

ESG FRAMEWORK STOCHASTIC RISK NEUTRAL_OPEN

February 10, 2025

1 STOCHASTIC SCENARIO TESTS

This test checks the consistency of a market-consistent and risk-neutral output from an Economic Scenario Generator (ESG). The goal of this script is to:

- Verify if average performance of the stochastic output is consistent with the deterministic equivalent as implied by the risk-neutral property
- Perform a series of Martingale tests using multiple investment strategies

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Description of the stochastic scenario tests 1. Average of the stochastic deflator is equal to the deterministic deflator 2. Martingale test for a Forward return with 1-year spread strategy using spot rates 3. Martingale test for a Buy-and-hold strategy with spot rates 4. Martingale test for a Buy-and-hold strategy with a total-return index

Limitations of the implementation This implementation has the following limitations:

- The data is assumed to have a monthly granularity
- This implementation only looks at a subset of all scenarios
- Only one currency at a time is checked
- Only one set of deterministic/stochastic scenario pairs are used
- Only one total-return index is used
- Only two spot rates with term 1-and 2-years are considered

The accounting convention used is that all months are assumed to be of equal length and represent exactly 1/12 of a year.

Note that the Extra content sections contain plots for the spot curve with up to the 5-year term.

Success criteria definition The following success criteria is proposed:

- Difference between the average of stochastic deflator curve and the deterministic deflator is never larger than 300bps
- Average difference between the average of stochastic deflator curve and the deterministic deflator is never larger than 500bps
- The maximum difference between the theoretical and observed return of a Martingale strategy is not larger than 100bps
- Average difference between the theoretical and observed return of a Martingale strategy is not larger than 50bps

```
[509]: test_threshold_max = 0.03  
test_threshold_mean = 0.05
```

```
[510]: test_martingale_threshold_max = 0.01  
test_martingale_threshold_mean = 0.005
```

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

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1.1 Data requirements

To perform this test, the following data was provided:

- Deterministic curves: File:TEST_DATA_DETERMINISTIC.csv
- Stochastic scenario: File:TEST_DATA_STOCHASTIC.csv

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1.2 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 Mat-

plotlib (<https://matplotlib.org/stable/index.html>).

```
[514]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import matplotlib.ticker as mtick
%matplotlib inline
```

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1.2.1 Importing data

Two input files are used in this test.

```
[516]: file_det = 'TEST_DATA_DETERMINISTIC.csv'
file_stoch = 'TEST_DATA_STOCHASTIC.csv'
```

```
[517]: data_det_raw = pd.read_csv(file_det, sep=',', index_col=0)
```

```
[518]: data_stoch_raw = pd.read_csv(file_stoch, sep=',', index_col=0, low_memory=False)
```

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1.3 Data manipulation

This step converts the raw data from the input file into a format that can be digested by the testing framework.

To uniquely identify the correct scenarios, 4 header types are used: 1. Description ex. DEFLATOR ASSET, SPOT RATE, ... 2. Term ex. 1, 2, 3, ... 3. Currency ex. EUR, USD, GBP, ... 4. Scenario number ex. 0, 1, 2, ...

The dataset of both deterministic and stochastic scenarios is filtered for desired curves. The filtering is done using the `filter_scenario_file()` function to avoid duplication of code:

```
[522]: def filter_scenario_file(data, mydescription, myterm, mycurrency):
description_row = data.iloc[0,:]
term_row = data.iloc[1,:]
currency_row = data.iloc[2,:]
true_description = description_row.eq(mydescription, level=None)
true_term = term_row.eq(myterm, level=None)
true_currency = currency_row.eq(mycurrency, level=None)
out = true_description * true_term * true_currency
out = out.tolist()
return out
```

RUN PARAMETER
Currency selection

```
[524]: currency = "EUR"
```

In both files, the currency denominated deflator asset can be identified by:

```
[527]: deflator_variable = "DEFLATOR ASSET"
deflator_term = "0"
```

The currency denominated spot rates can be identified by:

```
[529]: spot_variable = "SPOT RATE"
```

Note that the term of the spot curve can vary between the tests.

The currency denominated total-return index can be identified by:

```
[532]: index_variable = "INDEX"
index_term = "0"
```

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1.3.1 Deterministic scenario preparation

This subsection splits the raw file containing the deterministic scenarios into two tables. The first one contains the headers that are used to identify individual scenarios. The second one contains the numeric data.

```
[535]: data_det_headers = data_det_raw.iloc[:4,:]
```

Numeric data is extracted and transposed such that the time runs across rows as is expected by the testing framework.

```
[537]: data_det = data_det_raw.iloc[4:,:].astype(float)
```

VISUAL INSPECTION

Deterministic headers and scenario data

```
[571]: data_det_headers.head()
```

```
[571]:
```

	DEFLATOR	SPOT_1	SPOT_2	SPOT_3	SPOT_4	\
VARIABLE_KEY						
DESCRIPTION	DEFLATOR ASSET	SPOT RATE	SPOT RATE	SPOT RATE	SPOT RATE	
TERM	0	1	2	3	4	
CURRENCY	EUR	EUR	EUR	EUR	EUR	
SCENARIO	0	0	0	0	0	

	SPOT_5	INDEX
VARIABLE_KEY		
DESCRIPTION	SPOT RATE	INDEX

TERM	5	0
CURRENCY	EUR	EUR
SCENARIO	0	0

```
[572]: data_det.head()
```

```
[572]:
```

	DEFLATOR	SPOT_1	SPOT_2	SPOT_3	SPOT_4	SPOT_5	\
VARIABLE_KEY							
30/09/2022	1.000000	0.025484	0.028942	0.029624	0.030070	0.030430	
31/10/2022	0.998170	0.026134	0.029381	0.029853	0.030273	0.030593	
30/11/2022	0.996310	0.026804	0.029787	0.030077	0.030467	0.030751	
31/12/2022	0.994414	0.027485	0.030155	0.030293	0.030650	0.030900	
31/01/2023	0.992476	0.028166	0.030483	0.030499	0.030821	0.031041	

	INDEX
VARIABLE_KEY	
30/09/2022	1.000000
31/10/2022	1.001861
30/11/2022	1.003674
31/12/2022	1.005618
31/01/2023	1.007612

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1.3.2 Stochastic scenarios preparation

This subsection splits the raw file containing the stochastic scenarios into two tables. One with descriptive headers and the other one with numeric data.

```
[576]: data_stoch_headers = data_stoch_raw.iloc[:4,:]
```

```
[577]: data_stoch = data_stoch_raw.iloc[4:,:].astype(float)
```

VISUAL INSPECTION

Stochastic headers and scenario data

```
[590]: data_stoch_headers.head()
```

```
[590]:
```

	DEFLATOR	SPOT_1	SPOT_2	SPOT_3	SPOT_4	\
VARIABLE_KEY						
DESCRIPTION	DEFLATOR ASSET	SPOT RATE	SPOT RATE	SPOT RATE	SPOT RATE	
TERM	0	1	2	3	4	
CURRENCY	EUR	EUR	EUR	EUR	EUR	
SCENARIO	1	1	1	1	1	

	SPOT_5	INDEX	DEFLATOR.1	SPOT_1.1	SPOT_2.1	...	\
--	--------	-------	------------	----------	----------	-----	---

VARIABLE_KEY	SPOT RATE	INDEX	DEFLATOR	ASSET	SPOT RATE	SPOT RATE	...
DESCRIPTION							...
TERM	5	0		0	1	2	...
CURRENCY	EUR	EUR		EUR	EUR	EUR	...
SCENARIO	1	1		2	2	2	...

	SPOT_4.498	SPOT_5.498	INDEX.498	DEFLATOR.499	SPOT_1.499	\
VARIABLE_KEY						
DESCRIPTION	SPOT RATE	SPOT RATE	INDEX	DEFLATOR	ASSET	SPOT RATE
TERM	4	5	0		0	1
CURRENCY	EUR	EUR	EUR		EUR	EUR
SCENARIO	499	499	499		500	500

	SPOT_2.499	SPOT_3.499	SPOT_4.499	SPOT_5.499	INDEX.499
VARIABLE_KEY					
DESCRIPTION	SPOT RATE	SPOT RATE	SPOT RATE	SPOT RATE	INDEX
TERM	2	3	4	5	0
CURRENCY	EUR	EUR	EUR	EUR	EUR
SCENARIO	500	500	500	500	500

[4 rows x 3500 columns]

```
[591]: data_stoch.head()
```

```
[591]:
```

	DEFLATOR	SPOT_1	SPOT_2	SPOT_3	SPOT_4	SPOT_5	\
VARIABLE_KEY							
30/09/2022	1.000000	0.025484	0.028942	0.029624	0.030070	0.030430	
31/10/2022	0.998170	0.025951	0.029216	0.029703	0.030137	0.030470	
30/11/2022	0.996326	0.026256	0.029277	0.029602	0.030025	0.030339	
31/12/2022	0.994477	0.024064	0.026857	0.027118	0.027591	0.027951	
31/01/2023	0.992825	0.024050	0.026521	0.026690	0.027154	0.027509	

	INDEX	DEFLATOR.1	SPOT_1.1	SPOT_2.1	...	SPOT_4.498	\
VARIABLE_KEY					...		
30/09/2022	1.000000	1.000000	0.025484	0.028942	...	0.030070	
31/10/2022	1.004252	0.998170	0.024965	0.028254	...	0.029700	
30/11/2022	1.013114	0.996408	0.023438	0.026530	...	0.021411	
31/12/2022	1.019324	0.994793	0.023999	0.026794	...	0.022387	
31/01/2023	1.022011	0.993146	0.026649	0.029053	...	0.025759	

	SPOT_5.498	INDEX.498	DEFLATOR.499	SPOT_1.499	SPOT_2.499	\
VARIABLE_KEY						
30/09/2022	0.030430	1.000000	1.000000	0.025484	0.028942	
31/10/2022	0.030045	0.989699	0.998170	0.022864	0.026205	
30/11/2022	0.021956	0.997647	0.996583	0.019003	0.022206	
31/12/2022	0.022887	1.006041	0.995337	0.014633	0.017664	
31/01/2023	0.026152	1.007956	0.994471	0.017642	0.020277	

	SPOT_3.499	SPOT_4.499	SPOT_5.499	INDEX.499
VARIABLE_KEY				
30/09/2022	0.029624	0.030070	0.030430	1.000000
31/10/2022	0.026774	0.027288	0.027698	1.011132
30/11/2022	0.022726	0.023335	0.023829	1.007644
31/12/2022	0.018178	0.018892	0.019486	1.009948
31/01/2023	0.020618	0.021247	0.021761	1.011125

[5 rows x 3500 columns]

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1.4 Currency and indices selection

Each data file contains multiple currencies and curves. In this section, the required curves are extracted. The scope is described in the Limitation section. For each run, only a single currency is chosen.

```
[600]: data_det_headers.head()
```

```
[600]:
```

	DEFLATOR	SPOT_1	SPOT_2	SPOT_3	SPOT_4	\
VARIABLE_KEY						
DESCRIPTION	DEFLATOR ASSET	SPOT RATE	SPOT RATE	SPOT RATE	SPOT RATE	
TERM	0	1	2	3	4	
CURRENCY	EUR	EUR	EUR	EUR	EUR	
SCENARIO	0	0	0	0	0	

	SPOT_5	INDEX
VARIABLE_KEY		
DESCRIPTION	SPOT RATE	INDEX
TERM	5	0
CURRENCY	EUR	EUR
SCENARIO	0	0

Deflator asset; deterministic scenario The deflator asset is represented as a curve of prices of a risk-free bond with a notional value of 1, issued at time 0 and maturing at time t :

$$ZCB_{DET,RF}(0,t)$$

Where: - DET denotes that this Zero Coupon Bond (ZCB) was priced using the deterministic curve

- RF denotes that this Zero Coupon Bond represents a risk-free asset (deflator)
- 0 is the Issue time
- t is the Maturity time

```
[603]: locationindex = filter_scenario_file(\
        data_det_headers, deflator_variable, deflator_term, \
        currency)
```

```
[606]: data_det_headers.iloc[:,locationindex]
```

```
[606]:          DEFLATOR
VARIABLE_KEY
DESCRIPTION  DEFLATOR ASSET
TERM                0
CURRENCY          EUR
SCENARIO          0
```

```
[607]: deflator_det = data_det.iloc[:, locationindex]
```

Spot rate with a 1-year term; deterministic scenario The spot rate with term of 1-year is one of the simulated financial curves. It is represented as a series of annualized returns. The value at time t represents the return of investing 1 unit of currency at time t until time $t+12$ (months):

$$r_{DET,T1}(t)$$

Where:

- DET denotes that this is a return calculated using the deterministic scenario curve
- T1 denotes that this return is calculated using the spot curve with a 1-year term
- t is the time of investment

```
[615]: locationindex = filter_scenario_file(data_det_headers, spot_variable, \
        "1", currency)
```

```
[616]: data_det_headers.iloc[:,locationindex]
```

```
[616]:          SPOT_1
VARIABLE_KEY
DESCRIPTION  SPOT RATE
TERM                1
CURRENCY          EUR
SCENARIO          0
```

```
[617]: spot_1_det = data_det.iloc[:, locationindex]
```

Total-return index; deterministic scenario The total-return indices represent the second kind of simulated financial curves. It is represented as the value of an index with a starting value of 100 at modelling time 0.

It represents the value of the normalized index at time t :

$$I_{DET,1}(t)$$

Where:

- DET denotes that this is a return calculated using the deterministic scenario curve
- 1 denotes that this total-return index represents the first hypothetical assets
- t is the Modelling time

Note that the index is normalized to the starting value of 1 currency unit.

```
[620]: locationindex = filter_scenario_file(data_det_headers, \
                                         index_variable, index_term, currency)
```

```
[621]: data_det_headers.iloc[:,locationindex]
```

```
[621]:          INDEX
VARIABLE_KEY
DESCRIPTION    INDEX
TERM           0
CURRENCY       EUR
SCENARIO       0
```

```
[622]: index_det = data_det.iloc[:, locationindex]
```

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1.4.1 Stochastic scenarios filter

Deflator asset; stochastic scenarios The deflator asset is represented as the price curve of a risk-free bond with a notional value of 1, issued at time 0 and maturing at time t. The curves from the stochastic scenarios file do not have the subscript DET, but instead the stochastic scenario number i. This is represented as:

$$ZCB_{i,RF}(0,t)$$

Where:

- i denotes that this is a Zero Coupon Bond (ZCB) was priced using the i-th stochastic scenario
- RF denotes that this Zero Coupon Bond represents a risk-free asset (deflator)
- 0 is the Issue time
- t is the Maturity time

```
[630]: locationindex = filter_scenario_file(data_stoch_headers, \
                                         deflator_variable, deflator_term, currency)
```

```
[631]: data_stoch_headers.iloc[:,locationindex]
```

```

[631]:
      DEFLATOR      DEFLATOR.1      DEFLATOR.2      DEFLATOR.3 \
VARIABLE_KEY
DESCRIPTION  DEFLATOR ASSET  DEFLATOR ASSET  DEFLATOR ASSET  DEFLATOR ASSET
TERM          0            0            0            0
CURRENCY      EUR          EUR          EUR          EUR
SCENARIO      1            2            3            4

      DEFLATOR.4      DEFLATOR.5      DEFLATOR.6      DEFLATOR.7 \
VARIABLE_KEY
DESCRIPTION  DEFLATOR ASSET  DEFLATOR ASSET  DEFLATOR ASSET  DEFLATOR ASSET
TERM          0            0            0            0
CURRENCY      EUR          EUR          EUR          EUR
SCENARIO      5            6            7            8

      DEFLATOR.8      DEFLATOR.9 ... DEFLATOR.490 \
VARIABLE_KEY
DESCRIPTION  DEFLATOR ASSET  DEFLATOR ASSET ... DEFLATOR ASSET
TERM          0            0 ... 0
CURRENCY      EUR          EUR ... EUR
SCENARIO      9            10 ... 491

      DEFLATOR.491      DEFLATOR.492      DEFLATOR.493      DEFLATOR.494 \
VARIABLE_KEY
DESCRIPTION  DEFLATOR ASSET  DEFLATOR ASSET  DEFLATOR ASSET  DEFLATOR ASSET
TERM          0            0            0            0
CURRENCY      EUR          EUR          EUR          EUR
SCENARIO      492          493          494          495

      DEFLATOR.495      DEFLATOR.496      DEFLATOR.497      DEFLATOR.498 \
VARIABLE_KEY
DESCRIPTION  DEFLATOR ASSET  DEFLATOR ASSET  DEFLATOR ASSET  DEFLATOR ASSET
TERM          0            0            0            0
CURRENCY      EUR          EUR          EUR          EUR
SCENARIO      496          497          498          499

      DEFLATOR.499
VARIABLE_KEY
DESCRIPTION  DEFLATOR ASSET
TERM          0
CURRENCY      EUR
SCENARIO      500

[4 rows x 500 columns]

```

```

[632]: deflator = data_stoch.iloc[:,locationindex]

```

VISUAL INSPECTION

Stochastic deflator asset headers and scenarios

```
[634]: data_stoch_headers.iloc[:,locationindex]
```

```
[634]:
```

	DEFLATOR		DEFLATOR.1		DEFLATOR.2		DEFLATOR.3 \	
VARIABLE_KEY								
DESCRIPTION	DEFLATOR	ASSET	DEFLATOR	ASSET	DEFLATOR	ASSET	DEFLATOR	ASSET
TERM		0		0		0		0
CURRENCY		EUR		EUR		EUR		EUR
SCENARIO		1		2		3		4

	DEFLATOR.4		DEFLATOR.5		DEFLATOR.6		DEFLATOR.7 \	
VARIABLE_KEY								
DESCRIPTION	DEFLATOR	ASSET	DEFLATOR	ASSET	DEFLATOR	ASSET	DEFLATOR	ASSET
TERM		0		0		0		0
CURRENCY		EUR		EUR		EUR		EUR
SCENARIO		5		6		7		8

	DEFLATOR.8		DEFLATOR.9 ...		DEFLATOR.490 \	
VARIABLE_KEY						
DESCRIPTION	DEFLATOR	ASSET	DEFLATOR	ASSET	DEFLATOR	ASSET
TERM		0		0		0
CURRENCY		EUR		EUR		EUR
SCENARIO		9		10		491

	DEFLATOR.491		DEFLATOR.492		DEFLATOR.493		DEFLATOR.494 \	
VARIABLE_KEY								
DESCRIPTION	DEFLATOR	ASSET	DEFLATOR	ASSET	DEFLATOR	ASSET	DEFLATOR	ASSET
TERM		0		0		0		0
CURRENCY		EUR		EUR		EUR		EUR
SCENARIO		492		493		494		495

	DEFLATOR.495		DEFLATOR.496		DEFLATOR.497		DEFLATOR.498 \	
VARIABLE_KEY								
DESCRIPTION	DEFLATOR	ASSET	DEFLATOR	ASSET	DEFLATOR	ASSET	DEFLATOR	ASSET
TERM		0		0		0		0
CURRENCY		EUR		EUR		EUR		EUR
SCENARIO		496		497		498		499

	DEFLATOR.499	
VARIABLE_KEY		
DESCRIPTION	DEFLATOR	ASSET
TERM		0
CURRENCY		EUR
SCENARIO		500

[4 rows x 500 columns]

```
[635]: deflator.head(4)
```

```
[635]:
```

	DEFLATOR	DEFLATOR.1	DEFLATOR.2	DEFLATOR.3	DEFLATOR.4	\
VARIABLE_KEY						
30/09/2022	1.000000	1.000000	1.000000	1.000000	1.000000	
31/10/2022	0.998170	0.998170	0.998170	0.998170	0.998170	
30/11/2022	0.996326	0.996408	0.995904	0.996705	0.996666	
31/12/2022	0.994477	0.994793	0.993855	0.994944	0.994744	

	DEFLATOR.5	DEFLATOR.6	DEFLATOR.7	DEFLATOR.8	DEFLATOR.9	...	\
VARIABLE_KEY						...	
30/09/2022	1.000000	1.000000	1.000000	1.000000	1.000000	...	
31/10/2022	0.998170	0.998170	0.998170	0.998170	0.998170	...	
30/11/2022	0.995663	0.996545	0.996361	0.996389	0.996160	...	
31/12/2022	0.993562	0.994015	0.994973	0.994567	0.994002	...	

	DEFLATOR.490	DEFLATOR.491	DEFLATOR.492	DEFLATOR.493	\
VARIABLE_KEY					
30/09/2022	1.000000	1.000000	1.000000	1.000000	
31/10/2022	0.998170	0.998170	0.998170	0.998170	
30/11/2022	0.996747	0.995999	0.996333	0.996731	
31/12/2022	0.995077	0.993976	0.994701	0.995361	

	DEFLATOR.494	DEFLATOR.495	DEFLATOR.496	DEFLATOR.497	\
VARIABLE_KEY					
30/09/2022	1.000000	1.000000	1.000000	1.000000	
31/10/2022	0.998170	0.998170	0.998170	0.998170	
30/11/2022	0.996151	0.996685	0.996308	0.996043	
31/12/2022	0.995770	0.995182	0.994799	0.994050	

	DEFLATOR.498	DEFLATOR.499
VARIABLE_KEY		
30/09/2022	1.000000	1.000000
31/10/2022	0.998170	0.998170
30/11/2022	0.996366	0.996583
31/12/2022	0.995295	0.995337

[4 rows x 500 columns]

Spot rate with a 1-year term; stochastic scenarios The spot rate with term of 1-year is represented as a series of returns. The value at time t represents the return of investing 1 at time t until time $t+12$ (months) in scenario i .

$$r_{i,T1}(t)$$

Where:

- *i* denotes that this is the return calculated using the *i*-th stochastic scenario
- *T1* denotes that this return is calculated using the spot curve with a 1-year term
- *t* is the time of investment

```
[639]: locationindex = filter_scenario_file(data_stoch_headers,\
spot_variable, "1", currency)
```

```
[640]: spot_1 = data_stoch.iloc[:,locationindex]
```

VISUAL INSPECTION

Stochastic spot rate with 1-year term; headers and scenarios

```
[642]: data_stoch_headers.iloc[:,locationindex]
```

```
[642]:
```

	SPOT_1	SPOT_1.1	SPOT_1.2	SPOT_1.3	SPOT_1.4	\
VARIABLE_KEY						
DESCRIPTION	SPOT RATE	SPOT RATE	SPOT RATE	SPOT RATE	SPOT RATE	
TERM	1	1	1	1	1	
CURRENCY	EUR	EUR	EUR	EUR	EUR	
SCENARIO	1	2	3	4	5	

	SPOT_1.5	SPOT_1.6	SPOT_1.7	SPOT_1.8	SPOT_1.9	...	\
VARIABLE_KEY						...	
DESCRIPTION	SPOT RATE	SPOT RATE	SPOT RATE	SPOT RATE	SPOT RATE	...	
TERM	1	1	1	1	1	...	
CURRENCY	EUR	EUR	EUR	EUR	EUR	...	
SCENARIO	6	7	8	9	10	...	

	SPOT_1.490	SPOT_1.491	SPOT_1.492	SPOT_1.493	SPOT_1.494	\
VARIABLE_KEY						
DESCRIPTION	SPOT RATE	SPOT RATE	SPOT RATE	SPOT RATE	SPOT RATE	
TERM	1	1	1	1	1	
CURRENCY	EUR	EUR	EUR	EUR	EUR	
SCENARIO	491	492	493	494	495	

	SPOT_1.495	SPOT_1.496	SPOT_1.497	SPOT_1.498	SPOT_1.499
VARIABLE_KEY					
DESCRIPTION	SPOT RATE	SPOT RATE	SPOT RATE	SPOT RATE	SPOT RATE
TERM	1	1	1	1	1
CURRENCY	EUR	EUR	EUR	EUR	EUR
SCENARIO	496	497	498	499	500

[4 rows x 500 columns]

```
[643]: spot_1.head()
```

```
[643]:
```

	SPOT_1	SPOT_1.1	SPOT_1.2	SPOT_1.3	SPOT_1.4	SPOT_1.5	\
VARIABLE_KEY							
30/09/2022	0.025484	0.025484	0.025484	0.025484	0.025484	0.025484	
31/10/2022	0.025951	0.024965	0.031043	0.021404	0.021871	0.033967	
30/11/2022	0.026256	0.023438	0.028679	0.025185	0.027129	0.029316	
31/12/2022	0.024064	0.023999	0.031913	0.029466	0.026192	0.024812	
31/01/2023	0.024050	0.026649	0.033978	0.029959	0.036508	0.022529	

	SPOT_1.6	SPOT_1.7	SPOT_1.8	SPOT_1.9	...	SPOT_1.490	\
VARIABLE_KEY					...		
30/09/2022	0.025484	0.025484	0.025484	0.025484	...	0.025484	
31/10/2022	0.023326	0.025541	0.025197	0.027956	...	0.020897	
30/11/2022	0.034493	0.020698	0.025932	0.029994	...	0.024095	
31/12/2022	0.039323	0.023184	0.024342	0.030728	...	0.019358	
31/01/2023	0.045296	0.012303	0.029277	0.031384	...	0.020451	

	SPOT_1.491	SPOT_1.492	SPOT_1.493	SPOT_1.494	SPOT_1.495	\
VARIABLE_KEY						
30/09/2022	0.025484	0.025484	0.025484	0.025484	0.025484	
31/10/2022	0.029902	0.025875	0.021099	0.028071	0.021642	
30/11/2022	0.028359	0.023632	0.020486	0.008710	0.022090	
31/12/2022	0.031593	0.021744	0.029503	0.006404	0.022279	
31/01/2023	0.037064	0.031838	0.031011	0.013253	0.020579	

	SPOT_1.496	SPOT_1.497	SPOT_1.498	SPOT_1.499
VARIABLE_KEY				
30/09/2022	0.025484	0.025484	0.025484	0.025484
31/10/2022	0.026177	0.029374	0.025478	0.022864
30/11/2022	0.022156	0.027993	0.016918	0.019003
31/12/2022	0.023948	0.027462	0.018420	0.014633
31/01/2023	0.019591	0.021532	0.022537	0.017642

[5 rows x 500 columns]

Spot rate with a 2-year term; stochastic scenarios The spot rate with term of 2 years is represented as a series of annualized returns. The value at time t represents the annualized return of investing 1 unit of currency at time t until time $t+24$ (months) in scenario i . This is represented as:

$$r_{i,T2}(t)$$

Where:

- i denotes that this is the return calculated using the i -th stochastic scenario
- T2 denotes that this return is calculated using the spot curve with a 2-year term
- t is the time of investing

```
[647]: locationindex = filter_scenario_file(data_stoch_headers, \
                                         spot_variable, "2", currency)
```

```
[648]: spot_2 = data_stoch.iloc[:,locationindex]
```

VISUAL INSPECTION

Stochastic spot rate with 2-year term; headers and scenarios

```
[654]: data_stoch_headers.iloc[:,locationindex]
```

```
[654]:
```

	SPOT_2	SPOT_2.1	SPOT_2.2	SPOT_2.3	SPOT_2.4	\
VARIABLE_KEY						
DESCRIPTION	SPOT RATE	SPOT RATE	SPOT RATE	SPOT RATE	SPOT RATE	
TERM	2	2	2	2	2	
CURRENCY	EUR	EUR	EUR	EUR	EUR	
SCENARIO	1	2	3	4	5	

	SPOT_2.5	SPOT_2.6	SPOT_2.7	SPOT_2.8	SPOT_2.9	...	\
VARIABLE_KEY						...	
DESCRIPTION	SPOT RATE	SPOT RATE	SPOT RATE	SPOT RATE	SPOT RATE	...	
TERM	2	2	2	2	2	...	
CURRENCY	EUR	EUR	EUR	EUR	EUR	...	
SCENARIO	6	7	8	9	10	...	

	SPOT_2.490	SPOT_2.491	SPOT_2.492	SPOT_2.493	SPOT_2.494	\
VARIABLE_KEY						
DESCRIPTION	SPOT RATE	SPOT RATE	SPOT RATE	SPOT RATE	SPOT RATE	
TERM	2	2	2	2	2	
CURRENCY	EUR	EUR	EUR	EUR	EUR	
SCENARIO	491	492	493	494	495	

	SPOT_2.495	SPOT_2.496	SPOT_2.497	SPOT_2.498	SPOT_2.499
VARIABLE_KEY					
DESCRIPTION	SPOT RATE	SPOT RATE	SPOT RATE	SPOT RATE	SPOT RATE
TERM	2	2	2	2	2
CURRENCY	EUR	EUR	EUR	EUR	EUR
SCENARIO	496	497	498	499	500

[4 rows x 500 columns]

```
[655]: spot_2.head()
```

```
[655]:
```

	SPOT_2	SPOT_2.1	SPOT_2.2	SPOT_2.3	SPOT_2.4	SPOT_2.5	\
VARIABLE_KEY							
30/09/2022	0.028942	0.028942	0.028942	0.028942	0.028942	0.028942	
31/10/2022	0.029216	0.028254	0.034180	0.024781	0.025237	0.037030	
30/11/2022	0.029277	0.026530	0.031639	0.028232	0.030128	0.032259	
31/12/2022	0.026857	0.026794	0.034505	0.032121	0.028931	0.027586	
31/01/2023	0.026521	0.029053	0.036191	0.032277	0.038656	0.025039	

	SPOT_2.6	SPOT_2.7	SPOT_2.8	SPOT_2.9	...	SPOT_2.490	\
VARIABLE_KEY					...		
30/09/2022	0.028942	0.028942	0.028942	0.028942	...	0.028942	
31/10/2022	0.026655	0.028815	0.028480	0.031170	...	0.024287	
30/11/2022	0.037305	0.023859	0.028961	0.032920	...	0.027170	
31/12/2022	0.041724	0.025999	0.027128	0.033350	...	0.022270	
31/01/2023	0.047213	0.015074	0.031613	0.033665	...	0.023014	

	SPOT_2.491	SPOT_2.492	SPOT_2.493	SPOT_2.494	SPOT_2.495	\
VARIABLE_KEY						
30/09/2022	0.028942	0.028942	0.028942	0.028942	0.028942	
31/10/2022	0.033067	0.029142	0.024484	0.031282	0.025014	
30/11/2022	0.031326	0.026718	0.023651	0.012170	0.025215	
31/12/2022	0.034193	0.024596	0.032158	0.009642	0.025118	
31/01/2023	0.039197	0.034107	0.033302	0.016000	0.023138	

	SPOT_2.496	SPOT_2.497	SPOT_2.498	SPOT_2.499
VARIABLE_KEY				
30/09/2022	0.028942	0.028942	0.028942	0.028942
31/10/2022	0.029436	0.032552	0.028754	0.026205
30/11/2022	0.025280	0.030970	0.020173	0.022206
31/12/2022	0.026744	0.030169	0.021356	0.017664
31/01/2023	0.022176	0.024067	0.025046	0.020277

[5 rows x 500 columns]

Extra content

```
[657]: locationindex = filter_scenario_file(data_stoch_headers, spot_variable, "3",  
      ↪currency)  
spot_3 = data_stoch.iloc[:,locationindex]
```

```
[658]: locationindex = filter_scenario_file(data_stoch_headers, spot_variable, "4",  
      ↪currency)  
spot_4 = data_stoch.iloc[:,locationindex]
```

```
[659]: locationindex = filter_scenario_file(data_stoch_headers, spot_variable, "5",  
      ↪currency)
```



```
spot_5 = data_stoch.iloc[:,locationindex]
```

Total-return index stochastic scenarios The total-return indices are represented as the value of an index at time t with a starting value of 1 units of currency at modelling time 0 in scenario i .

$$I_{i,1}(t)$$

Where:

- i denotes that this index is calculated using the i -th stochastic scenario
- 1 denotes that this total-return index represents the first hypothetical asset
- t is the Modelling time

```
[671]: locationindex = filter_scenario_file(data_stoch_headers, index_variable,
↪index_term, currency)
```

```
[672]: first_index = data_stoch.iloc[:,locationindex]
```

VISUAL INSPECTION

Total-return index; headers and scenarios

```
[674]: data_stoch_headers.iloc[:,locationindex]
```

```
[674]:
```

	INDEX	INDEX.1	INDEX.2	INDEX.3	INDEX.4	INDEX.5	INDEX.6	INDEX.7	\
VARIABLE_KEY									
DESCRIPTION	INDEX	INDEX	INDEX	INDEX	INDEX	INDEX	INDEX	INDEX	
TERM	0	0	0	0	0	0	0	0	
CURRENCY	EUR	EUR	EUR	EUR	EUR	EUR	EUR	EUR	
SCENARIO	1	2	3	4	5	6	7	8	

	INDEX.8	INDEX.9	...	INDEX.490	INDEX.491	INDEX.492	INDEX.493	\
VARIABLE_KEY			...					
DESCRIPTION	INDEX	INDEX	...	INDEX	INDEX	INDEX	INDEX	
TERM	0	0	...	0	0	0	0	
CURRENCY	EUR	EUR	...	EUR	EUR	EUR	EUR	
SCENARIO	9	10	...	491	492	493	494	

	INDEX.494	INDEX.495	INDEX.496	INDEX.497	INDEX.498	INDEX.499
VARIABLE_KEY						
DESCRIPTION	INDEX	INDEX	INDEX	INDEX	INDEX	INDEX
TERM	0	0	0	0	0	0
CURRENCY	EUR	EUR	EUR	EUR	EUR	EUR
SCENARIO	495	496	497	498	499	500

[4 rows x 500 columns]

```
[675]: first_index.head()
```

```
[675]:
```

	INDEX	INDEX.1	INDEX.2	INDEX.3	INDEX.4	INDEX.5	\
VARIABLE_KEY							
30/09/2022	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	
31/10/2022	1.004252	0.991676	0.996692	1.009442	1.006826	0.998891	
30/11/2022	1.013114	0.998047	1.004273	1.013157	0.995257	0.998347	
31/12/2022	1.019324	1.000392	1.007218	1.019881	0.995524	0.989838	
31/01/2023	1.022011	1.004860	1.006790	1.023282	1.002880	0.989899	

	INDEX.6	INDEX.7	INDEX.8	INDEX.9	...	INDEX.490	\
VARIABLE_KEY					...		
30/09/2022	1.000000	1.000000	1.000000	1.000000	...	1.000000	
31/10/2022	1.015911	0.987568	1.006754	1.008715	...	0.998135	
30/11/2022	1.019634	0.987449	1.009823	1.019360	...	1.000675	
31/12/2022	1.012374	0.975855	1.005642	1.009754	...	1.009533	
31/01/2023	1.021796	0.980949	1.011963	1.019708	...	1.011067	

	INDEX.491	INDEX.492	INDEX.493	INDEX.494	INDEX.495	\
VARIABLE_KEY						
30/09/2022	1.000000	1.000000	1.000000	1.000000	1.000000	
31/10/2022	0.994408	1.013690	0.991886	1.006662	1.009607	
30/11/2022	1.003585	1.008598	0.999374	1.016761	1.001836	
31/12/2022	1.011131	1.027138	1.015096	1.023693	1.016197	
31/01/2023	1.017271	1.028211	1.010807	1.021912	1.017094	

	INDEX.496	INDEX.497	INDEX.498	INDEX.499
VARIABLE_KEY				
30/09/2022	1.000000	1.000000	1.000000	1.000000
31/10/2022	1.009335	0.992735	0.989699	1.011132
30/11/2022	1.001443	0.997197	0.997647	1.007644
31/12/2022	0.991262	0.995540	1.006041	1.009948
31/01/2023	0.994203	1.001523	1.007956	1.011125

[5 rows x 500 columns]

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1.4.2 Stochastic deflator performance calculation

To calculate the performance of the deflator asset in different time increments, first the monthly return at each time increment t in each scenario i is calculated:

$$r_{i,RF}^{1M}(t) = \frac{ZCB_{i,RF}(0, t-1)}{ZCB_{i,RF}(0, t)} - 1$$

Annualized (12 month) performance can be calculated as the 12-month rolling product of the monthly capitalisation factor:

$$r_{i,RF}(t) = \prod_{j=t}^{t+12} (1 + r_{i,RF}^{1M}(j)) - 1$$

Where:

- $ZCB_{i,RF}(0, t)$ is the value of a ZCB issued at time 0 with a Maturity at time t priced using the i -th scenario
- $r_{i,RF}^{1M}(t)$ is the 1-month return of the deflator asset at time t under the i -th scenario
- $r_{i,RF}(t)$ is the yearly (12-month) return of the deflator asset between times t and $t+12$ under the i -th scenario

```
[704]: r_d_1_t = deflator.shift()/ deflator # 1+ monthly return
```

```
[705]: r_d_12_t = r_d_1_t.rolling(window=12).apply(np.prod, raw=True) - 1 # annual
↪return
```

```
[706]: r_d_12_t = r_d_12_t.iloc[12:,:] # remove NaN (13) entries;
```

```
[707]: r_d_12_t.index = r_d_1_t.index[:-12] # Corrected index dates
```

VISUAL INSPECTION

Deflator asset monthly and annual performance

```
[732]: r_d_1_t.head()
```

```
[732]:
```

	DEFLATOR	DEFLATOR.1	DEFLATOR.2	DEFLATOR.3	DEFLATOR.4	\
VARIABLE_KEY						
30/09/2022	NaN	NaN	NaN	NaN	NaN	
31/10/2022	1.001833	1.001833	1.001833	1.001833	1.001833	
30/11/2022	1.001851	1.001768	1.002275	1.001470	1.001509	
31/12/2022	1.001859	1.001623	1.002062	1.001770	1.001932	
31/01/2023	1.001664	1.001658	1.002318	1.002115	1.001842	

	DEFLATOR.5	DEFLATOR.6	DEFLATOR.7	DEFLATOR.8	DEFLATOR.9	...	\
VARIABLE_KEY						...	
30/09/2022	NaN	NaN	NaN	NaN	NaN	...	
31/10/2022	1.001833	1.001833	1.001833	1.001833	1.001833	...	
30/11/2022	1.002518	1.001631	1.001816	1.001787	1.002018	...	
31/12/2022	1.002115	1.002545	1.001395	1.001832	1.002171	...	
31/01/2023	1.001726	1.002932	1.001590	1.001688	1.002220	...	

	DEFLATOR.490	DEFLATOR.491	DEFLATOR.492	DEFLATOR.493	\
VARIABLE_KEY					

30/09/2022	NaN	NaN	NaN	NaN
31/10/2022	1.001833	1.001833	1.001833	1.001833
30/11/2022	1.001428	1.002180	1.001844	1.001444
31/12/2022	1.001678	1.002035	1.001641	1.001376
31/01/2023	1.001269	1.002292	1.001469	1.002118

	DEFLATOR.494	DEFLATOR.495	DEFLATOR.496	DEFLATOR.497	\
VARIABLE_KEY					
30/09/2022	NaN	NaN	NaN	NaN	
31/10/2022	1.001833	1.001833	1.001833	1.001833	
30/11/2022	1.002027	1.001490	1.001869	1.002135	
31/12/2022	1.000383	1.001510	1.001517	1.002005	
31/01/2023	1.000175	1.001515	1.001654	1.001948	

	DEFLATOR.498	DEFLATOR.499
VARIABLE_KEY		
30/09/2022	NaN	NaN
31/10/2022	1.001833	1.001833
30/11/2022	1.001811	1.001592
31/12/2022	1.001076	1.001252
31/01/2023	1.001191	1.000871

[5 rows x 500 columns]

```
[734]: r_d_12_t.head()
```

	DEFLATOR	DEFLATOR.1	DEFLATOR.2	DEFLATOR.3	DEFLATOR.4	\
VARIABLE_KEY						
30/09/2022	0.020827	0.026285	0.034641	0.022062	0.032137	
31/10/2022	0.022666	0.027230	0.038219	0.021999	0.034811	
30/11/2022	0.024252	0.028432	0.041514	0.022603	0.037150	
31/12/2022	0.025756	0.030051	0.044917	0.022805	0.039650	
31/01/2023	0.028001	0.031371	0.047627	0.022665	0.042654	

	DEFLATOR.5	DEFLATOR.6	DEFLATOR.7	DEFLATOR.8	DEFLATOR.9	...	\
VARIABLE_KEY						...	
30/09/2022	0.020858	0.038673	0.018733	0.029472	0.022307	...	
31/10/2022	0.020307	0.040746	0.019605	0.029984	0.020637	...	
30/11/2022	0.019316	0.043069	0.020066	0.031586	0.018773	...	
31/12/2022	0.018710	0.045469	0.020515	0.032690	0.017706	...	
31/01/2023	0.018060	0.047791	0.020101	0.033502	0.016815	...	

	DEFLATOR.490	DEFLATOR.491	DEFLATOR.492	DEFLATOR.493	\
VARIABLE_KEY					
30/09/2022	0.019504	0.036817	0.032964	0.018647	
31/10/2022	0.020299	0.039327	0.034376	0.017440	
30/11/2022	0.020958	0.041613	0.035730	0.016307	

31/12/2022	0.021372	0.044232	0.037461	0.015837
31/01/2023	0.022236	0.046724	0.039662	0.015525

	DEFLATOR.494	DEFLATOR.495	DEFLATOR.496	DEFLATOR.497 \
VARIABLE_KEY				
30/09/2022	0.020413	0.023793	0.020776	0.022708
31/10/2022	0.021587	0.025199	0.021842	0.023527
30/11/2022	0.022378	0.026491	0.022654	0.024252
31/12/2022	0.025352	0.028517	0.023763	0.024909
31/01/2023	0.028822	0.031275	0.024574	0.026397

	DEFLATOR.498	DEFLATOR.499
VARIABLE_KEY		
30/09/2022	0.018307	0.010119
31/10/2022	0.018646	0.008269
30/11/2022	0.019662	0.006607
31/12/2022	0.021212	0.005481
31/01/2023	0.022691	0.004750

[5 rows x 500 columns]

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2 Test 1; Mean of stochastic deflator is equal to the deterministic

The objective of this test is to compare the average performance of the stochastic deflation asset and the deterministic deflator.

The two test statistics used are:

$$S_{AVERAGE}^1 = \frac{1}{T-12} \sum_{t=0}^{T-12} \frac{1}{N} \sum_{i=0}^N |ZCB_{i,RF}(0, t) - ZCB_{DET,RF}(0, t)|$$

$$S_{MAX}^1 = \max_t \frac{1}{N} \left| \sum_{i=0}^N ZCB_{i,RF}(0, t) - ZCB_{DET,RF}(0, t) \right|$$

Where:

- N ... is the number scenarios
- T ... is the number of time increments
- $ZCB_{i,RF}(0, t)$... is the deflator asset between time 0 and t , priced using the stochastic deflator scenario i
- $ZCB_{DET,RF}(0, t)$... is the deflator asset between time 0 and t priced using the deterministic scenario

- $S_{AVERAGE}^1$... is the average test statistics for the first test
- S_{MAX}^1 ... is the maximum deviation test statistics for the first test

```
[759]: deflator_stoch_mean = pd.DataFrame(deflator.mean(axis=1), columns=["Deflator_↵
↵mean"])
```

```
[760]: test_statistics_1 = np.abs(deflator_det.values-deflator_stoch_mean.values)
```

VISUAL INSPECTION

Deflator asset test

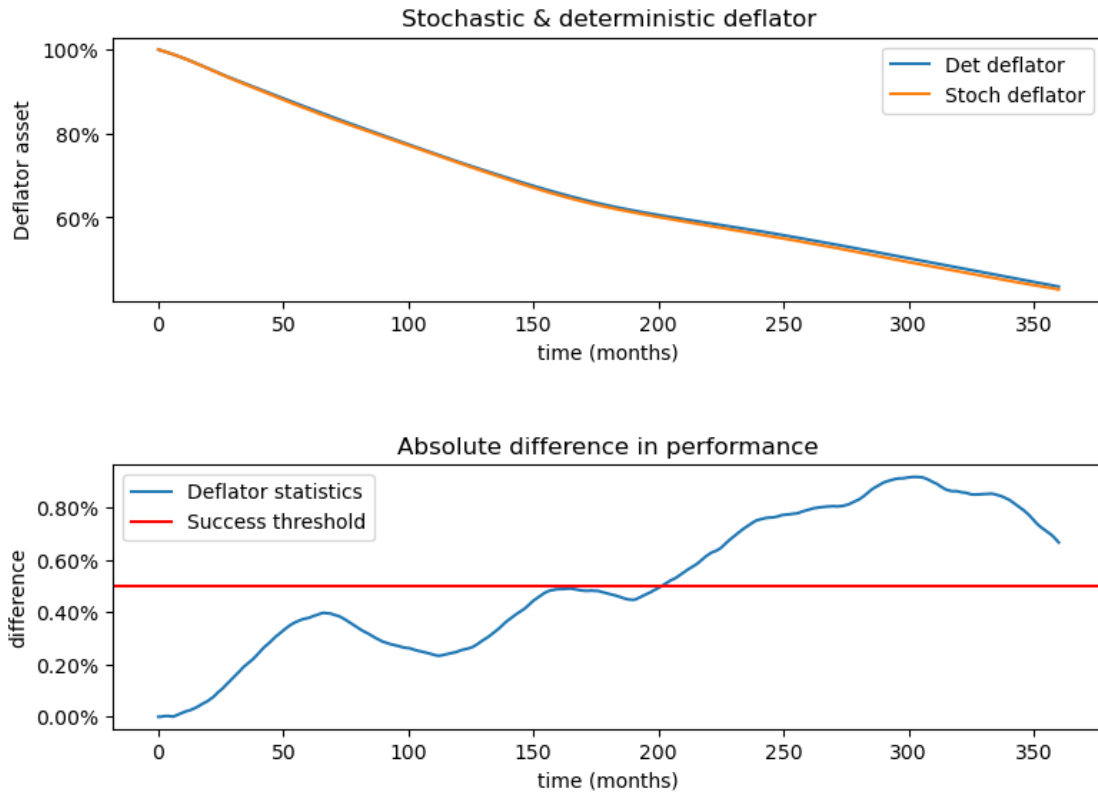
```
[762]: x_data_label = range(0, deflator_det.shape[0], 1)
fig, (ax1, ax2) = plt.subplots(2, 1)
ax1.plot(x_data_label, deflator_det.values*100, color='tab:blue', label="Det_↵
↵deflator")
ax1.plot(x_data_label, deflator_stoch_mean.values*100, color='tab:
↵orange', label="Stoch deflator")
ax1.legend()
ax1.set_xlabel("time (months)")
ax1.set_ylabel("Deflator asset")
ax1.set_title('Stochastic & deterministic deflator')

ax2.plot(x_data_label, test_statistics_1*100, label="Deflator statistics")
ax2.axhline(y = test_martingale_threshold_mean*100, color = 'r', label="Success_↵
↵threshold")
ax2.legend()
ax2.set_xlabel("time (months)")
ax2.set_ylabel("difference")
ax2.set_title('Absolute difference in performance')
fig.tight_layout(h_pad=2)

ax1.yaxis.set_major_formatter(mtick.PercentFormatter())
ax2.yaxis.set_major_formatter(mtick.PercentFormatter())

fig.set_figwidth(8)
fig.set_figheight(6)

plt.show()
```



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2.1 Test 1; Success criteria

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

```
[775]: result1 = success_test(test_statistics_1, test_threshold_max,
    ↪test_threshold_mean)
```

Test passed

Test passed

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2.2 Spot curve cumulative performance

The spot curve time-series contains annualized returns starting at every month. To calculate the cumulative performance for longer time periods, three extra steps must be performed:

1. Separate non-chained returns
2. Convert the annualized return into the actual return for the entire period
3. Aggregate the chained returns through time

The function `spot_cumulative()` applies the 3 steps above. This example is limited to terms 1 and 2 with terms 3 to 10 calculated in the Extra content section. It is possible to extend the methodology

to higher terms.

```
[778]: def spot_cumulative(spot, term):
        out = (1+spot).pow(term)
        for iRow in range(term*12,out.shape[0]):
            out.iloc[iRow, :] = out.iloc[iRow, :]*out.iloc[iRow-term*12, :]
        out = out.iloc[:(-term*12),:]
        return out
```

```
[779]: spot_1_cum = spot_cumulative(spot_1, 1)
```

```
[780]: spot_2_cum = spot_cumulative(spot_2, 2)
```

Extra content

```
[782]: spot_3_cum = spot_cumulative(spot_3, 3)
        spot_4_cum = spot_cumulative(spot_4, 4)
        spot_5_cum = spot_cumulative(spot_5, 5)
```

VISUAL INSPECTION

Spot rate cumulative performance

```
[791]: spot_1_cum.head()
```

```
[791]:
```

	SPOT_1	SPOT_1.1	SPOT_1.2	SPOT_1.3	SPOT_1.4	SPOT_1.5	\
VARIABLE_KEY							
30/09/2022	1.025484	1.025484	1.025484	1.025484	1.025484	1.025484	
31/10/2022	1.025951	1.024965	1.031043	1.021404	1.021871	1.033967	
30/11/2022	1.026256	1.023438	1.028679	1.025185	1.027129	1.029316	
31/12/2022	1.024064	1.023999	1.031913	1.029466	1.026192	1.024812	
31/01/2023	1.024050	1.026649	1.033978	1.029959	1.036508	1.022529	

	SPOT_1.6	SPOT_1.7	SPOT_1.8	SPOT_1.9	...	SPOT_1.490	\
VARIABLE_KEY					...		
30/09/2022	1.025484	1.025484	1.025484	1.025484	...	1.025484	
31/10/2022	1.023326	1.025541	1.025197	1.027956	...	1.020897	
30/11/2022	1.034493	1.020698	1.025932	1.029994	...	1.024095	
31/12/2022	1.039323	1.023184	1.024342	1.030728	...	1.019358	
31/01/2023	1.045296	1.012303	1.029277	1.031384	...	1.020451	

	SPOT_1.491	SPOT_1.492	SPOT_1.493	SPOT_1.494	SPOT_1.495	\
VARIABLE_KEY						
30/09/2022	1.025484	1.025484	1.025484	1.025484	1.025484	
31/10/2022	1.029902	1.025875	1.021099	1.028071	1.021642	
30/11/2022	1.028359	1.023632	1.020485	1.008710	1.022090	
31/12/2022	1.031593	1.021744	1.029503	1.006404	1.022279	
31/01/2023	1.037064	1.031838	1.031011	1.013253	1.020579	

	SPOT_1.496	SPOT_1.497	SPOT_1.498	SPOT_1.499
VARIABLE_KEY				
30/09/2022	1.025484	1.025484	1.025484	1.025484
31/10/2022	1.026177	1.029373	1.025478	1.022864
30/11/2022	1.022156	1.027993	1.016918	1.019003
31/12/2022	1.023948	1.027462	1.018420	1.014633
31/01/2023	1.019591	1.021532	1.022537	1.017642

[5 rows x 500 columns]

[792]: `spot_2_cum.head()`

[792]:

	SPOT_2	SPOT_2.1	SPOT_2.2	SPOT_2.3	SPOT_2.4	SPOT_2.5	\
VARIABLE_KEY							
30/09/2022	1.058721	1.058721	1.058721	1.058721	1.058721	1.058721	
31/10/2022	1.059285	1.057306	1.069527	1.050177	1.051110	1.075431	
30/11/2022	1.059411	1.053763	1.064278	1.057262	1.061163	1.065559	
31/12/2022	1.054435	1.054305	1.070200	1.065274	1.058698	1.055934	
31/01/2023	1.053745	1.058949	1.073692	1.065596	1.078806	1.050705	

	SPOT_2.6	SPOT_2.7	SPOT_2.8	SPOT_2.9	...	SPOT_2.490	\
VARIABLE_KEY					...		
30/09/2022	1.058721	1.058721	1.058721	1.058721	...	1.058721	
31/10/2022	1.054021	1.058461	1.057771	1.063311	...	1.049165	
30/11/2022	1.076002	1.048287	1.058760	1.066923	...	1.055077	
31/12/2022	1.085189	1.052674	1.054992	1.067813	...	1.045036	
31/01/2023	1.096655	1.030376	1.064225	1.068463	...	1.046557	

	SPOT_2.491	SPOT_2.492	SPOT_2.493	SPOT_2.494	SPOT_2.495	\
VARIABLE_KEY						
30/09/2022	1.058721	1.058721	1.058721	1.058721	1.058721	
31/10/2022	1.067228	1.059132	1.049568	1.063542	1.050653	
30/11/2022	1.063634	1.054150	1.047862	1.024487	1.051066	
31/12/2022	1.069555	1.049797	1.065349	1.019377	1.050866	
31/01/2023	1.079930	1.069377	1.067713	1.032255	1.046812	

	SPOT_2.496	SPOT_2.497	SPOT_2.498	SPOT_2.499
VARIABLE_KEY				
30/09/2022	1.058721	1.058721	1.058721	1.058721
31/10/2022	1.059738	1.066164	1.058334	1.053097
30/11/2022	1.051200	1.062898	1.040754	1.044905
31/12/2022	1.054203	1.061247	1.043169	1.035641
31/01/2023	1.044843	1.048714	1.050720	1.040965

[5 rows x 500 columns]

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3 Martingale tests

To evaluate the market-consistency, IVASS regulation 18 article 57 recommends to perform the Martingale test on asset classes used in the calibration process of the ESG and for simple financial strategies. The following asset classes are covered:

- Spot curves
- Total-return indices

Article 57 recommends using multiple simple investment strategies. Two are used in this example:

- Forward return with a spread of 1 year
- Buy-and-hold strategy

3.1 Forward return with a spread of 1 year investment strategy

Forward return with a spread of 1-year strategy consists of buying and selling fixed income assets in a way that the return is equal to a forward rate with constant maturity. In this example, a constant maturity of 1-year is implemented by buying (long position) a ZCB with a maturity of $K+1$ years and selling (short position) a ZCB with a maturity of K -years.

$$\begin{aligned}
 ZCB_{i,TK}(t) &= \frac{1}{(1 + r_{i,TK}(t))^K} \\
 Perf_{i,TK,TK+1}^{FR}(t) &= \frac{ZCB_{i,TK}(t)}{ZCB_{i,TK+1}(t)} = \\
 &= \frac{\frac{1}{(1+r_{i,TK}(t))^K}}{\frac{1}{(1+r_{i,TK+1}(t))^{K+1}}} = \frac{(1 + r_{i,TK+1}(t))^{K+1}}{(1 + r_{i,TK}(t))^K}
 \end{aligned}$$

Therefore:

$$DPerf_{i,TK,TK+1}^{FR}(t) = \frac{Perf_{i,TK,TK+1}^{FR}(t)}{(1 + r_{i,RF}(t + 12 * K))}$$

is a Martingale.

The test verifies the following asymptotic identity:

$$E^Q[Perf_{i,TK,TK+1}^{FR}(t)] \stackrel{?}{=} 1$$

Where:

- Q is the numeraire used in this test which is the 1-year forward rate of the deflator

The function `forward_strategy_test()` performs the beforementioned steps for any sequential spot rate terms

```
[801]: def forward_strategy_test(annualdef, spot, spotnext, term):
    if term>1:
        annualdef= annualdef.iloc[:-(term-1)*12,:]

    perf = ((1+spotnext).pow(term+1))/((1+spot).pow(term)).values
    perf_shifted = perf.shift(term*12)
    Mspot = perf_shifted.iloc[:term*12,:].values/(1+annualdef).values
    Mspot = pd.DataFrame(Mspot)
    Mspot = Mspot.iloc[term*12:,:]
    return (Mspot - 1).mean(axis=1)
```

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3.2 Test 2; Forward return with a spread of 1-year investment strategy

Forward return with a spread of 1-year strategy using 1- and 2-year spot rates consists of buying and selling ZCB bonds with 1- and 2-year term. In this example, this consists of buying (long position) a ZCB with a maturity of 2 years and selling (short position) a ZCB with a maturity of 1 year.

$$ZCB_{i,T1}(t) = \frac{1}{(1 + r_{i,T1}(t))}$$

$$ZCB_{i,T2}(t) = \frac{1}{(1 + r_{i,T2}(t))^2}$$

$$Perf_{i,T1,T2}^{FR}(t) = \frac{(1 + r_{i,T2}(t))^2}{(1 + r_{i,T1}(t))}$$

The strategy consists of selling the 1-year ZCB bond and using the proceeds to buy a ZCB bond with a maturity of 2 years.

In a risk-neutral scenario this strategy can be replicated by depositing one unit at year 1 into the deflator asset and withdrawing it at time 2.

Therefore:

$$DPerf_{i,T1,T2}^{FR}(t) = \frac{Perf_{i,T1,T2}^{FR}(t)}{(1 + r_{i,RF}(t + 12))}$$

The test verifies the following asymptotic identity:

$$E^Q[Perf_{i,T1,T2}^{FR}(t)] \stackrel{?}{=} 1$$

Where:

- Q is the numeraire used in this test which is the 1-year forward rate of the deflator

The two test statistics used are:

$$S_{AVERAGE} = \frac{1}{T-12} \sum_{t=0}^{T-12} \frac{1}{N} \sum_{i=0}^N |DPerf_{i,T1,T2}^{FR}(t) - 1|$$

$$S_{MAX} = \max_t \frac{1}{N} \left| \sum_{i=0}^N (DPerf_{i,T1,T2}^{FR}(t) - 1) \right|$$

```
[809]: test_statistics_2 = forward_strategy_test(r_d_12_t, spot_1, spot_2, 1)
```

VISUAL INSPECTION

Forward return strategy using the 1- and 2-year spot rate

```
[811]: x_data_label = range(0, test_statistics_2.shape[0], 1)
fig, (ax1, ax2) = plt.subplots(2, 1)
ax1.plot(x_data_label, test_statistics_2.values*100, color='tab:blue')

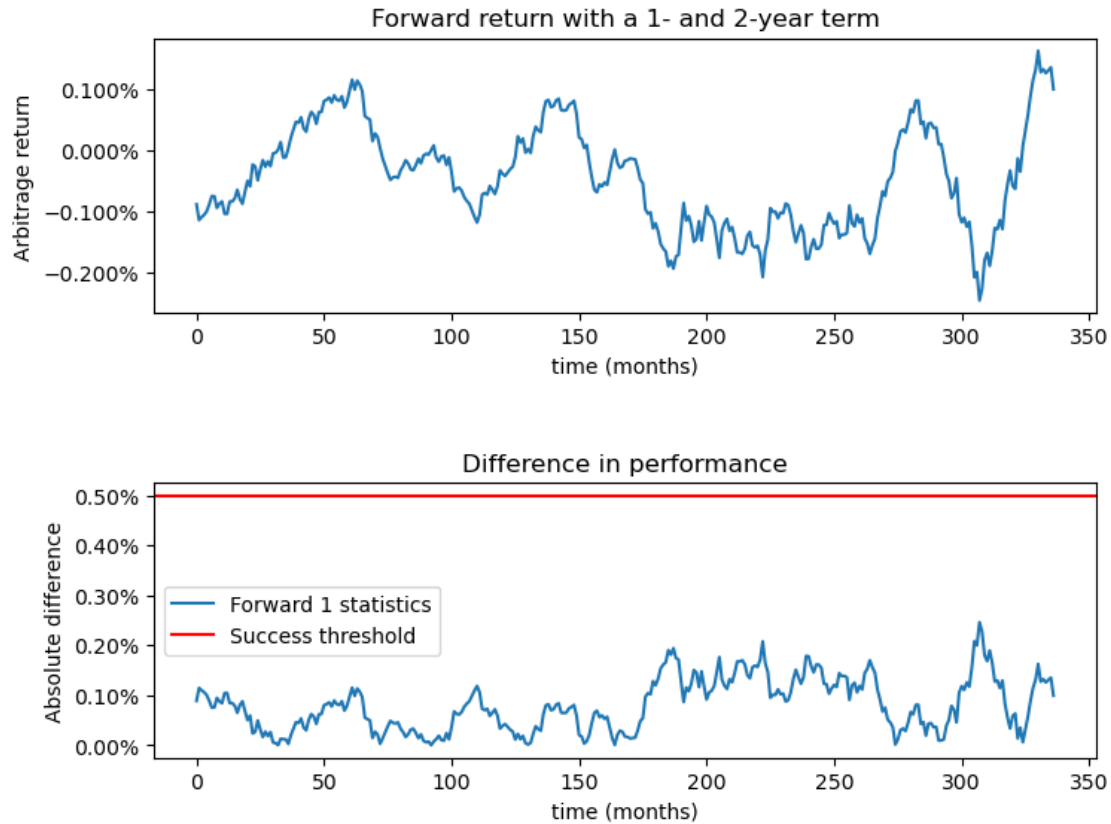
ax1.set_xlabel("time (months)")
ax1.set_ylabel("Arbitrage return")
ax1.set_title('Forward return with a 1- and 2-year term')

ax1.yaxis.set_major_formatter(mtick.PercentFormatter())

ax2.plot(x_data_label, np.abs(test_statistics_2)*100, label="Forward 1-  
↪statistics")
ax2.axhline(y = test_martingale_threshold_mean*100, color = 'r', label="Success_  
↪threshold")
ax2.legend()
ax2.set_xlabel("time (months)")
ax2.set_ylabel("Absolute difference")
ax2.set_title('Difference in performance')
fig.tight_layout(h_pad=2)

ax2.yaxis.set_major_formatter(mtick.PercentFormatter())
fig.set_figwidth(8)
fig.set_figheight(6)

plt.show()
```



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3.3 Test 2; Success criteria

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

```
[814]: result2 = success_test(np.abs(test_statistics_2),
    ↪ test_martingale_threshold_max, test_martingale_threshold_mean)
```

Test passed

Test passed

EXTRA CONTENT

```
[816]: test2_diff_2 = forward_strategy_test(r_d_12_t, spot_2, spot_3, 2)
test2_diff_3 = forward_strategy_test(r_d_12_t, spot_3, spot_4, 3)

x_data_label_2 = range(0, test2_diff_2.shape[0], 1)
x_data_label_3 = range(0, test2_diff_3.shape[0], 1)

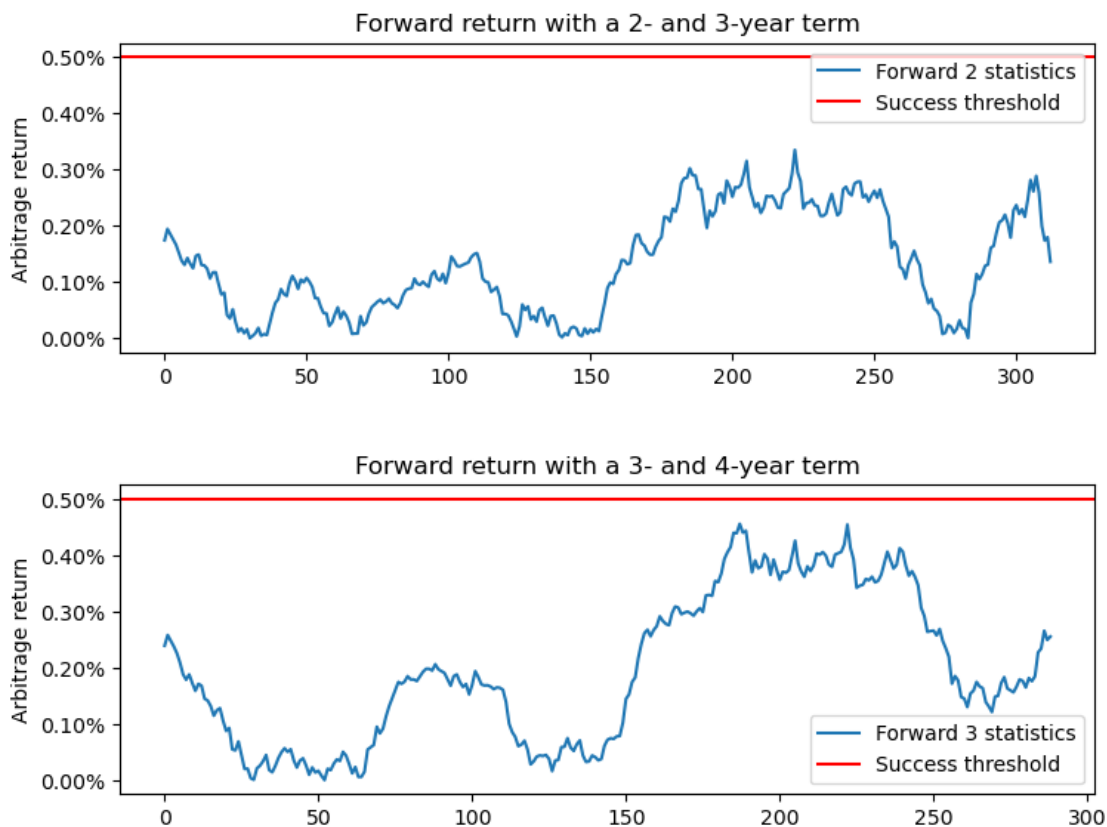
fig, (ax1, ax2) = plt.subplots(2, 1)
ax1.plot(x_data_label_2, np.abs(test2_diff_2).values*100, color='tab:blue',
    ↪ label="Forward 2 statistics")
```

```

ax1.axhline(y = test_martingale_threshold_mean*100, color = 'r', label="Success_
↳threshold")
ax1.set_title('Forward return with a 2- and 3-year term')
ax1.yaxis.set_major_formatter(mtick.PercentFormatter())
ax1.set_ylabel("Arbitrage return")
ax1.legend()
ax2.plot(x_data_label_3, np.abs(test2_diff_3).values*100, color='tab:blue',
↳label="Forward 3 statistics")
ax2.axhline(y = test_martingale_threshold_mean*100, color = 'r', label="Success_
↳threshold")
ax2.yaxis.set_major_formatter(mtick.PercentFormatter())
ax2.set_title('Forward return with a 3- and 4-year term')
ax2.set_ylabel("Arbitrage return")
ax2.legend()
fig.tight_layout(h_pad=2)
fig.set_figwidth(8)
fig.set_figheight(6)

plt.show()

```



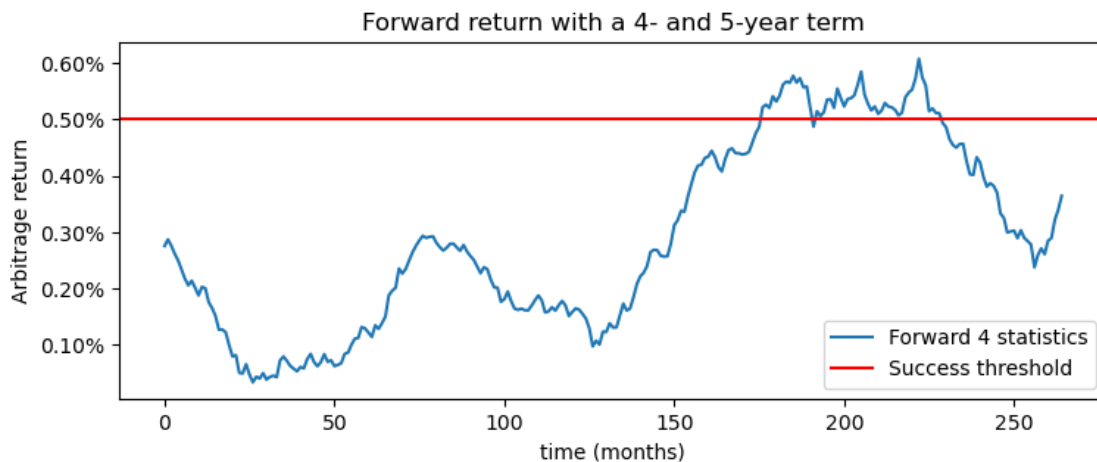
```
[817]: test2_diff_4 = forward_strategy_test(r_d_12_t, spot_4, spot_5, 4)

x_data_label_4 = range(0, test2_diff_4.shape[0], 1)

fig, ax3 = plt.subplots(1, 1)

ax3.plot(x_data_label_4, np.abs(test2_diff_4).values*100, color='tab:blue',
        label="Forward 4 statistics")
ax3.axhline(y = test_martingale_threshold_mean*100, color = 'r', label="Success
        threshold")
ax3.yaxis.set_major_formatter(mtick.PercentFormatter())
ax3.set_ylabel("Arbitrage return")
ax3.set_title('Forward return with a 4- and 5-year term')
ax3.set_xlabel("time (months)")
ax3.legend()
fig.tight_layout(h_pad=2)
fig.set_figwidth(8)
fig.set_figheight(3)

plt.show()
```



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3.4 Test 3; Buy-and-hold investment strategy

The Buy-and-hold strategy consists of investing all funds into ZCBs of a selected maturity, holding them until maturity and rolling all proceeds to buy another round of ZCB with the same selected maturity.

The strategy consists of investing for example 1 unit of currency in month k and withdrawing the

accrued payoff at time T.

k is one of the first $K * 12$ months where K is the term of the spot rate curve used.

This payoff can be replicated by investing the initial amount in a risk-free asset represented by the deflator asset at time k and withdrawing everything at the end time T.

If the strategy has a term of K years, $K * 12$ separate starting points need to be tested (to cover the entire curve). For each starting point, the performance can be calculated as:

$$Perf_{i,TK,k}^{BH}(t^{\wedge}) = \prod_{j=0}^{t^{\wedge}} ((1 + r_{i,TK}(k + 12Kj))^K$$

Where $t^{\wedge} = 1, 2, 3, \dots$. Note that t^{\wedge} represents year multiples of the selected term K .

Therefore:

$$DPerf_{i,TK,k}^{BH}(t^{\wedge}) = Perf_{i,TK,k}^{BH}(t^{\wedge}) * ZCB_{i,RF}(0, k + 12Kt^{\wedge})$$

is a Martingale and therefore the test verifies the following asymptotic identity:

$$E^{Q_k}[Perf_{i,TK,k}^{BH}(t^{\wedge})] = 1$$

Where:

- Q is the numeraire used in this test which is the terminal value of the deflator asset starting at time k

The function `deflator_normalisation()` normalizes the deflator asset such that the resulting series has a value of 1 at starting month k .

```
[823]: def deflator_normalisation(deflator, n):
        out = deflator.copy()
        for iRow in range(0, out.shape[0]):
            out.iloc[iRow, :] = out.iloc[iRow, :] / deflator.iloc[iRow % (n*12), :]
        ↪ #ToDo modulo in python
        out = out.iloc[n*12:, :]
        return out
```

The function `month_decomposition()` separates the time-series that starts at each month in a separate row, making it possible to compare them.

```
[825]: def month_decomposition(data, term):
        years = int(np.floor(data.shape[0] / (term*12))) - 1
        out = np.zeros((years, term*12))

        for iStart in range(0, years*(term*12), 1):
            position = iStart % (term*12)
            time = int(np.floor(iStart / (term*12)))
            out[time, position] = data[iStart]
```



```
return out
```

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Test 3.1; Spot rate 1-year term

$$Perf_{i,T1,k}^{BH}(t^{\wedge}) = \prod_{j=0}^{t^{\wedge}} (1 + r_{i,T1}(k + 12 * j))$$

Where $t^{\wedge} = 1, 2, 3, \dots$ and $k = 0, 1, 2, \dots, 11$

Therefore

$$DPerf_{i,T1,k}^{BH}(t^{\wedge}) = Perf_{i,T1,k}^{BH}(t^{\wedge}) * ZCB_{i,RF}(0, k + t^{\wedge} * 12)$$

is a Martingale and therefore the test verifies the following asymptotic identity

$$E^{Q_k}[Perf_{i,T1,k}^{BH}(t^{\wedge})] = 1$$

The two test statistics used are:

$$S_{AVERAGE} = \frac{1}{T^{\wedge}} \sum_{t^{\wedge}=0}^{T^{\wedge}} \frac{1}{N} \left| \sum_{i=0}^N (DPerf_{i,T1,k}^{BH}(t^{\wedge}) - 1) \right|$$

$$S_{MAX} = \max_{t^{\wedge}} \frac{1}{N} \left| \sum_{i=0}^N (DPerf_{i,T1,k}^{BH}(t^{\wedge}) - 1) \right|$$

```
[829]: deflator_1 = deflator_normalisation(deflator, 1)
      MPerf1 = pd.DataFrame(spot_1_cum.values*deflator_1.values)
```

```
[830]: test_statistics_3_1 = (MPerf1 - 1).mean(axis=1)
```

VISUAL INSPECTION

1-year Buy-and-hold strategy test

```
[832]: MPerf1.head()
```

```
[832]:
```

	0	1	2	3	4	5	6	\
0	1.004562	0.999219	0.991150	1.003348	0.993555	1.004531	0.987302	
1	1.003213	0.997795	0.993088	0.999417	0.987495	1.013388	0.983262	
2	1.001956	0.995144	0.987677	1.002524	0.990338	1.009810	0.991778	
3	0.998350	0.994125	0.987555	1.006513	0.987055	1.005991	0.994121	
4	0.996157	0.995422	0.986971	1.007133	0.994106	1.004390	0.997618	
	7	8	9	...	490	491	492	493 \

0	1.006627	0.996127	1.003108	...	1.005866	0.989069	0.992759	1.006712
1	1.005822	0.995352	1.007171	...	1.000586	0.990931	0.991782	1.003596
2	1.000620	0.994519	1.011014	...	1.003072	0.987275	0.988319	1.004111
3	1.002615	0.991917	1.012795	...	0.998028	0.987896	0.984851	1.013453
4	0.992355	0.995912	1.014328	...	0.998254	0.990771	0.992474	1.015249

	494	495	496	497	498	499
0	1.004969	1.001652	1.004612	1.002714	1.007048	1.015211
1	1.006347	0.996530	1.004242	1.005712	1.006707	1.014475
2	0.986631	0.995712	0.999514	1.003652	0.997309	1.012315
3	0.981520	0.993935	1.000180	1.002491	0.997266	1.009101
4	0.984867	0.989628	0.995137	0.995261	0.999849	1.012831

[5 rows x 500 columns]

```
[833]: plot_strategy_1 = month_decomposition(test_statistics_3_1, 1)
```

```
[834]: x_data_label_1 = range(0,plot_strategy_1.shape[0],1)
x_data_label_2 = range(0,test_statistics_3_1.shape[0],1)

fig, (ax1,ax2) = plt.subplots(2,1)
ax1.plot(x_data_label_1, plot_strategy_1*100)
ax1.yaxis.set_major_formatter(mtick.PercentFormatter())
ax1.set_xlabel("time (months)")
ax1.set_ylabel("Arbitrage return")
ax1.set_title('Discounted 1-year Buy-and-hold strategy return')

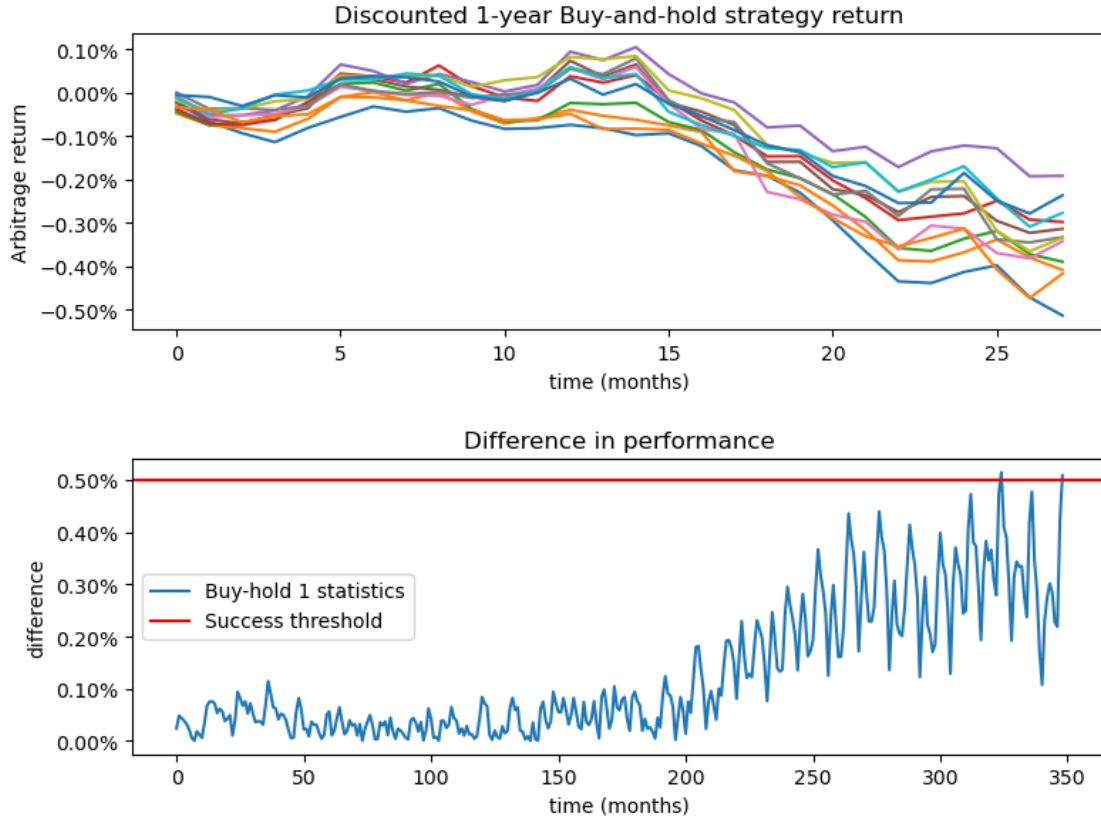
ax2.plot(x_data_label_2, np.abs(test_statistics_3_1)*100, label="Buy-hold 1_
↳statistics")
ax2.axhline(y = test_martingale_threshold_mean*100, color = 'r', label="Success_
↳threshold")

ax2.set_xlabel("time (months)")
ax2.set_ylabel("difference")
ax2.set_title('Difference in performance')
fig.tight_layout(h_pad=2)
ax2.legend()
ax2.yaxis.set_major_formatter(mtick.PercentFormatter())

fig.set_figwidth(8)
fig.set_figheight(6)

fig.tight_layout(h_pad=2)

plt.show()
```



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3.5 Test 3.1; Success criteria

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

```
[836]: result31 = success_test(np.abs(test_statistics_3_1),
    ↪ test_martingale_threshold_max, test_martingale_threshold_mean)
```

Test passed

Test passed

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Test 3.2; Spot rate 2-year term

$$Perf_{i,T2,k}^{BH}(t^{\wedge}) = \prod_{j=0}^{t^{\wedge}} ((1 + r_{i,T2}(k + 24j))^2$$

Where $t^{\wedge} = 1, 2, 3, \dots$ and $k = 0, 1, 2, \dots, 23$.

Therefore

$$DPerf_{i,T2,k}^{BH}(t^{\setminus}) = Perf_{i,T2,k}^{BH}(t^{\setminus}) * ZCB_{i,RF}(0, k + 24t^{\setminus})$$

is a Martingale and therefore the test verifies the following asymptotic identity:

$$E^{Q_k}[Perf_{i,T2,k}^{BH}(t^{\setminus})] = 1$$

The two test statistics used are:

$$S_{AVERAGE} = \frac{1}{T-24} \sum_{t=0}^T \left| \frac{1}{N} \sum_{i=0}^N (DPerf_{i,T2,k}^{BH}(t) - 1) \right|$$

$$S_{MAX} = \max_t \left| \frac{1}{N} \sum_{i=0}^N (DPerf_{i,T2,k}^{BH}(t) - 1) \right|$$

```
[841]: deflator_2 = deflator_normalisation(deflator, 2)
      MPerf2 = pd.DataFrame(spot_2_cum.values*deflator_2.values)
```

```
[842]: test_statistics_3_2 = (MPerf2 - 1).mean(axis=1)
```

VISUAL INSPECTION

2-year Buy-and-hold strategy test

```
[844]: MPerf2.head()
```

```
[844]:
```

	0	1	2	3	4	5	6	\
0	0.987804	0.995772	0.958710	1.023029	0.966309	1.024074	0.957349	
1	0.986066	0.992620	0.965574	1.017481	0.956770	1.039609	0.950018	
2	0.984107	0.987466	0.958412	1.026432	0.963630	1.030374	0.966157	
3	0.977909	0.985546	0.961410	1.036239	0.960092	1.020917	0.971313	
4	0.975116	0.987340	0.962555	1.039016	0.976373	1.015152	0.978943	
	7	8	9	...	490	491	492	493 \
0	1.010823	1.005664	1.022176	...	1.017128	0.972413	0.986842	1.015954
1	1.010024	1.005465	1.026817	...	1.008156	0.978269	0.985488	1.006890
2	1.000004	1.007448	1.031612	...	1.013869	0.973820	0.978974	1.004639
3	1.003374	1.004735	1.033812	...	1.004674	0.978008	0.972903	1.020778
4	0.981117	1.014926	1.036062	...	1.005777	0.986106	0.988721	1.023611
	494	495	496	497	498	499		
0	0.993610	0.990125	0.999217	0.996569	1.014536	1.044291		
1	0.995958	0.981113	0.998096	1.003136	1.013737	1.040211		
2	0.957384	0.979795	0.988150	0.999873	0.995640	1.034115		
3	0.948895	0.977734	0.988632	0.997856	0.996031	1.026409		

```
4  0.957155  0.972586  0.977612  0.986324  1.001435  1.032183
```

```
[5 rows x 500 columns]
```

```
[845]: plot_strategy_2 = month_decomposition(test_statistics_3_2, 2)
```

```
[846]: x_data_label_1 = range(0,plot_strategy_2.shape[0],1)
x_data_label_2 = range(0,test_statistics_3_2.shape[0],1)

fig, (ax1, ax2) = plt.subplots(2,1)
ax1.plot(x_data_label_1, plot_strategy_2)
ax1.yaxis.set_major_formatter(mtick.PercentFormatter())

ax1.set_xlabel("time (months)")
ax1.set_ylabel("Arbitrage return")
ax1.set_title('Discounted 2-year Buy-and-hold strategy return')

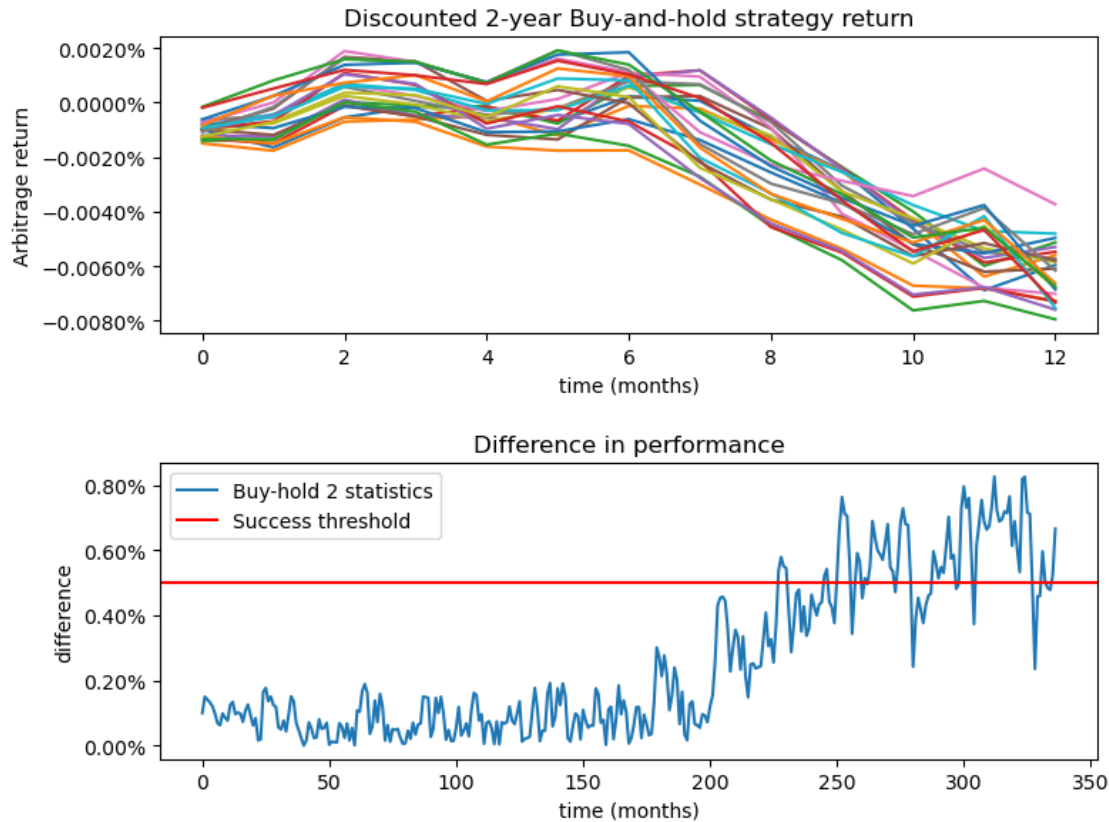
ax2.plot(x_data_label_2, np.abs(test_statistics_3_2)*100, label="Buy-hold 2-
↳statistics")
ax2.axhline(y = test_martingale_threshold_mean*100, color = 'r', label="Success-
↳threshold")

ax2.set_xlabel("time (months)")
ax2.set_ylabel("difference")
ax2.set_title('Difference in performance')
fig.tight_layout(h_pad=2)
ax2.legend()
ax2.yaxis.set_major_formatter(mtick.PercentFormatter())

fig.set_figwidth(8)
fig.set_figheight(6)

fig.tight_layout(h_pad=2)

plt.show()
```



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3.6 Test 3.2; Success criteria

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

```
[848]: result32 = success_test(np.abs(test_statistics_3_2),
    ↪ test_martingale_threshold_max, test_martingale_threshold_mean)
```

Test passed

Test passed

3.6.1 Extra content

```
[850]: deflator_3 = deflator_normalisation(deflator, 3)
MPerf3 = pd.DataFrame(spot_3_cum.values*deflator_3.values)
test3_diff_3 = (MPerf3 - 1).mean(axis=1)
plot_strategy_3 = month_decomposition(test3_diff_3, 3)
```

```
[851]: deflator_4 = deflator_normalisation(deflator, 4)
MPerf4 = pd.DataFrame(spot_4_cum.values*deflator_4.values)
test3_diff_4 = (MPerf4 - 1).mean(axis=1)
```

```
plot_strategy_4 = month_decomposition(test3_diff_4, 4)
```

```
[852]: deflator_5 = deflator_normalisation(deflator,5)
MPerf5 = pd.DataFrame(spot_5_cum.values*deflator_5.values)
test3_diff_5 = (MPerf5 - 1).mean(axis=1)
plot_strategy_5 = month_decomposition(test3_diff_5, 5)
```

```
[853]: x_data_label_3 = range(0,plot_strategy_3.shape[0],1)
x_data_label_4 = range(0,plot_strategy_4.shape[0],1)
x_data_label_5 = range(0,plot_strategy_5.shape[0],1)

fig, (ax1, ax2, ax3) = plt.subplots(3,1)

ax1.plot(x_data_label_3, np.abs(plot_strategy_3)*100)
ax1.set_title('Discounted 3-year Buy-and-hold strategy return')
ax1.yaxis.set_major_formatter(mtick.PercentFormatter())

ax2.plot(x_data_label_4, np.abs(plot_strategy_4)*100)
ax2.yaxis.set_major_formatter(mtick.PercentFormatter())
ax2.set_ylabel("Absolute difference")
ax2.set_title('Discounted 4-year Buy-and-hold strategy return')

ax3.plot(x_data_label_5, np.abs(plot_strategy_5)*100)
ax3.yaxis.set_major_formatter(mtick.PercentFormatter())
ax3.set_title('Discounted 5-year Buy-and-hold strategy return')

fig.tight_layout(h_pad=2)

fig.set_figwidth(8)
fig.set_figheight(6)

plt.show()
```



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3.7 Test 4; Total-return index

The total-return index in a risk-neutral framework has on average the same performance as the stochastic deflator asset.

$$DPerf_{i,1}^I(t) = I_{i,1}(t) * ZCB_{i,RF}(0, t)$$

Therefore, the performance of the index is a Martingale if the deflator asset is used as the numeraire:

$$E^Q[I_{i,1}(t)] \stackrel{?}{=} 1$$

Where:

- Q is the numeraire used in this test which is the deflator asset at time t

The two test statistics used are:

$$S_{AVERAGE} = \frac{1}{T} \sum_{t=0}^T \frac{1}{N} \left| \sum_{i=0}^N DPerf_{i,1}^I(t) - 1 \right|$$

$$S_{MAX} = \max_t \frac{1}{N} \left| \sum_{i=0}^N (DPerf_{i,1}^I(t) - 1) \right|$$

```
[858]: test_statistics_4 = pd.DataFrame(first_index.values * deflator.values).
      ↪mean(axis=1)-1
```

```
[860]: x_data_label = range(0,test_statistics_4.shape[0],1)

fig, (ax1,ax2) = plt.subplots(2,1)
ax1.plot(x_data_label, test_statistics_4.values*100, color='tab:blue')

ax1.set_ylabel("Arbitrage return")
ax1.set_title('Discounted 2-year Buy-and-hold strategy return')
ax1.yaxis.set_major_formatter(mtick.PercentFormatter())

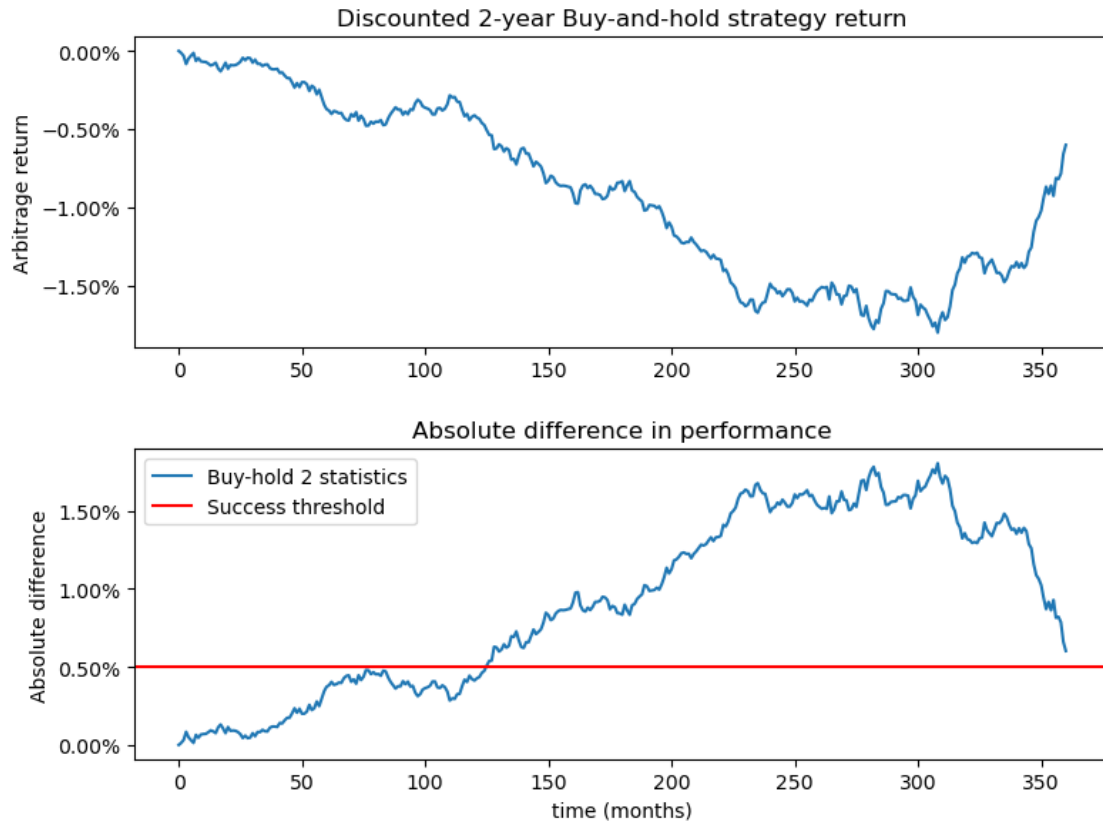
ax2.plot(x_data_label, np.abs(test_statistics_4)*100, label="Buy-hold 2_
      ↪statistics")
ax2.axhline(y = test_martingale_threshold_mean*100, color = 'r', label="Success_
      ↪threshold")

ax2.set_xlabel("time (months)")
ax2.set_ylabel("Absolute difference")
ax2.set_title('Absolute difference in performance')
fig.tight_layout(h_pad=2)
ax2.legend()
ax2.yaxis.set_major_formatter(mtick.PercentFormatter())

fig.set_figwidth(8)
fig.set_figheight(6)

fig.tight_layout(h_pad=2)

plt.show()
```



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3.8 Test 4; Success criteria

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

```
[862]: result4 = success_test(np.abs(test_statistics_4),
    ↪ test_martingale_threshold_max, test_martingale_threshold_mean)
```

Test failed

Test failed

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3.9 Conclusion

This script conducted tests on the output of the stochastic scenario output. The files selected are:

```
[864]: file_det
```

```
[864]: 'TEST_DATA_DETERMINISTIC.csv'
```

```
[865]: file_stoch
```

```
[865]: 'TEST_DATA_STOCHASTIC.csv'
```

The currency selected for this run is

```
[867]: currency
```

```
[867]: 'EUR'
```

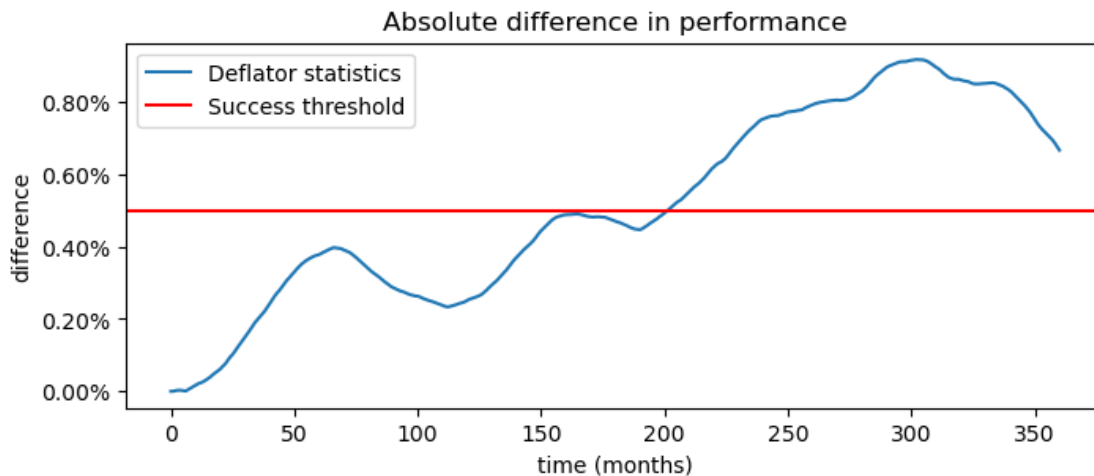
Test 1 compares the average performance of the stochastic deflator asset compared with the deterministic deflator asset.

```
[869]: x_data_label = range(0,deflator_det.shape[0],1)
fig, ax = plt.subplots(1,1)
ax.plot(x_data_label, test_statistics_1*100, label="Deflator statistics")
ax.axhline(y = test_martingale_threshold_mean*100, color = 'r', label="Success_
↪threshold")

ax.set_xlabel("time (months)")
ax.set_ylabel("difference")
ax.set_title('Absolute difference in performance')
ax.legend()
ax.yaxis.set_major_formatter(mtick.PercentFormatter())

fig.set_figwidth(8)
fig.set_figheight(3)

plt.show()
```

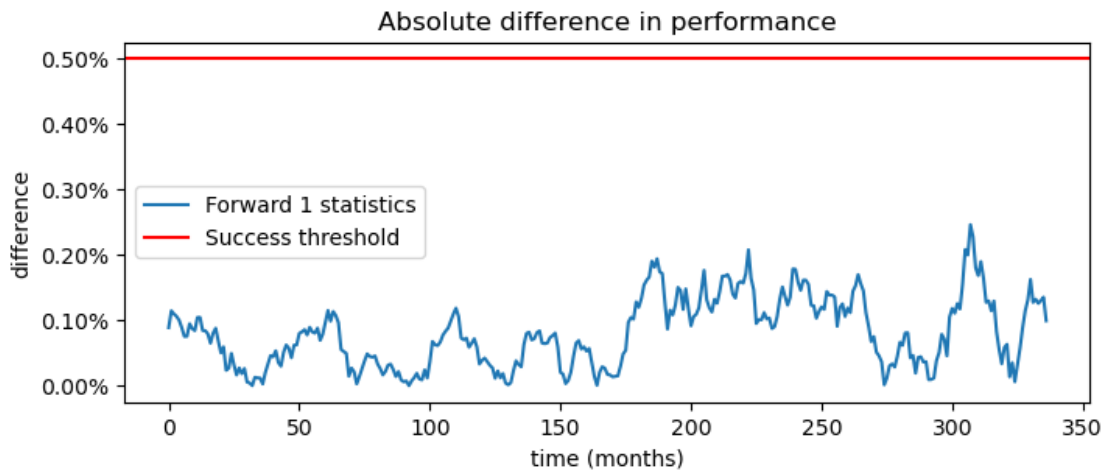


Test 2 performs a Martingale test on a specific investment strategy using spot curves. The Forward return with a spread of 1-year strategy using 1- and 2-year spot rates consists of buying and selling ZCB bonds with 1- and 2-year term. In this example, this consists of buying (long position) a ZCB with a maturity of 2 years and selling (short position) a ZCB with a maturity of 1 year.

```
[871]: x_data_label = range(0,test_statistics_2.shape[0],1)
fig, ax = plt.subplots(1,1)
ax.plot(x_data_label, np.abs(test_statistics_2)*100, label="Forward 1_
↳statistics")
ax.axhline(y = test_martingale_threshold_mean*100, color = 'r', label="Success_
↳threshold")
ax.legend()
ax.set_xlabel("time (months)")
ax.set_ylabel("difference")
ax.set_title('Absolute difference in performance')

ax.yaxis.set_major_formatter(mtick.PercentFormatter())
fig.set_figwidth(8)
fig.set_figheight(3)

plt.show()
```



Tests 3.1 and 3.2 perform a Martingale test on the second financial strategy using spot rate curves. This Buy-and-hold strategy consists of investing all funds into ZCBs of a selected maturity, holding them until maturity and rolling all proceeds to buy another round of ZCB with the same selected maturity.

Test 3.1 uses the 1-year spot curve.

```
[874]: x_data_label_2 = range(0,test_statistics_3_1.shape[0],1)

fig, ax = plt.subplots(1, 1)

ax.plot(x_data_label_2, np.abs(test_statistics_3_1)*100, label="Buy-hold 1_
↳statistics")
```

```

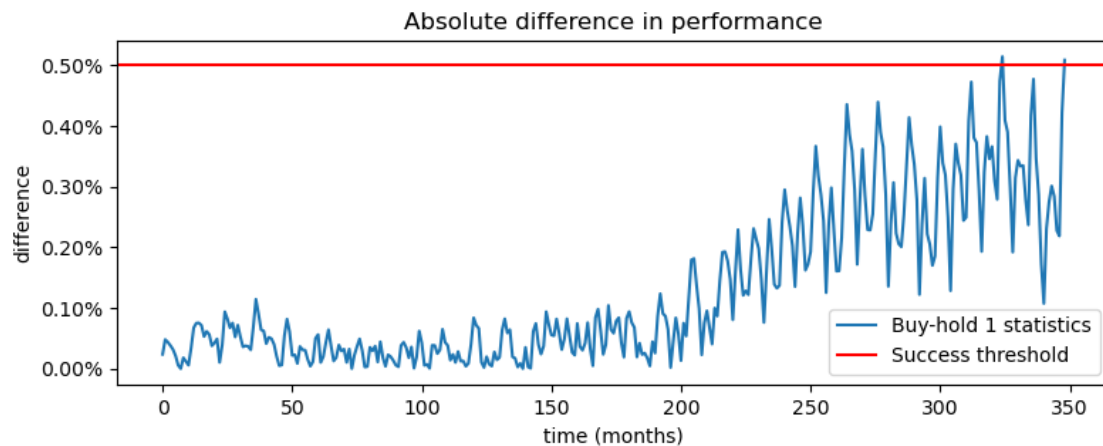
ax.axhline(y = test_martingale_threshold_mean*100, color = 'r', label="Success_
↳threshold")
ax.legend()
ax.set_xlabel("time (months)")
ax.set_ylabel("difference")
ax.set_title('Absolute difference in performance')
fig.tight_layout(h_pad=2)

ax.yaxis.set_major_formatter(mtick.PercentFormatter())

fig.set_figwidth(8)
fig.set_figheight(3)

plt.show()

```



Test 3.2 uses the 2-year spot curve.

```

[876]: x_data_label_2 = range(0, test_statistics_3_2.shape[0], 1)

fig, ax = plt.subplots(1, 1)

ax.plot(x_data_label_2, np.abs(test_statistics_3_2)*100, label="Buy-hold 2_
↳statistics")
ax.axhline(y = test_martingale_threshold_mean*100, color = 'r', label="Success_
↳threshold")

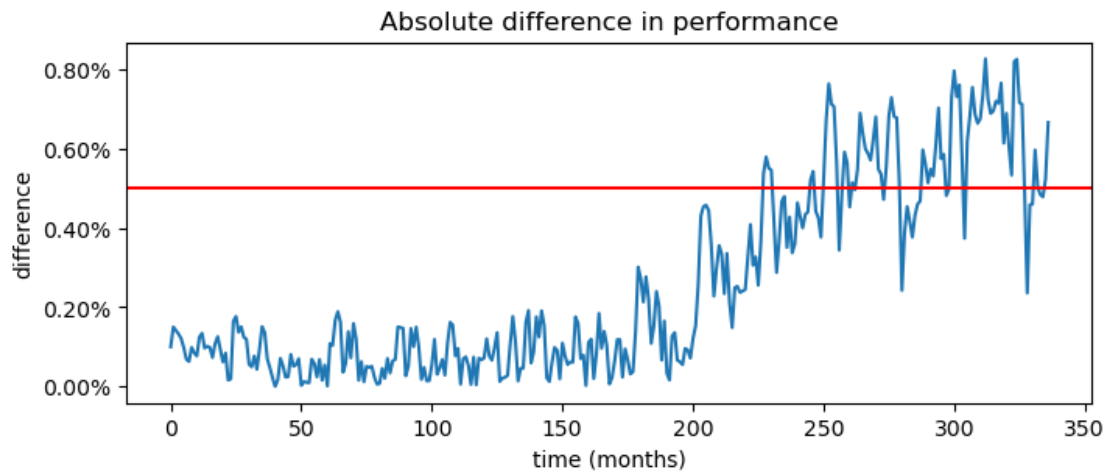
ax.set_xlabel("time (months)")
ax.set_ylabel("difference")
ax.set_title('Absolute difference in performance')

ax.yaxis.set_major_formatter(mtick.PercentFormatter())

```

```
fig.set_figwidth(8)
fig.set_figheight(3)

plt.show()
```



Test 4 performs the Martingale test using the total-return index. This test checks if the performance of the index using the deflator as numeraire is a Martingale.

```
[878]: x_data_label = range(0, test_statistics_4.shape[0], 1)

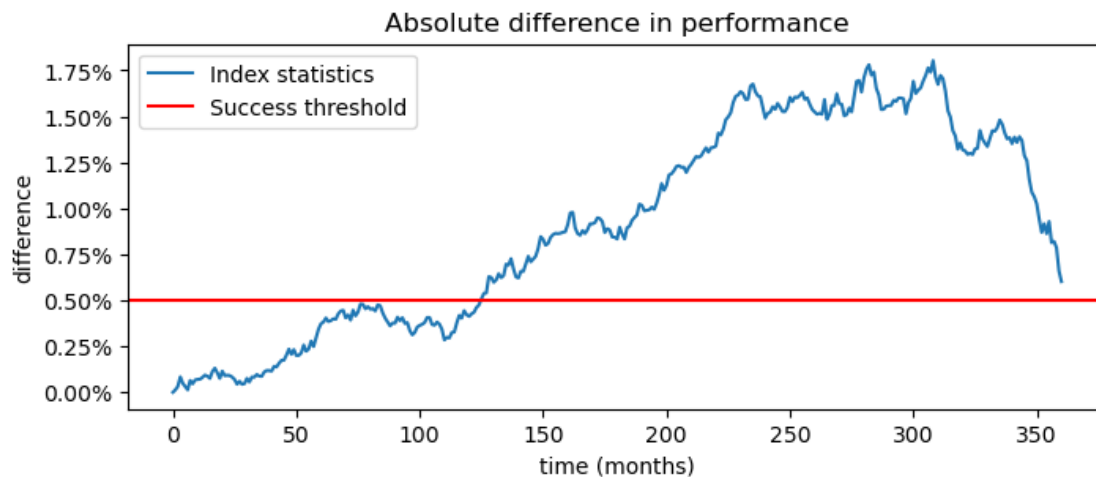
fig, ax = plt.subplots(1, 1)

ax.plot(x_data_label, np.abs(test_statistics_4)*100, label="Index statistics")
ax.axhline(y = test_martingale_threshold_mean*100, color = 'r', label="Success_
↳threshold")

ax.set_xlabel("time (months)")
ax.set_ylabel("difference")
ax.set_title('Absolute difference in performance')
ax.legend()
ax.yaxis.set_major_formatter(mtick.PercentFormatter())

fig.set_figwidth(8)
fig.set_figheight(3)

plt.show()
```



Given the success criteria, the tests show the following success rate:

```
[881]: pd.DataFrame(data = [result1, result2, result31, result32, result4], columns = ["Mean test", "Max test"], \
    index= ["Deflator/deterministic", "Martingale Forward return 12", \
    "Martingale Buy-and-hold 1", \
    "Martingale Buy-and-hold 2", "Martingale Total-return \
    index"])
```

```
[881]:
```

	Mean test	Max test
Deflator/deterministic	True	True
Martingale Forward return 12	True	True
Martingale Buy-and-hold 1	True	True
Martingale Buy-and-hold 2	True	True
Martingale Total-return index	False	False