.
    #     Qb = n x 1 ndarray. Calibration vector calculated on observed bonds.
    #     ufr = 1 x 1 floating number. Representing the ultimate forward rate.
    #     Ex. ufr = 0.042
    #     alpha = 1 x 1 floating number. Representing the convergence speed parameter.
    #
    #
    # Returns:
    #     k x 1 ndarray. Represents the targeted rates for a zero-coupon bond. Each rate is
    #
    # For more information see https://www.eiopa.europa.eu/sites/default/files/risk_free_curve.ipynb

def SWHeart(u, v, alpha):
    # SWHEART Calculate the heart of the Wilson function.
    # H = SWHeart(u, v, alpha) calculates the matrix H (Heart of the Wilson
    # function) for maturities specified by vectors u and v. The formula is
    # taken from the EIOPA technical specifications paragraph 132.
    #
    # Arguments:
    #     u = n_1 x 1 vector of maturities. Ex. u = [1; 3]
    #     v = n_2 x 1 vector of maturities. Ex. v = [1; 2; 3; 5]
    #     alpha = 1 x 1 floating number representing the convergence speed parameter.
    #
    # Returns:
    #     n_1 x n_2 matrix representing the Heart of the Wilson function for selected
    #
    # For more information see https://www.eiopa.europa.eu/sites/default/files/risk_free_curve.ipynb

    u_Mat = np.tile(u, [v.size, 1]).transpose()
    v_Mat = np.tile(v, [u.size, 1])
    return 0.5 * (alpha * (u_Mat + v_Mat) + np.exp(-alpha * (u_Mat + v_Mat)) - alpha)

H = SWHeart(M_Target, M_Obs, alpha) # Heart of the Wilson function from paragraph 132
p = np.exp(-np.log(1+ufr)* M_Target) + np.diag(np.exp(-np.log(1+ufr) * M_Target))
return p ** (-1/ M_Target) -1 # Convert obtained prices to rates and return price

```

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Generation of the risk-free curve

The observed maturities, target maturities, and the model parameters provided by EIOPA are used to generate the target curve.

In [81]:

```

r_Target = SWExtrapolate(M_Target,M_Obs, Qb, ufr, alpha)
r_Target = pd.DataFrame(r_Target,columns=['Recalculated rates'])

```

Yield curve calculated

Yield curve calculated using the calibration vector **Qb**

In [82]:

```
r_Target.head(15)
```

Out[82]:

Recalculated rates

0	0.038530
1	0.035419
2	0.033076
3	0.031777
4	0.031117
5	0.030734
6	0.030516
7	0.030446
8	0.030461
9	0.030552
10	0.030700
11	0.030764
12	0.030832
13	0.030851
14	0.030747

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Test 1; Comparison test

Comparison of the calculated yield curve with the yield curve provided by EIOPA. The test is passed if the success criteria is reached.

The provided yield curve can be found in file *EIOPA_RFR_20230430_Term_Structures.xlsx*, sheet *RFR_spot_no_VA* if the test looks at the curve without the Volatility Adjustment and the sheet *RFR_spot_with_VA* if the test looks at the curve with the Volatility Adjustment.

In [83]:

```
target_curve = np.transpose(np.array(curve_country.values))
```

This implementation looks at two kinds of test statistics. The average deviation and the maximum deviation.

The average deviation is defined as:

$$S_{AVERAGE} = \frac{1}{T} \sum_{t=0}^T |r_{EIOPA}(t) - r_{EST}(t)|$$

The maximum deviation is defined as:

$$S_{MAX} = \max_t |r_{EIOPA}(t) - r_{EST}(t)|$$

Where T is the maximum maturity available.

The average difference test is successful if:

$$S_{AVERAGE} < 0.05bps$$

The maximum difference test is successful if:

$$S_{MAX} < 0.1bps$$

```
In [84]: target_curve = pd.DataFrame(target_curve,columns=['Given rates'])
```

EIOPA curve provided

Yield curve provided by EIOPA

```
In [85]: target_curve.head()
```

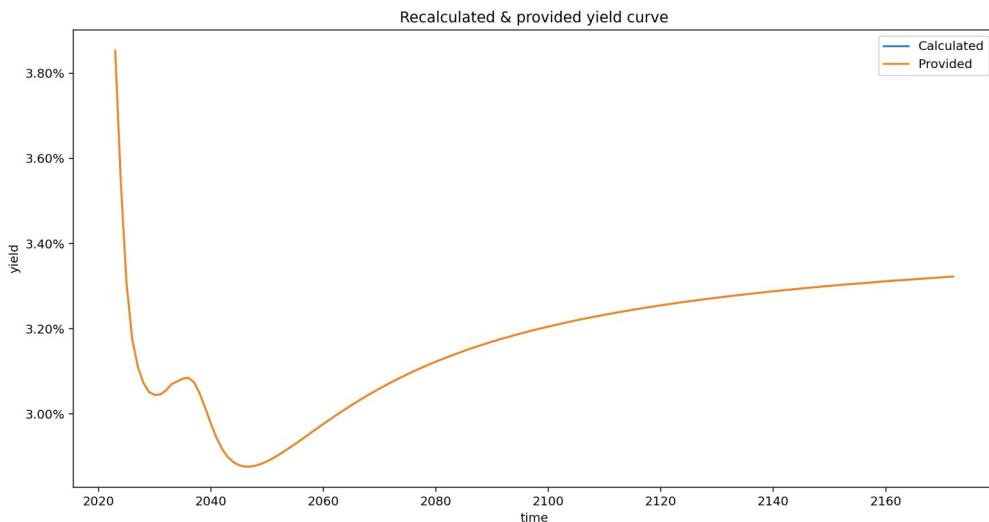
Out[85]: **Given rates**

0	0.03853
1	0.03542
2	0.03308
3	0.03178
4	0.03112

```
In [86]: x_data_label = range(2023,2023+r_Target.shape[0],1)
```

```
In [87]: fig, ax1 = plt.subplots(1,1)
ax1.plot(x_data_label, r_Target.values*100, color='tab:blue',label="Calculated")
ax1.plot(x_data_label, target_curve.values*100, color='tab:orange',label="Provided")

ax1.set_ylabel("yield")
ax1.set_title('Recalculated & provided yield curve')
ax1.set_xlabel("time")
ax1.legend()
ax1.yaxis.set_major_formatter(mtick.PercentFormatter())
fig.set_figwidth(6)
fig.set_figheight(3)
plt.show()
```



```
In [88]: test_statistics_bdp = pd.DataFrame(abs(r_Target.values-target_curve.values)*10000, co
```

EIOPA curve comparison

Absolute difference in bps

```
In [89]: test_statistics_bdp.head()
```

Out[89]: **Abs diff in bps**

0	0.000165
1	0.013756
2	0.040949
3	0.029202

Abs diff in bps[Back to the top](#)

Test 1; Success criteria

The successful application of the success criteria marks the completion of the test.

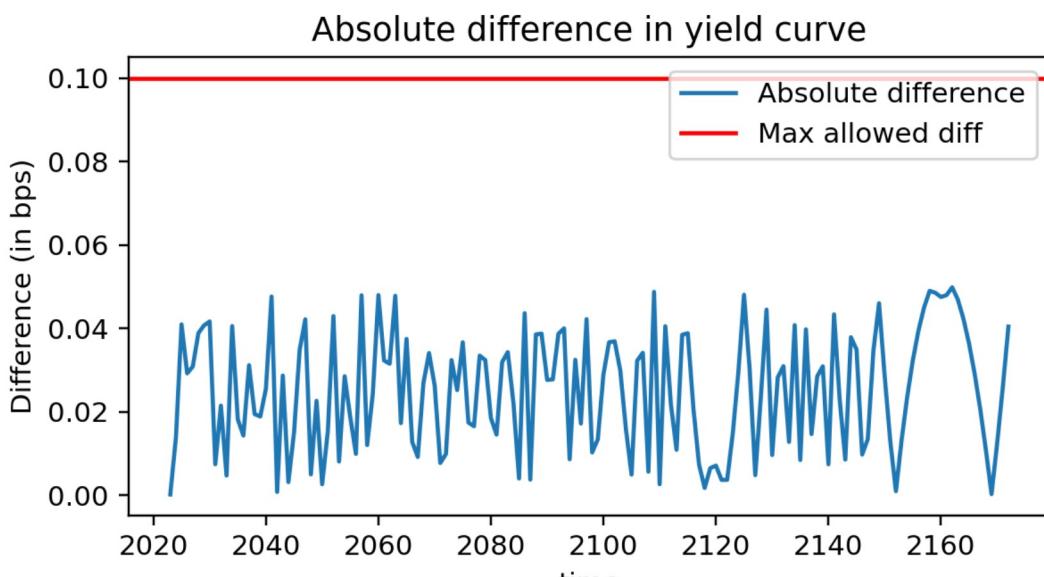
```
In [90]: result1 = SuccessTest(test_statistics_bdp.values, test_statistics_max_diff_in_bps, tolerance=0.01)
```

Test passed
Test passed

```
In [91]: x_data_label = range(2023,2023+r_Target.shape[0],1)
fig, ax1 = plt.subplots(1,1)
ax1.plot(x_data_label, test_statistics_bdp, label= "Absolute difference")
ax1.axhline(y = test_statistics_max_diff_in_bps, color = 'r', linestyle = '--',label='Max allowed diff')

ax1.set_xlabel("time")
ax1.set_ylabel("Difference (in bps)")
ax1.set_title('Absolute difference in yield curve')
ax1.legend()
fig.set_figwidth(6)
fig.set_figheight(3)

plt.show()
```



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Conclusion

This test checks the success criteria on the EIOPA curve generated for April 2023. If the tests are passed, it is likely that the curve was generated using the Smith & Wilson algorithm with the calibration vector that was provided in the file *EIOPA_RFR_20230430_Qb_SW.xlsx* and the parameters displayed in the file *EIOPA_RFR_20230430_Term_Structures.xlsx*.

In [92]:

```
pd.DataFrame(data = [result1], columns = ["Mean test", "Max test"], \
              index= ["Provided vs calculated"])
```

Out[92]:

	Mean test	Max test
Provided vs calculated	True	True

Final yield curve

Full yield curve provided by EIOPA in %

In [93]:

```
(curve_country*100).head(150)
```

Out[93]:

Country	
1	3.853
2	3.542
3	3.308
4	3.178
5	3.112
6	3.073
7	3.052
8	3.045
9	3.046
10	3.055
11	3.070
12	3.076
13	3.083
14	3.085
15	3.075
16	3.049
17	3.014
18	2.978
19	2.944
20	2.918
21	2.899
22	2.887
23	2.880
24	2.877
25	2.877
26	2.879
27	2.883
28	2.889

29	2.896
30	2.904
31	2.912
32	2.921
33	2.930
34	2.939
35	2.948
36	2.958
37	2.967
38	2.977
39	2.986
40	2.995
41	3.004
42	3.012
43	3.021
44	3.029
45	3.037
46	3.045
47	3.052
48	3.060
49	3.067
50	3.074
51	3.081
52	3.087
53	3.094
54	3.100
55	3.106
56	3.112
57	3.117
58	3.123
59	3.128
60	3.133
61	3.138
62	3.143
63	3.148
64	3.153
65	3.157
66	3.161
67	3.166
68	3.170
69	3.174
70	3.178
71	3.181
72	3.185
73	3.189
74	3.192
75	3.196
76	3.199
77	3.202
78	3.205
79	3.208
80	3.211
81	3.214
82	3.217
83	3.220
84	3.223
85	3.225
86	3.228
87	3.230

88	3.233
89	3.235
90	3.238
91	3.240
92	3.242
93	3.245
94	3.247
95	3.249
96	3.251
97	3.253
98	3.255
99	3.257
100	3.259
101	3.261
102	3.263
103	3.265
104	3.266
105	3.268
106	3.270
107	3.271
108	3.273
109	3.275
110	3.276
111	3.278
112	3.279
113	3.281
114	3.282
115	3.284
116	3.285
117	3.287
118	3.288
119	3.289
120	3.291
121	3.292
122	3.293
123	3.295
124	3.296
125	3.297
126	3.298
127	3.300
128	3.301
129	3.302
130	3.303
131	3.304
132	3.305
133	3.306
134	3.307
135	3.308
136	3.309
137	3.311
138	3.312
139	3.313
140	3.314
141	3.314
142	3.315
143	3.316
144	3.317
145	3.318
146	3.319

147 3.320
148 3.321
149 3.322
150 3.323