



## Book Presentation at:



**MATHEMATICS OF FINANCE MA PROGRAM**  
DEPARTMENT OF MATHEMATICS, COLUMBIA UNIVERSITY

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A quantitative methodology for risk assessment in financial products

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New York, 14th November 2012

Opinions expressed in this work are exclusively of the author

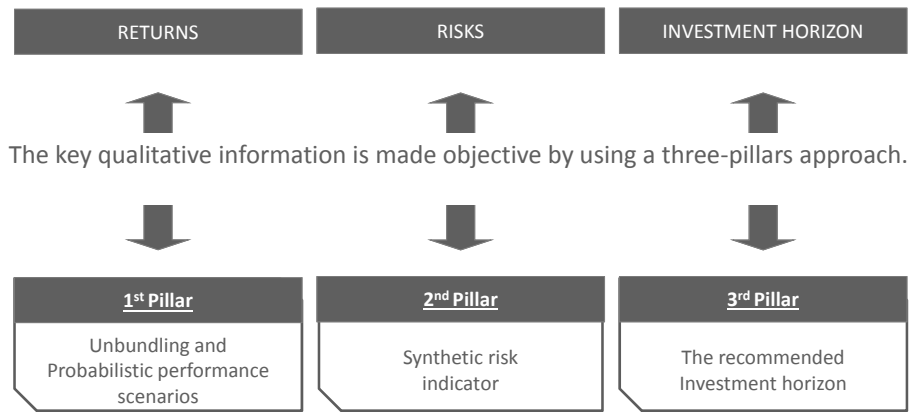


## Syllabus

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- Preliminaries: the three pillars
- Unbundling and Probabilistic performance scenarios
- Synthetic risk indicator
- The optimal time horizon
- An Application of the methodology

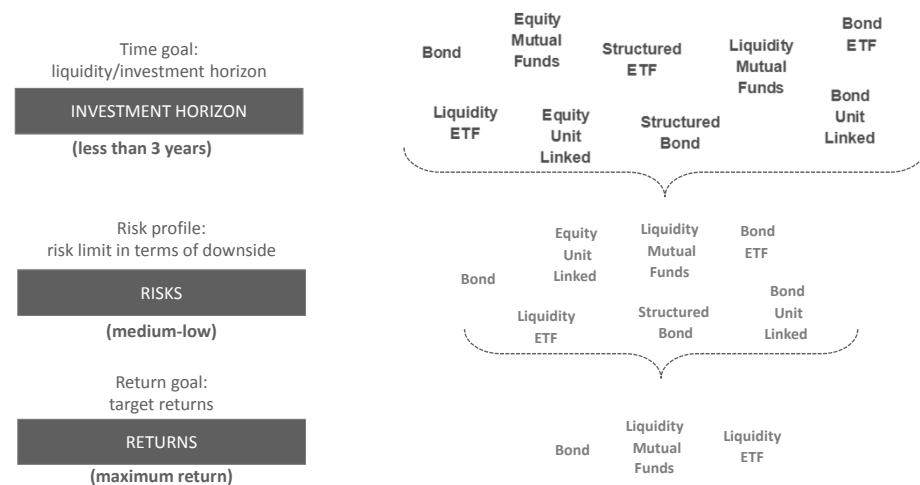
## Preliminaries



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## Preliminaries

These metrics provide a guide in the interpretation of complex information relating to financial products, by supporting the process of selection by means of a sequential filtering procedure:



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## Syllabus

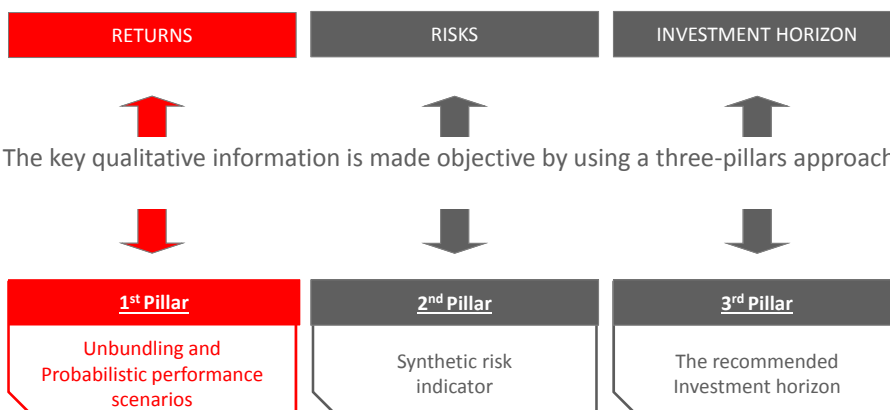
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- Preliminaries: the three pillars
- Unbundling and Probabilistic performance scenarios
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- The optimal time horizon
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## Unbundling and Probabilistic performance scenarios

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## Unbundling and Probabilistic performance scenarios

The returns evaluation requires the estimate of all the relevant risk factors connected with the financial structure of each product



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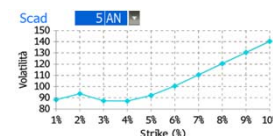
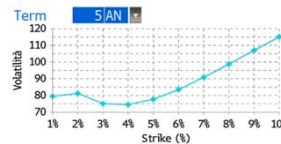
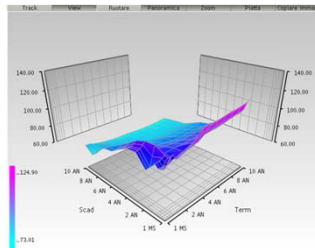
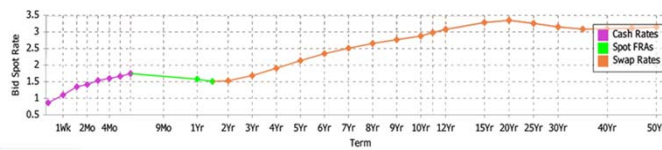
## Unbundling and Probabilistic performance scenarios

### DEFAULTABLE BOND



Interest Rate  
Volatility  
Significant exposure to credit risk

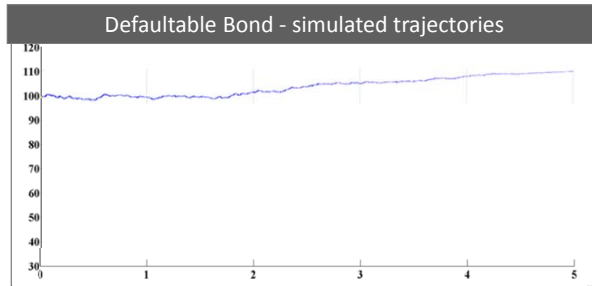
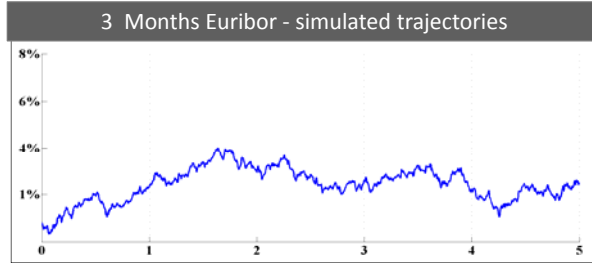
Markets data are used to estimate the relevant risk factors connected with the financial structure of the product



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## Unbundling and Probabilistic performance scenarios

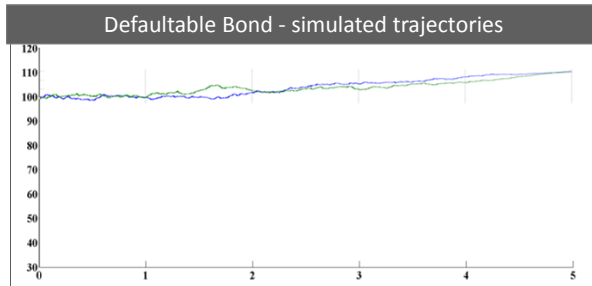
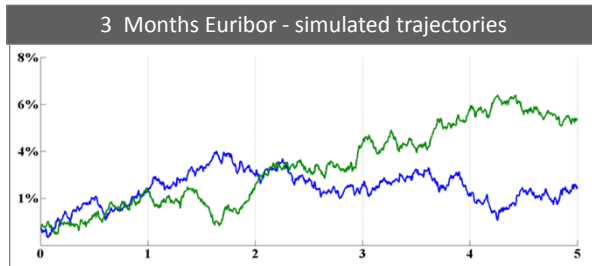
DEFAULTABLE  
BOND



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## Unbundling and Probabilistic performance scenarios

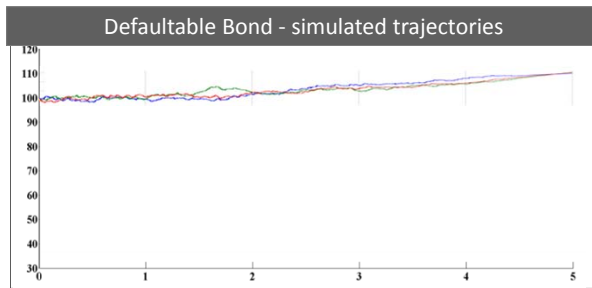
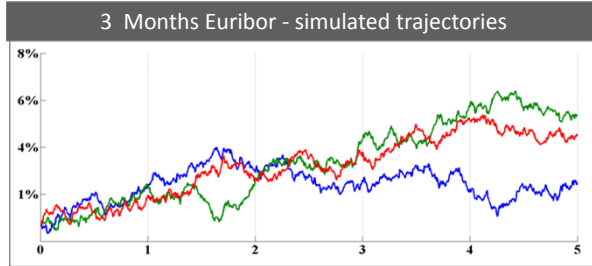
DEFAULTABLE  
BOND



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# Unbundling and Probabilistic performance scenarios

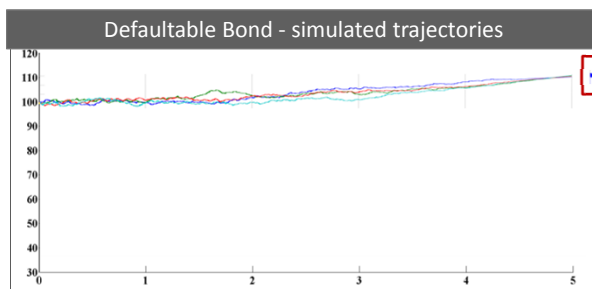
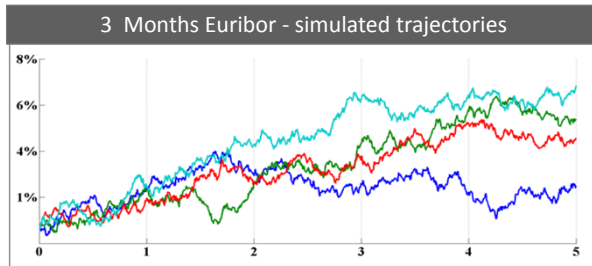
DEFAULTABLE  
BOND



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# Unbundling and Probabilistic performance scenarios

DEFAULTABLE  
BOND



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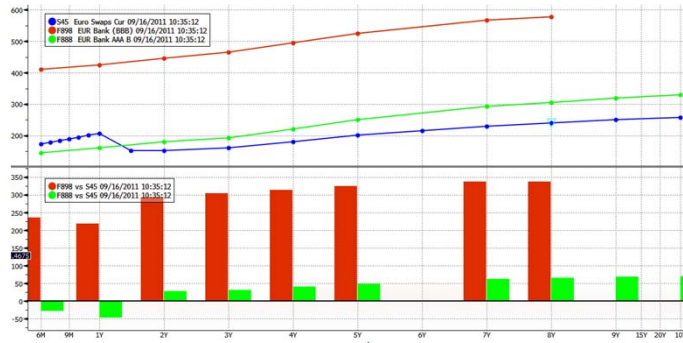
## Unbundling and Probabilistic performance scenarios

DEFAULTABLE  
BOND



Interest  
Rate  
Volatility  
**Significant  
exposure  
to credit risk**

Markets data are used to estimate the relevant risk factors connected with the financial structure of the product



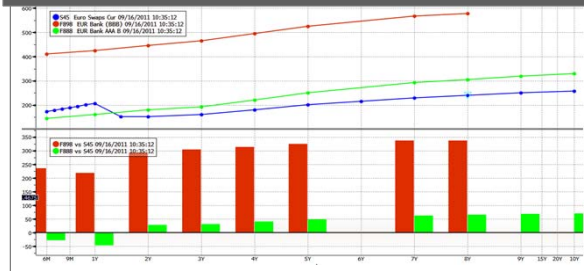
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## Unbundling and Probabilistic performance scenarios

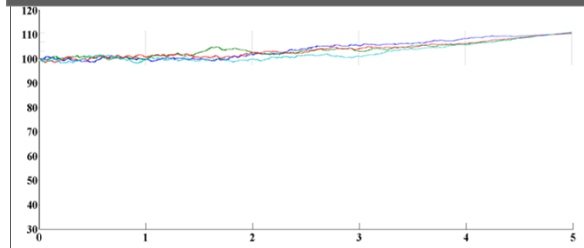
DEFAULTABLE  
BOND



Yield curve for different issuers



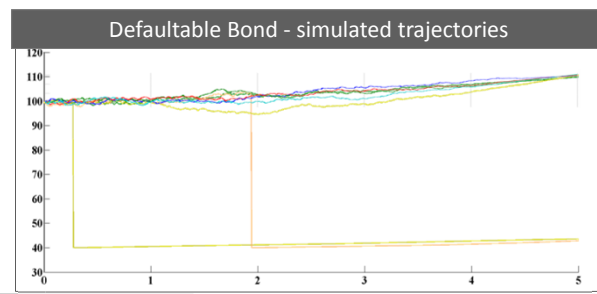
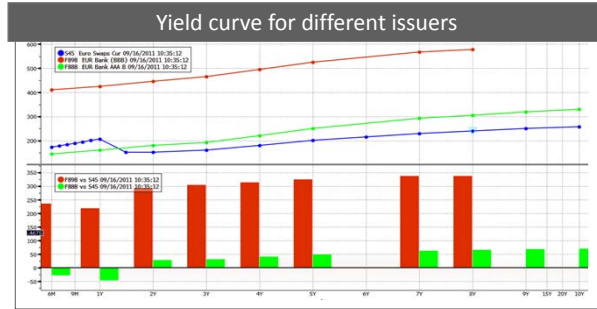
Defaultable Bond - simulated trajectories



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# Unbundling and Probabilistic performance scenarios

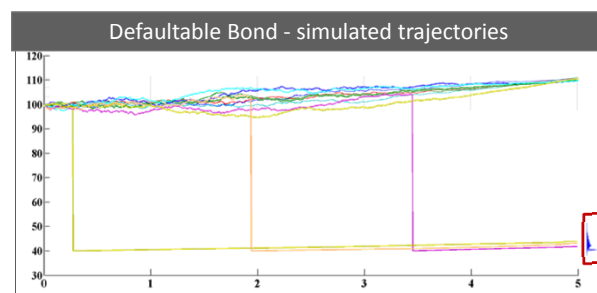
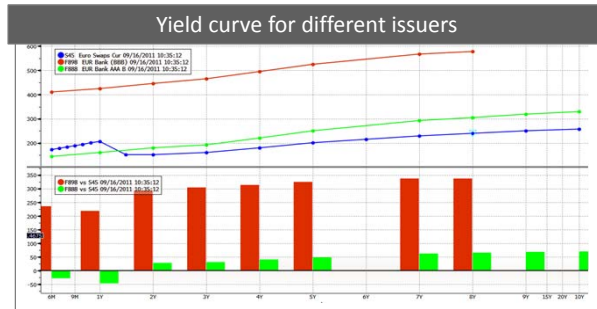
DEFAULTABLE  
BOND



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# Unbundling and Probabilistic performance scenarios

DEFAULTABLE  
BOND



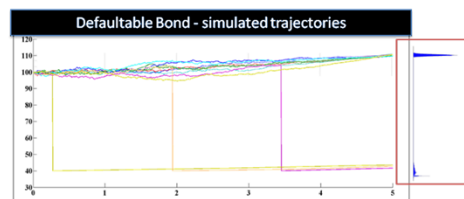
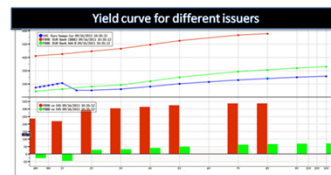
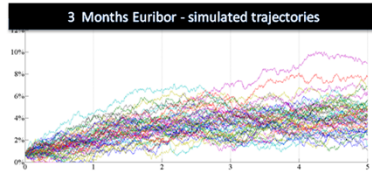
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## Unbundling and Probabilistic performance scenarios

DEFAULTABLE  
BOND

The risk factors define the product values over time and at expiry date

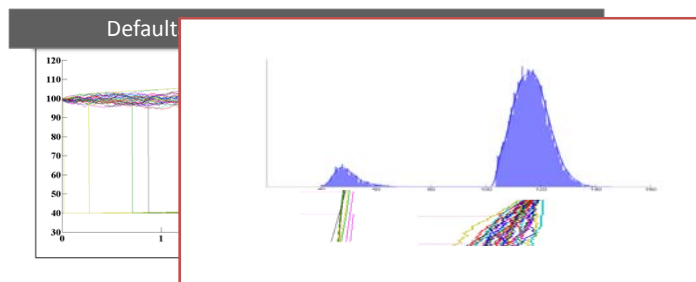


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## Unbundling and Probabilistic performance scenarios

DEFAULTABLE  
BOND

The final values of the product provide the probability distribution of the potential returns (so-called *pricing at maturity*)...



**Possible Outcomes**

*Pricing at maturity*

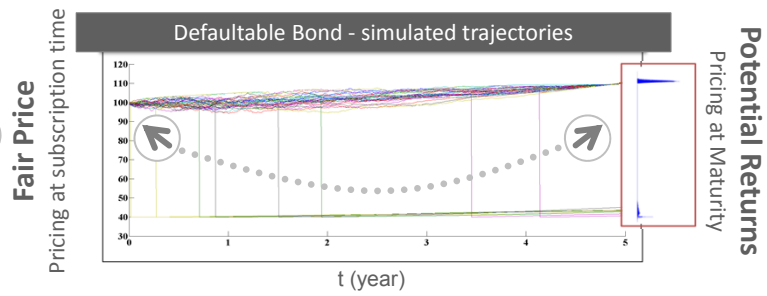
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## Unbundling and Probabilistic performance scenarios

### DEFAULTABLE BOND



... the "fair value" of the product at the issue date is obtained, like in the *best practice* of the pricing procedures of intermediaries, by evaluating the expected discounted value of this distribution.



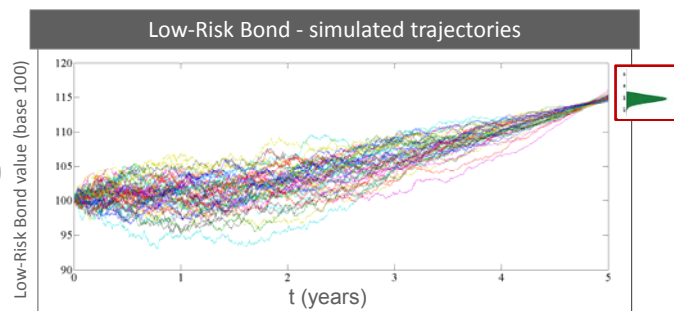
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## Unbundling and Probabilistic performance scenarios

### LOW-RISK BOND



Limited exposure to credit risk corresponds to a lower (or zero) number of trajectories incurring in a *default event*.

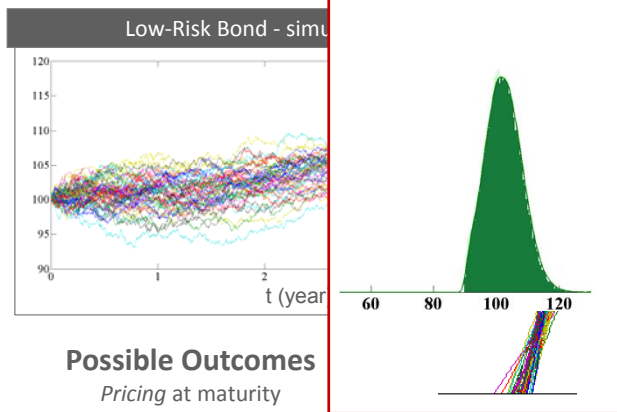


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## Unbundling and Probabilistic performance scenarios

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21

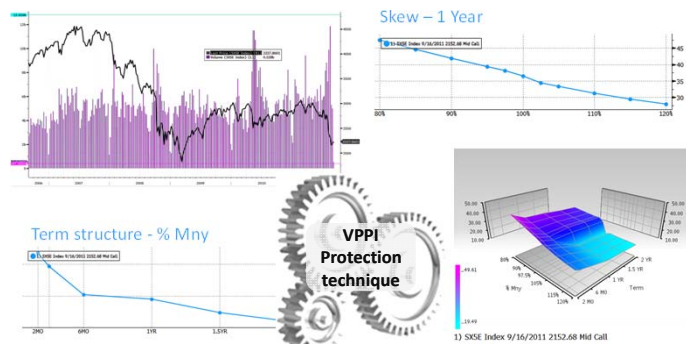
## Unbundling and Probabilistic performance scenarios

### VPPI PRODUCT

Markets data are used to estimate the relevant risk factors connected with the financial structure of the product



Interest Rate  
Rate  
Volatility  
Limited exposure  
to Market risk



VPPI technique is aimed at protecting the initial value of the financial investment over a specified time horizon and obtaining possible gains by limited exposure to the equity markets.

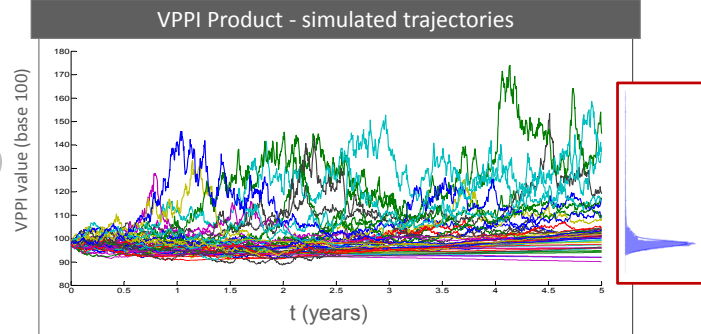
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## Unbundling and Probabilistic performance scenarios

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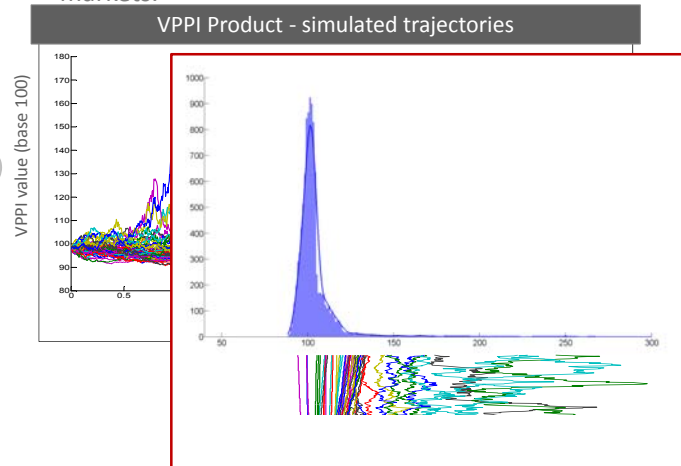
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## Unbundling and Probabilistic performance scenarios

### VPPI PRODUCT



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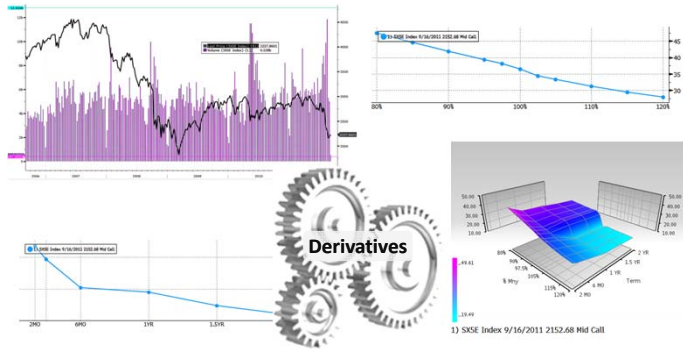
## Unbundling and Probabilistic performance scenarios

### INDEX LINKED CERTIFICATE



Significant exposure to Market risk

Markets data are used to estimate the relevant risk factors connected with the financial structure of the product



The index-linked certificate is characterised by a complex financial engineering that makes intensive use of different derivatives components. These derivatives link the performances of the product to the variability of an equity index.

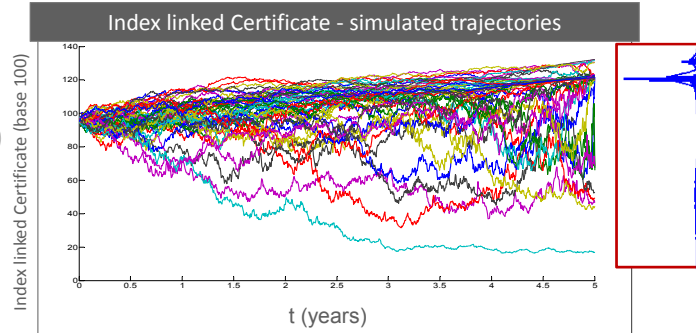
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## Unbundling and Probabilistic performance scenarios

### INDEX LINKED CERTIFICATE



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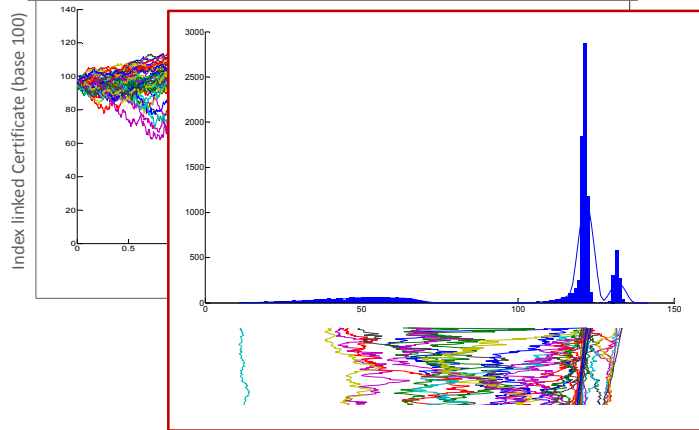
## Unbundling and Probabilistic performance scenarios

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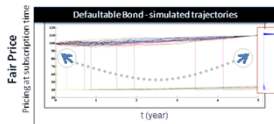
Index linked Certificate - simulated trajectories



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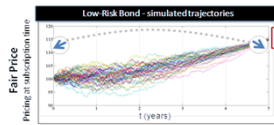
## Unbundling and Probabilistic performance scenarios

### DEFAULTABLE BOND



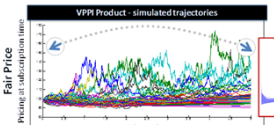
Potential Returns Pricing at maturity

### LOW-RISK BOND



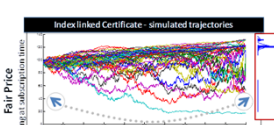
Potential Returns Pricing at maturity

### VPPI PRODUCT



Potential Returns Pricing at maturity

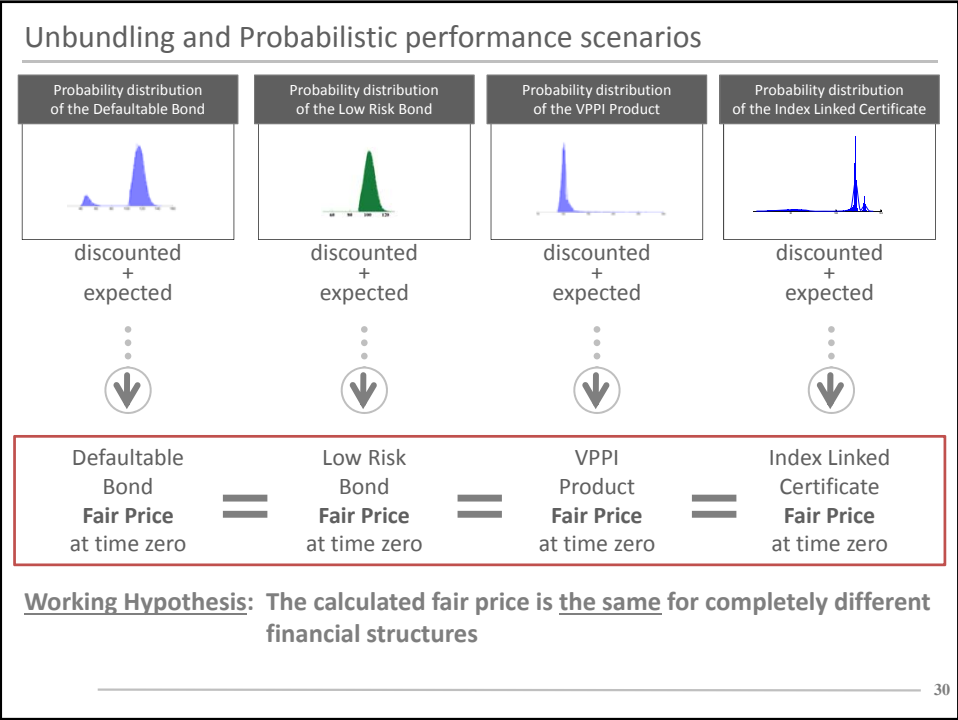
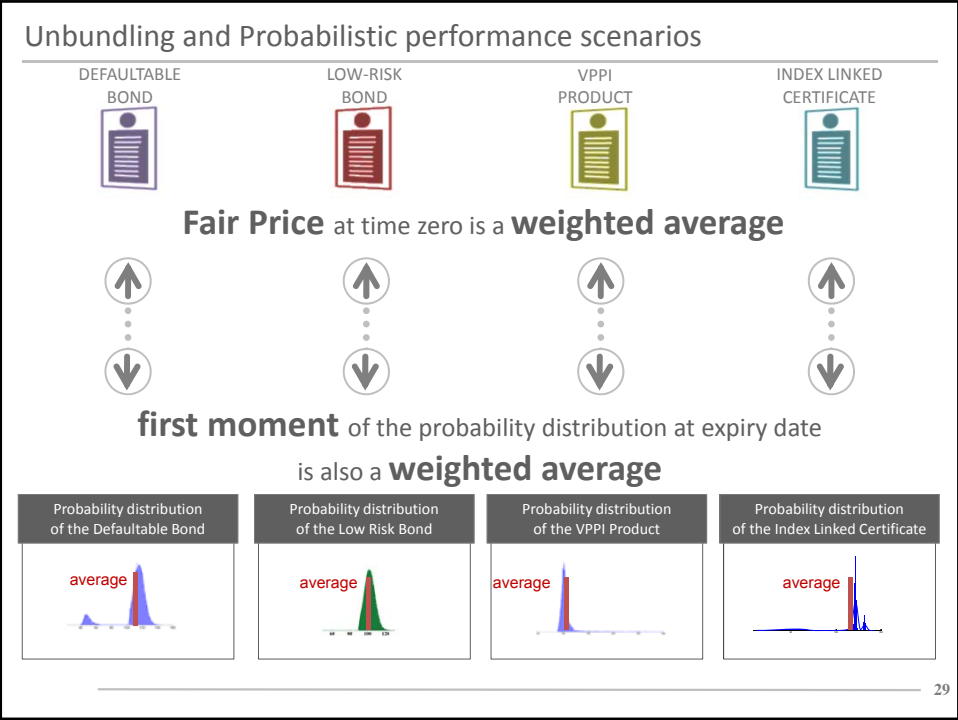
### INDEX LINKED CERTIFICATE



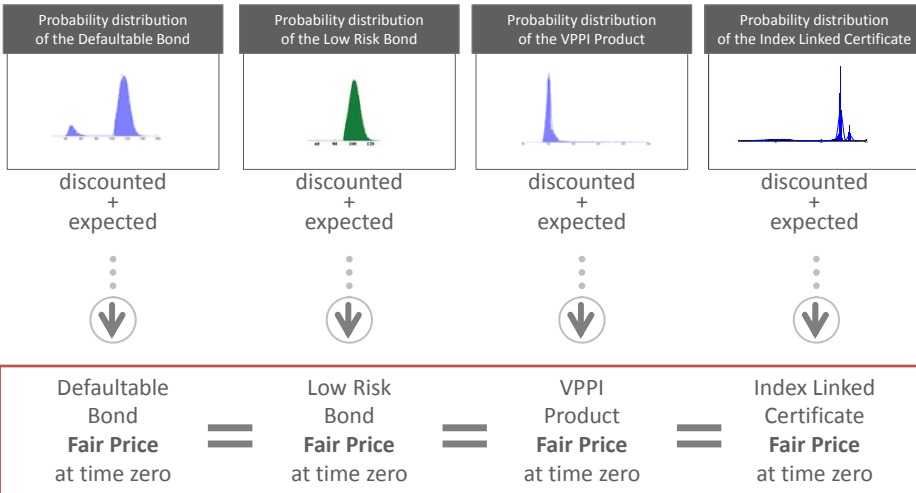
Potential Returns Pricing at maturity

Fair Price at time zero is a Weighted average

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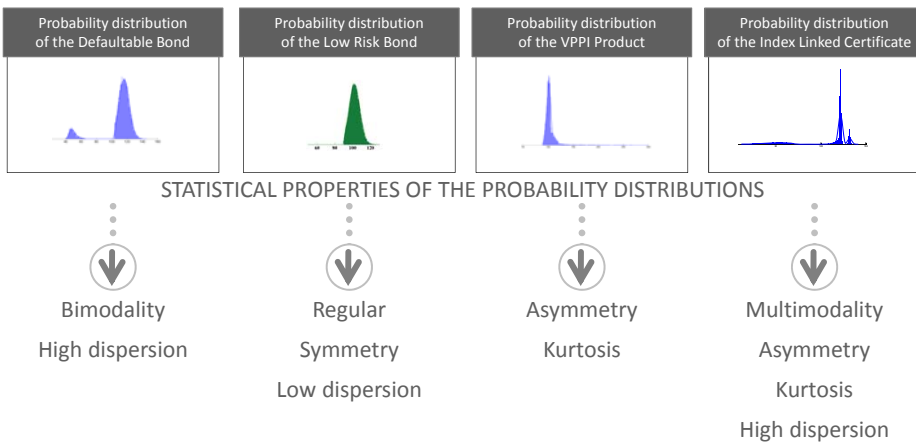
## Unbundling and Probabilistic performance scenarios



**Question:** How much information about the original probability distribution the price will convey in each case analyzed?

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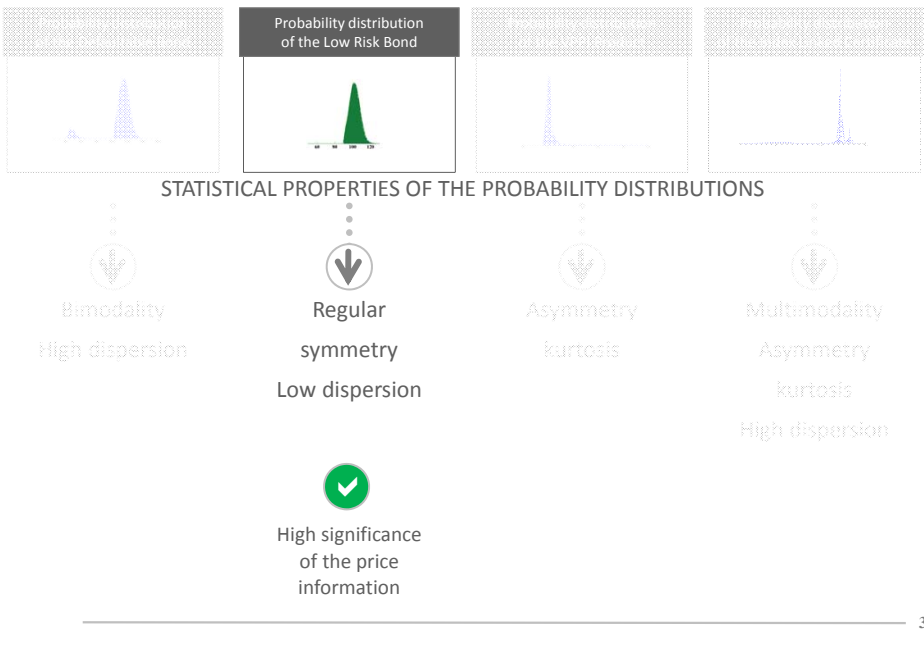
## Unbundling and Probabilistic performance scenarios



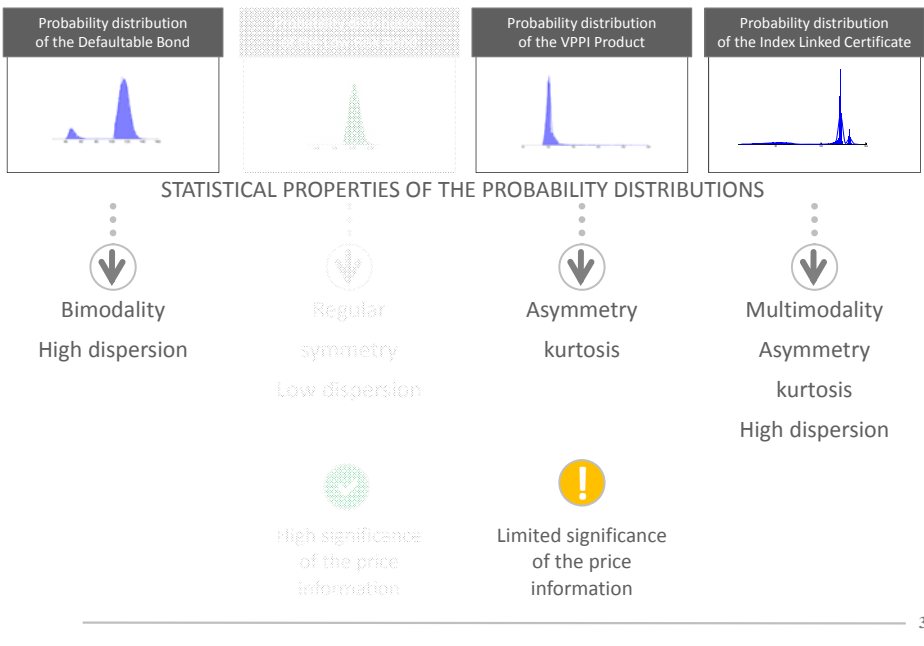
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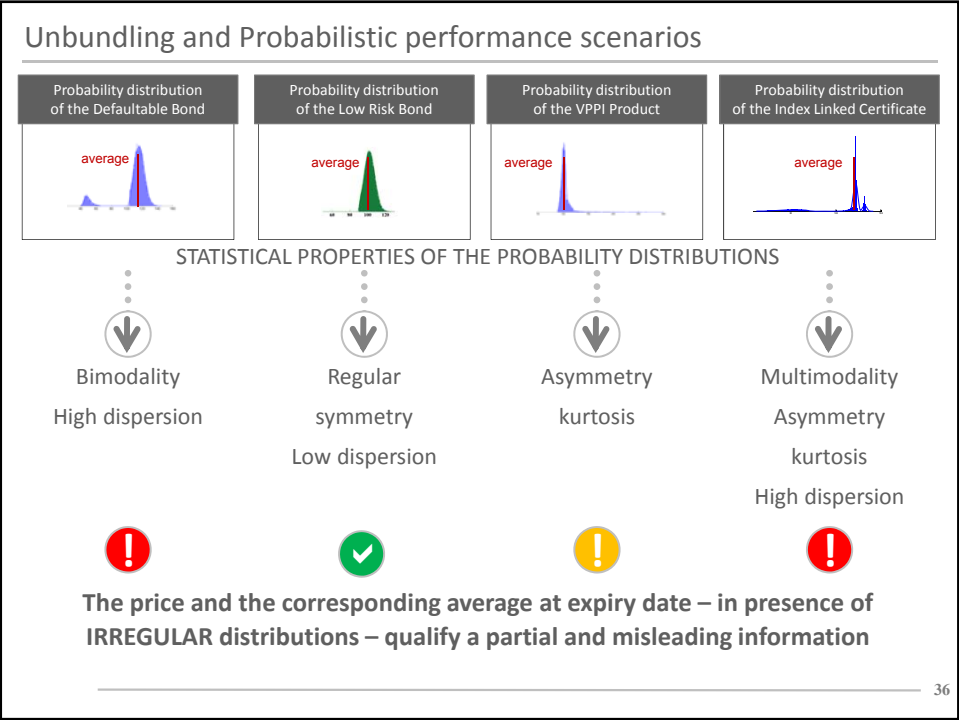
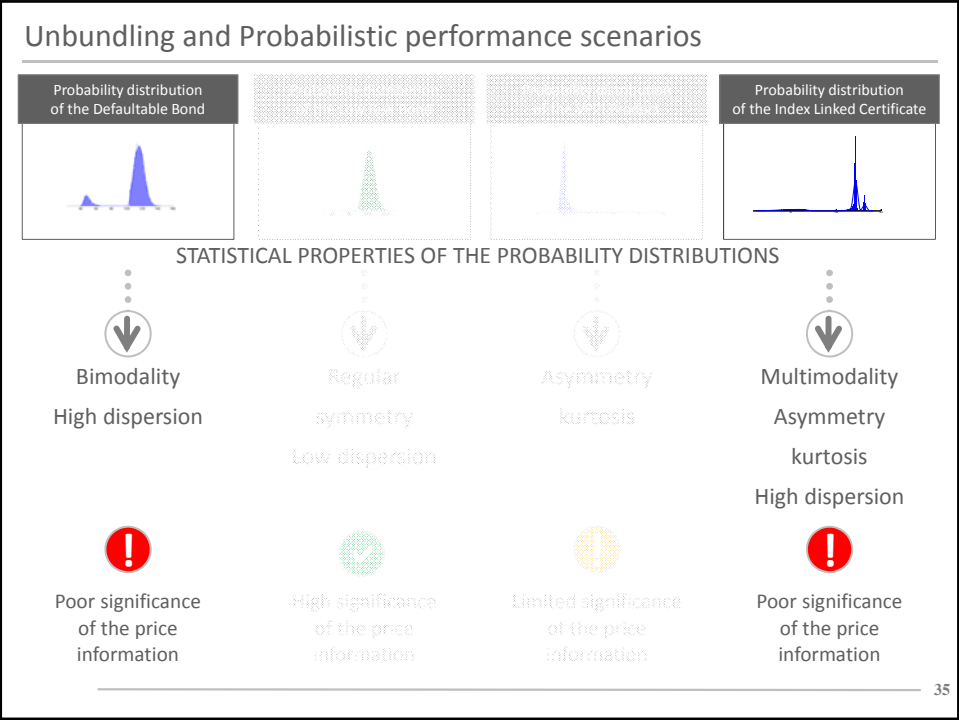


## Unbundling and Probabilistic performance scenarios



## Unbundling and Probabilistic performance scenarios





## Unbundling and Probabilistic performance scenarios



Significance  
of the price  
information



**As a weighted average, the price is strictly connected with the first moment of the probability distribution**

As the literature suggests, in presence of multimodality and irregular shapes for the probability distributions, the number of moments necessary to properly describe the probability distribution increases dramatically.

See:

- (1) Shohat, Tamarkin, 1943 - American Mathematical Survey
- (2) Szego, 1959 - American Mathematical Society
- (3) Totik, 2000 – Journal of Analytical Mathematics
- (4) Gavriiliadis, Athanassoulis, 2009 – Journal of Computational and Applied Mathematics

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## Unbundling and Probabilistic performance scenarios



Significance  
of the price  
information



**Mathematical Basis to test the significance of the price information**

Given a finite number of moments  $2k$ , it's possible to derive the following approximate relationship between the probability function  $f(x)$  and its Christoffel function of degree  $k$ :

$$f(x) \approx f_{AP,k}(x) = \frac{k}{c_0 \pi \sqrt{(x-a)(b-x)}} \lambda_k(x)$$


con  $x \in [a, b]$ .  $c_0$  è un fattore di normalizzazione.



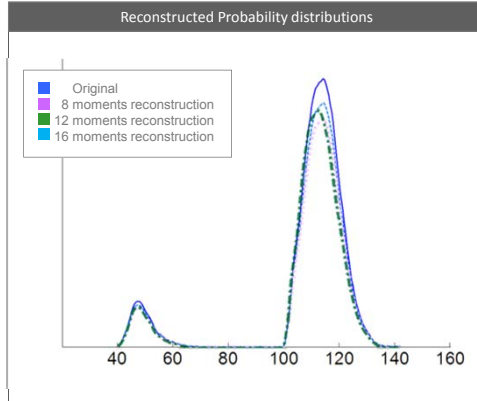
It's then immediate to apply the approximating formula for different values of  $k$  in order to test the accuracy of the approximation for the probability distributions corresponding to our different financial products

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Unbundling and Probabilistic performance scenarios


 Bimodality  
High dispersion

Significance test of the price information

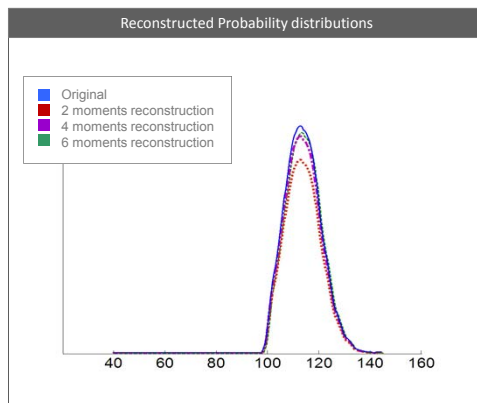


At least 16 moments are needed in order to obtain a satisfactory approximation of the original distribution. The information content of the first moment seems very limited.

Unbundling and Probabilistic performance scenarios

 Regular  
symmetry  
Low dispersion

Significance test of the price information

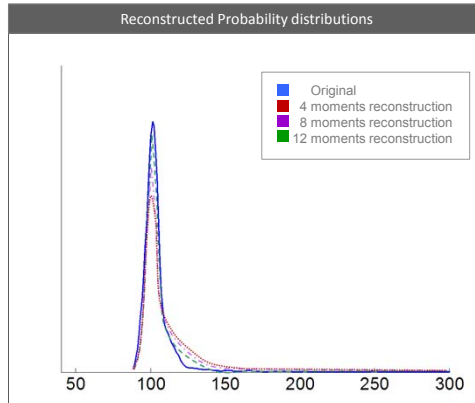


Only 4 moments are sufficient in order to describe properly the original distribution. The information content of the first moment can be considered adequate.

Unbundling and Probabilistic performance scenarios

Asymmetry  
kurtosis

Significance test of the price information

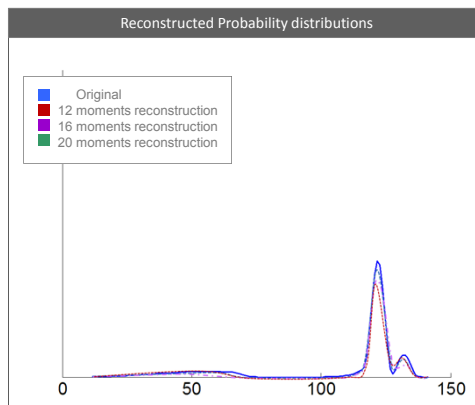


12 moments describe correctly the pattern of the original distribution. The information content of the first moment needs to be integrated.

Unbundling and Probabilistic performance scenarios

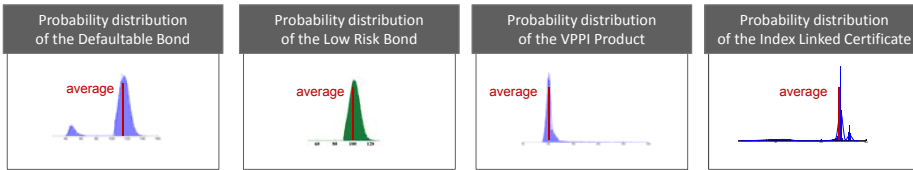
Multimodality  
Asymmetry  
kurtosis  
High dispersion

Significance test of the price information

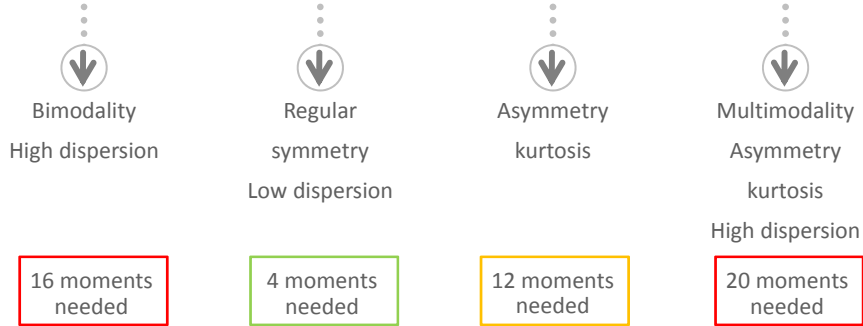


At least 20 moments are needed in order to obtain a satisfactory approximation of the original distribution. The information content of the first moment seems very limited.

## Unbundling and Probabilistic performance scenarios



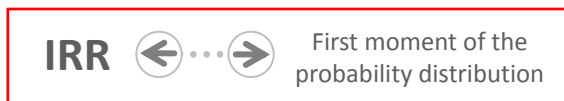
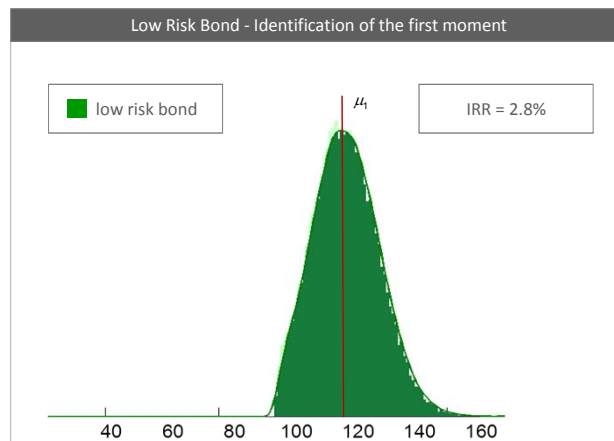
### STATISTICAL PROPERTIES OF THE PROBABILITY DISTRIBUTIONS



From a pure statistical point of view, a proper reconstruction of the original distribution needs at least 4 moments even for the most regular one

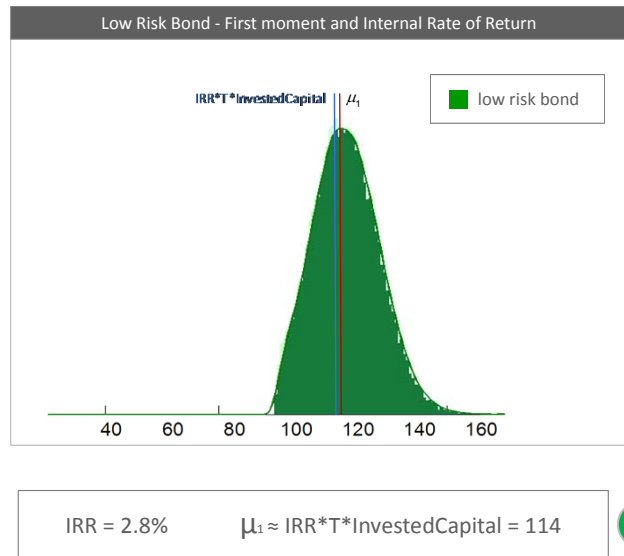
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## Unbundling and Probabilistic performance scenarios



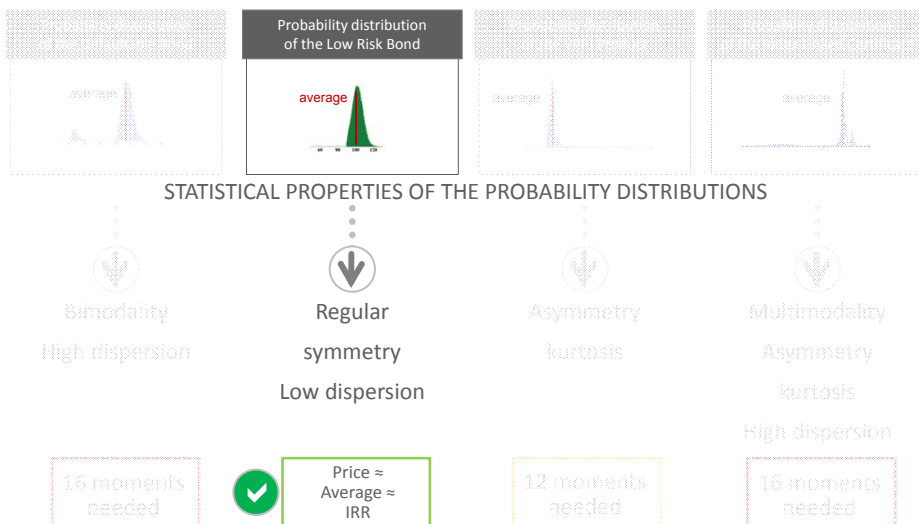
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## Unbundling and Probabilistic performance scenarios



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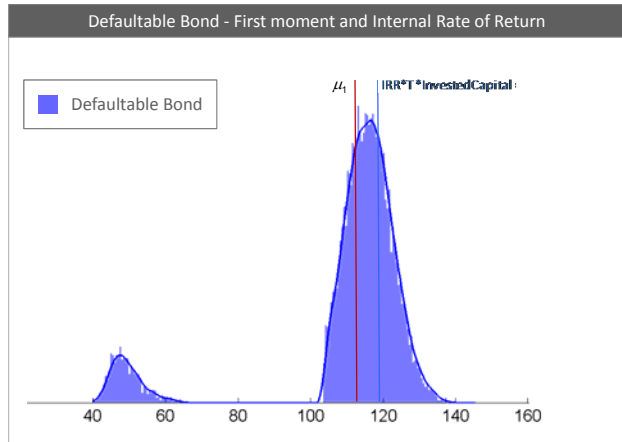
## Unbundling and Probabilistic performance scenarios



Even if 4 moments are needed for a proper reconstruction of the probability distribution, the average and its related measures (IRR and price), convey sufficient information for the investor decision process

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## Unbundling and Probabilistic performance scenarios



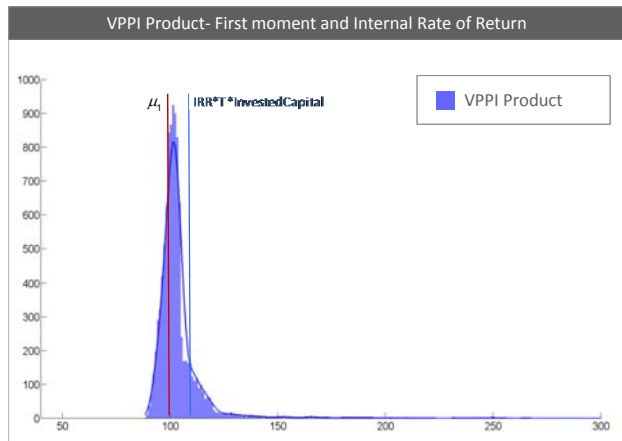
IRR = 3.85%

$\mu_1 \neq \text{IRR} * \text{InvestedCapital} = 119.25$



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## Unbundling and Probabilistic performance scenarios



IRR = 2.53%

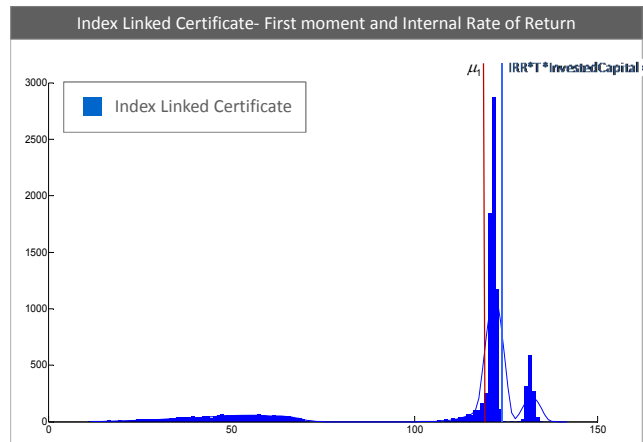
$\mu_1 \neq \text{IRR} * \text{InvestedCapital} = 112.65$



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## Unbundling and Probabilistic performance scenarios



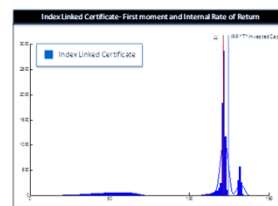
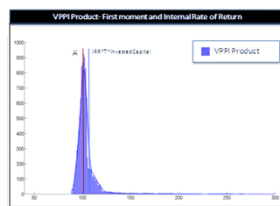
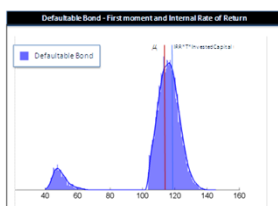
IRR = 5.91%

$\mu_1 \neq \text{IRR} * T * \text{InvestedCapital} = 129.55$



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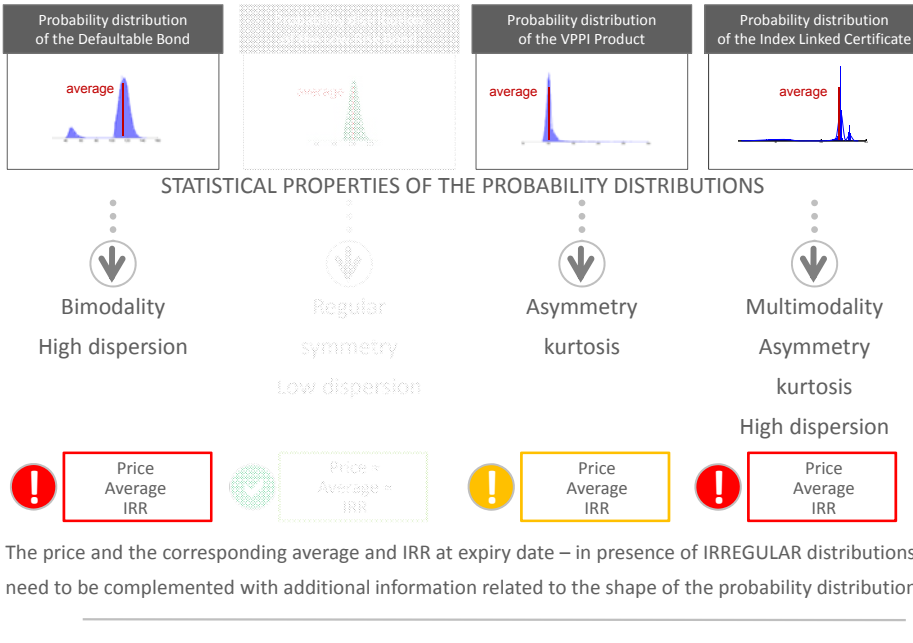
## Unbundling and Probabilistic performance scenarios



For more complex financial structures, the average progressively loses its connection with the internal rate of return of the investment, so reducing its usefulness as an effective tool for the decision process

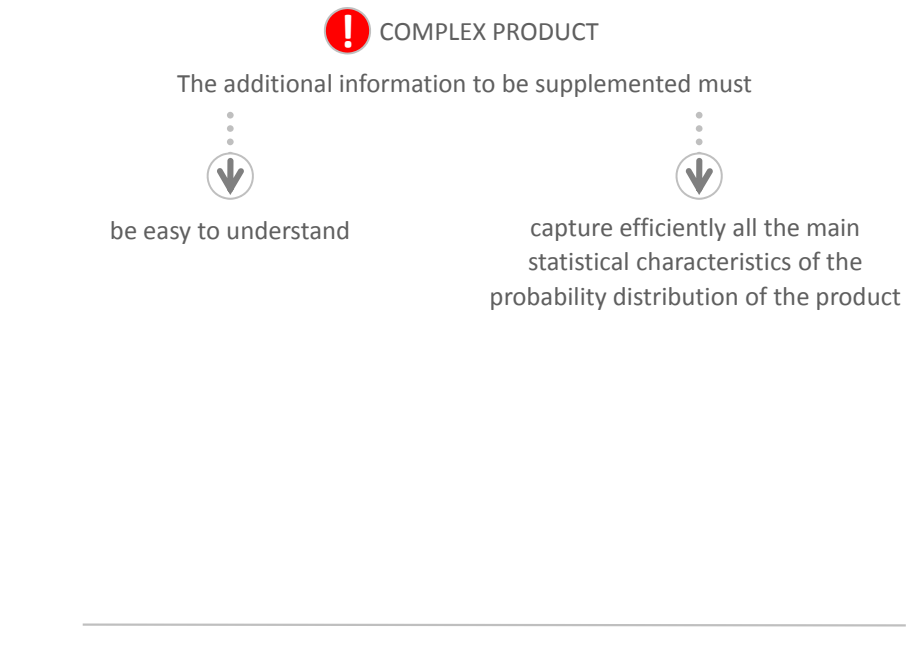
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## Unbundling and Probabilistic performance scenarios



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## Unbundling and Probabilistic performance scenarios



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## Unbundling and Probabilistic performance scenarios

### ! COMPLEX PRODUCT

The additional information to be supplemented must



be easy to understand



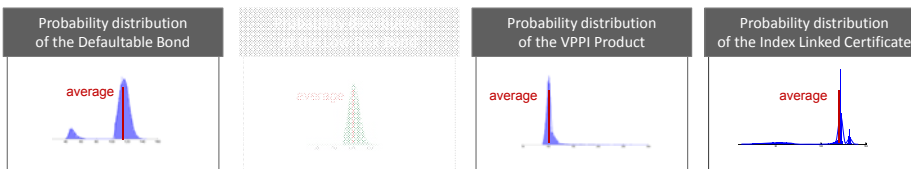
capture efficiently all the main statistical characteristics of the probability distribution of the product



**Proposal 1: Convey the entire probability distribution**

53

## Unbundling and Probabilistic performance scenarios



MODELLING CHOICES FOR THE SELECTED FINANCIAL PRODUCTS



2 Factor Short Interest Rate Hull-White Model

Short Interest Rate Cox Ingersoll Ross Model



Heston Stochastic Volatility Model for the Equity component

Barndorff Nielsen Normal Inverse Gaussian Model for the Equity component



Merton Jump Diffusion Model for the Equity component

Variance Gamma Model for the Equity component

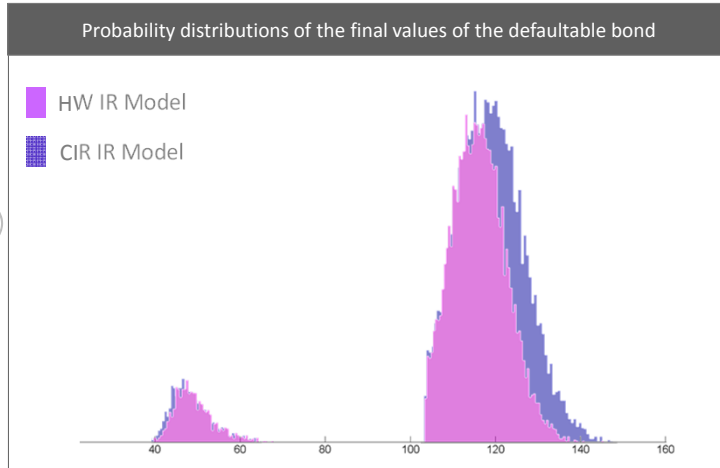
The shape of the probability distribution of the potential returns is obviously dependent on the modelling assumptions.

54

## Unbundling and Probabilistic performance scenarios

DEFAULTABLE  
BOND

MODELLING CHOICES FOR THE SELECTED FINANCIAL PRODUCTS

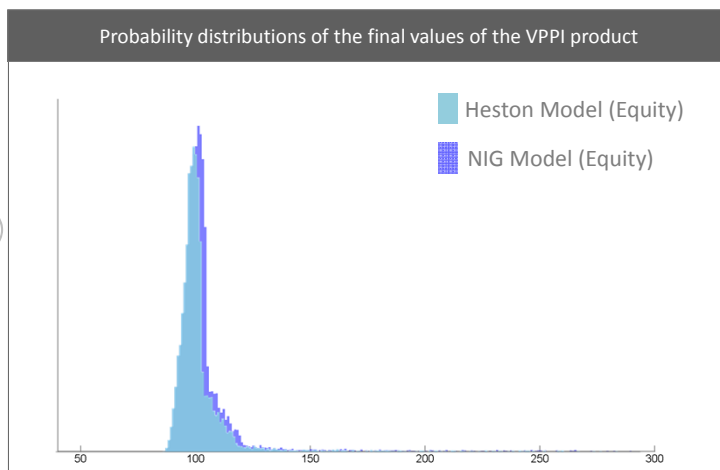


55

## Unbundling and Probabilistic performance scenarios

VPPI PRODUCT

MODELLING CHOICES FOR THE SELECTED FINANCIAL PRODUCTS



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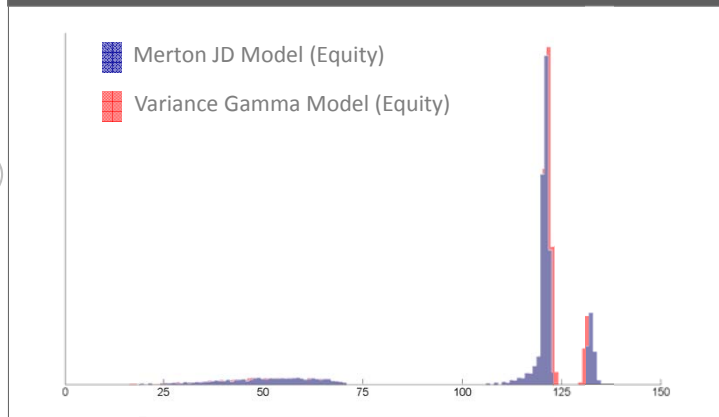
## Unbundling and Probabilistic performance scenarios

INDEX LINKED  
CERTIFICATE

MODELLING CHOICES FOR THE SELECTED FINANCIAL PRODUCTS



Probability distributions of the final values of the Index Linked Certificate



57

## Unbundling and Probabilistic performance scenarios

**!** COMPLEX PRODUCT

The additional information to be supplemented must



be easy to understand



the probability distribution is an abstract object not easy to handle



capture efficiently all the main statistical characteristics of the probability distribution of the product



the shape of the probability distribution is dependent on the modelling assumptions



**Proposal 1:** Convey the entire probability distribution

58

## Unbundling and Probabilistic performance scenarios

### ! COMPLEX PRODUCT

The additional information to be supplemented must



be easy to understand



capture efficiently all the main statistical characteristics of the probability distribution of the product



**Proposal 2: Unbundling the information content of the price**

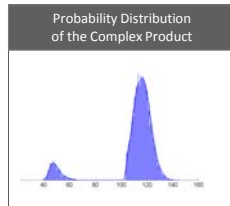
59

## Unbundling and Probabilistic performance scenarios

! COMPLEX PRODUCT



### Unbundling the information content of the price



DISCOUNTED  
EXPECTED  
VALUE



**Fair Value**  
(Complex Product)

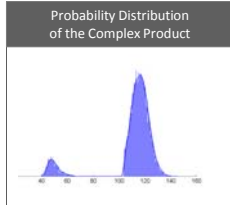
60

## Unbundling and Probabilistic performance scenarios

COMPLEX PRODUCT



### Unbundling the information content of the price

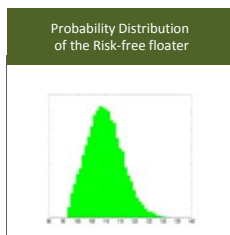


DISCOUNTED EXPECTED VALUE

Fair Value  
(Complex Product)

A risk-free floater with same fair value and coupon payment dates of the complex product is defined

=



DISCOUNTED EXPECTED VALUE

Fair Value  
(Risk-free floater)

61

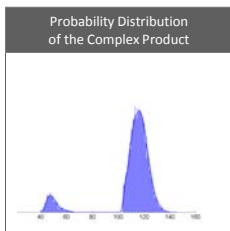
## Unbundling and Probabilistic performance scenarios

COMPLEX PRODUCT



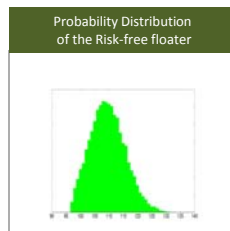
### Unbundling the information content of the price

Any non-elementary return-target product can be replicated by a portfolio composed of the associated risk-free floater and of a zero-value swap which transforms the cash flow structure of the risk-free security into the cash flow structure of the product itself, ie, denoting by  $\{swap_t\}_{t \in [0, T]}$  the value process of the swap



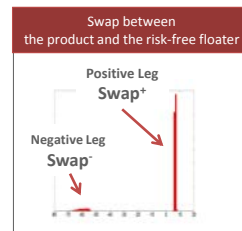
Fair Value  
(Complex Product)

=



Fair Value  
(Risk-free floater)

+



Fair Value  
(Swap = 0)

62

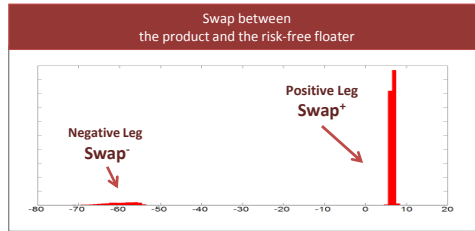
## Unbundling and Probabilistic performance scenarios

COMPLEX PRODUCT



Fair Value  
(Swap = 0)

### Unbundling the information content of the price



$$|FV(\text{Swap}^-)| = |FV(\text{Swap}^+)|$$



Theoretical value of the Risky component

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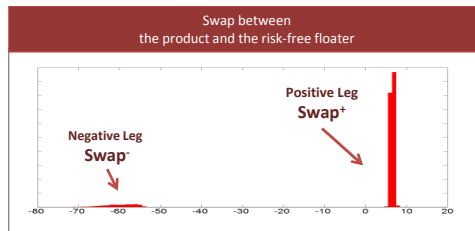
## Unbundling and Probabilistic performance scenarios

COMPLEX PRODUCT



Fair Value  
(Swap = 0)

### Unbundling the information content of the price



$$|FV(\text{Swap}^-)| = |FV(\text{Swap}^+)|$$



Theoretical value of the Risky component



C	Fair value
B	Theoretical value of the Risky component
A=C-B	Theoretical value of the Risk-Free component

64



## Unbundling and Probabilistic performance scenarios

### Financial investment table (Price Unbundling)

#### DEFAULTABLE BOND



A	Theoretical value of the Risk-Free component	91.3
B	Theoretical value of the Risky component	5
C = A + B	<i>Fair value</i>	96.3
D	Costs	3.7
E = C + D	Issue price	100

#### VPPI PRODUCT



A	Theoretical value of the Risk-Free component	90.1
B	Theoretical value of the Risky component	6.4
C = A + B	<i>Fair value</i>	96.5
D	Costs	3.5
E = C + D	Issue price	100

#### INDEX LINKED CERTIFICATE



A	Theoretical value of the Risk-Free component	86.2
B	Theoretical value of the Risky component	9.9
C = A + B	<i>Fair value</i>	96.1
D	Costs	3.9
E = C + D	Issue price	100

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## Unbundling and Probabilistic performance scenarios

### ! COMPLEX PRODUCT

The additional information to be supplemented must



✓ be easy to understand



capture efficiently ~~all~~ the main statistical characteristics of the probability distribution of the product

the unbundling represented by using a table is first level tool useful to appreciate the impact of the costs and the riskiness of the product

The unbundling exploits only the information contained in the first order moment of the probability distribution



**Proposal 2:** Unbundling the information content of the price

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## Unbundling and Probabilistic performance scenarios

### ! COMPLEX PRODUCT

The additional information to be supplemented must



be easy to understand



capture efficiently all the main statistical characteristics of the probability distribution of the product



**Proposal 3:** Perform a reduction in granularity by implementing a partition of the probability distribution

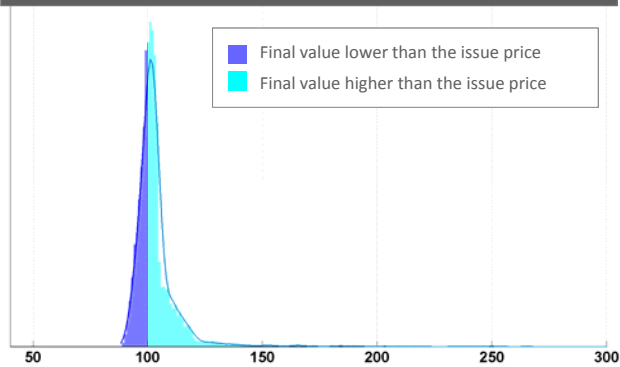
67

## Unbundling and Probabilistic performance scenarios

### ! COMPLEX PRODUCT



Partition of the Probability distribution of the Complex Product with respect to the point of zero return



The assessment of the probability of recovering at least the amount paid for the product is of great significance for the investor.

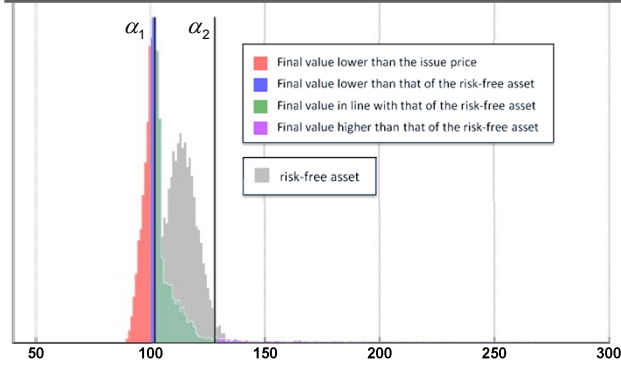
68

## Unbundling and Probabilistic performance scenarios

COMPLEX PRODUCT



Partition of the risk-neutral density of the Complex Product with respect to the point of zero return and to the two fixed positive thresholds  $\alpha_1$  and  $\alpha_2$



It is appropriate to explore further partitions of the macro-event “the final value of the investment is higher than the issue price” by performing a direct comparison with the final values of the risk-free asset.

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## Unbundling and Probabilistic performance scenarios

COMPLEX PRODUCT



Partition of the risk-neutral density of the Complex Product with respect to the point of zero return and to the two fixed positive thresholds  $\alpha_1$  and  $\alpha_2$

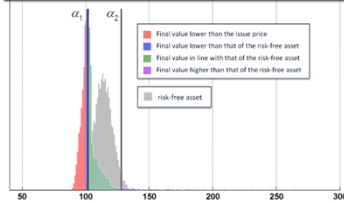


Table of the probabilistic performance scenarios

SCENARIOS	PROBABILITY	MEAN VALUES
The performance is <u>negative</u>	...	...
The performance is <u>positive but lower than the risk-free asset</u>	...	...
The performance is <u>positive and in line with the risk-free asset</u>	...	...
The performance is <u>positive and higher than the risk-free asset</u>	...	...

MEAN VALUES



$$E^P(S_T | S_T < 100) = \frac{1}{P(S_T < 100)} \int_{-\infty}^{100} x f_{S_T}(x) dx$$

$$E^P(S_T | 100 \leq S_T < \alpha_1) = \frac{1}{P(100 \leq S_T < \alpha_1)} \int_{100}^{\alpha_1} x f_{S_T}(x) dx$$

$$E^P(S_T | \alpha_1 \leq S_T < \alpha_2) = \frac{1}{P(\alpha_1 \leq S_T < \alpha_2)} \int_{\alpha_1}^{\alpha_2} x f_{S_T}(x) dx$$

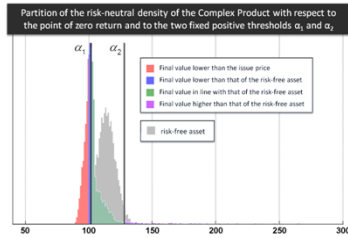


$$E^P(S_T | S_T \geq \alpha_2) = \frac{1}{P(S_T \geq \alpha_2)} \int_{\alpha_2}^{+\infty} x f_{S_T}(x) dx$$

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## Unbundling and Probabilistic performance scenarios

### COMPLEX PRODUCT



Benefits of this solution:

1. The reduction in granularity of the events determined by the partition involves only a very limited loss of information and the table, built by coupling for each scenario its risk-neutral probability and the associated mean value, is very easy to read;

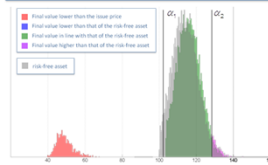
71

## Unbundling and Probabilistic performance scenarios

### DEFAULTABLE BOND



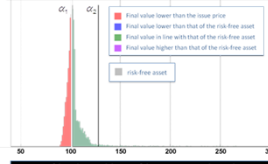
Partition of the risk-neutral density of the non-equity product with respect to the point of zero return and to the two fixed positive thresholds  $\alpha_1$  and  $\alpha_2$



### VPPI PRODUCT



Partition of the risk-neutral density of the Complex Product with respect to the point of zero return and to the two fixed positive thresholds  $\alpha_1$  and  $\alpha_2$



### INDEX LINKED CERTIFICATE



Partition of the risk-neutral density of the non-equity product with respect to the point of zero return and to the two fixed positive thresholds  $\alpha_1$  and  $\alpha_2$



SCENARIOS	PROBABILITY	MEAN VALUES
The performance is <u>negative</u>	9.5%	49.3
The performance is <u>positive but lower</u> than the risk-free asset	0.0%	-
The performance is <u>positive and in line</u> with the risk-free asset	87.4%	115.6
The performance is <u>positive and higher</u> than the risk-free asset	3.1%	131.1

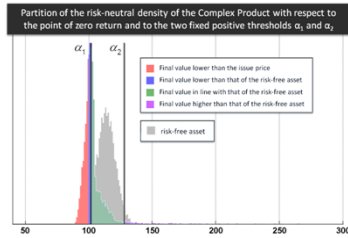
SCENARIOS	PROBABILITY	MEAN VALUES
The performance is <u>negative</u>	36.9%	96.9
The performance is <u>positive but lower</u> than the risk-free asset	18.5%	101
The performance is <u>positive and in line</u> with the risk-free asset	39.9%	107.1
The performance is <u>positive and higher</u> than the risk-free asset	4.7%	195.5

SCENARIOS	PROBABILITY	MEAN VALUES
The performance is <u>negative</u>	18.9%	49.1
The performance is <u>positive but lower</u> than the risk-free asset	0.0%	-
The performance is <u>positive and in line</u> with the risk-free asset	68.9%	120.9
The performance is <u>positive and higher</u> than the risk-free asset	12.2%	131.6

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## Unbundling and Probabilistic performance scenarios

### COMPLEX PRODUCT



Benefits of this solution:

1. The reduction in granularity of the events determined by the partition involves only a very limited loss of information; The table, built by coupling for each scenario its risk-neutral probability and the associated mean value, is very easy to read;
2. The model risk arising from the different proprietary models of the issuers has a limited impact.

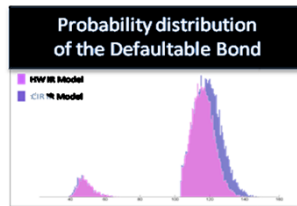
73

## Unbundling and Probabilistic performance scenarios

### DEFAULTABLE BOND



### MODELLING CHOICES FOR THE SELECTED FINANCIAL PRODUCTS



Difference less than 2%

#### HW IR MODEL

SCENARIOS	PROBABILITY	MEAN VALUES
The performance is <u>negative</u>	9.5%	49.3
The performance is <u>positive but lower</u> than the risk-free asset	0.0%	-
The performance is <u>positive and in line</u> with the risk-free asset	87.4%	115.6
The performance is <u>positive and higher</u> than the risk-free asset	3.1%	131.1

#### CIR IR MODEL

SCENARIOS	PROBABILITY	MEAN VALUES
The performance is <u>negative</u>	8.3%	49.9
The performance is <u>positive but lower</u> than the risk-free asset	0.0%	-
The performance is <u>positive and in line</u> with the risk-free asset	86.8%	117.9
The performance is <u>positive and higher</u> than the risk-free asset	4.9%	135.4

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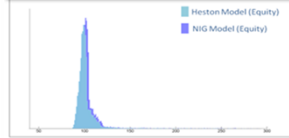
## Unbundling and Probabilistic performance scenarios

VPPI PRODUCT

MODELLING CHOICES FOR THE SELECTED FINANCIAL PRODUCTS



### Probability distribution of the VPPI Product



Difference less than 2%

#### HESTON MODEL

SCENARIOS	PROBABILITY	MEAN VALUES
The performance is <u>negative</u>	38.9%	95.5
The performance is <u>positive but lower</u> than the risk-free asset	18.9%	100.2
The performance is <u>positive and in line</u> with the risk-free asset	38.4%	106.3
The performance is <u>positive and higher</u> than the risk-free asset	3.8%	182.5

#### NIG MODEL

SCENARIOS	PROBABILITY	MEAN VALUES
The performance is <u>negative</u>	36.9%	96.9
The performance is <u>positive but lower</u> than the risk-free asset	18.5%	101
The performance is <u>positive and in line</u> with the risk-free asset	39.9%	107.1
The performance is <u>positive and higher</u> than the risk-free asset	4.7%	195.5

75

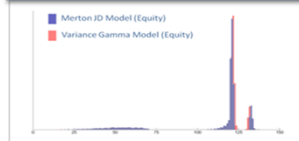
## Unbundling and Probabilistic performance scenarios

INDEX LINKED CERTIFICATE

MODELLING CHOICES FOR THE SELECTED FINANCIAL PRODUCTS



### Probability distribution of the Index Linked Certificate



Difference less than 4%

#### MERTON JD MODEL

SCENARIOS	PROBABILITY	MEAN VALUES
The performance is <u>negative</u>	18.9%	48.2
The performance is <u>positive but lower</u> than the risk-free asset	0.0%	-
The performance is <u>positive and in line</u> with the risk-free asset	65.8%	117.6
The performance is <u>positive and higher</u> than the risk-free asset	15.3%	132.7

#### VARIANCE GAMMA MODEL

SCENARIOS	PROBABILITY	MEAN VALUES
The performance is <u>negative</u>	18.9%	49.1
The performance is <u>positive but lower</u> than the risk-free asset	0.0%	-
The performance is <u>positive and in line</u> with the risk-free asset	68.9%	120.9
The performance is <u>positive and higher</u> than the risk-free asset	12.2%	131.6

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## Unbundling and Probabilistic performance scenarios

### ! COMPLEX PRODUCT

The additional information to be supplemented must

✓ be easy to understand

✓ capture efficiently all the main statistical characteristics of the probability distribution of the product

the partition should be done by choosing events that have a strong financial meaning

the reduction in granularity mitigates in a significant way the model risk

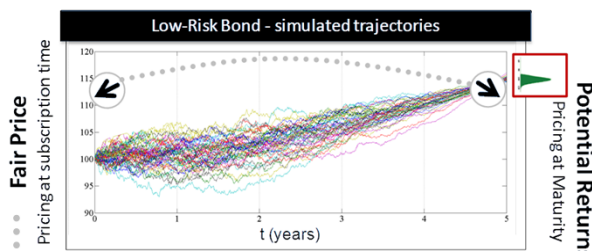
**Proposal 3:** Perform a reduction in granularity by implementing a partition of the probability distribution

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## Unbundling and Probabilistic performance scenarios

Since there's a close one-to-one relationship between the two tables, the two sets of information can be easily coupled in an easy-to-read sheet

COMPLEX PRODUCT



Financial investment table  
(Price Unbundling)

A	Theoretical value of the Risk-Free component	
B	Theoretical value of the Risky component	
C = A + B	Fair value	
D	Costs	
E = C + D	Issue price	

Table of the probabilistic performance scenarios

SCENARIOS	PROBABILITY	MEAN VALUES
The performance is <u>negative</u>	...	...
The performance is <u>positive but lower</u> than the risk-free asset	...	...
The performance is <u>positive and in line</u> with the risk-free asset	...	...
The performance is <u>positive and higher</u> than the risk-free asset	...	...

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## Syllabus

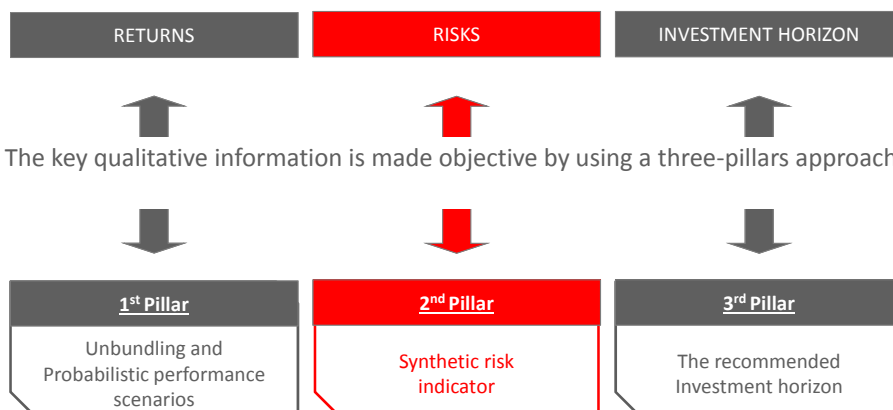
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- Preliminaries: the three pillars
- Unbundling and Probabilistic performance scenarios
- Synthetic risk indicator
- The optimal time horizon
- An Application of the methodology

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## Synthetic risk indicator

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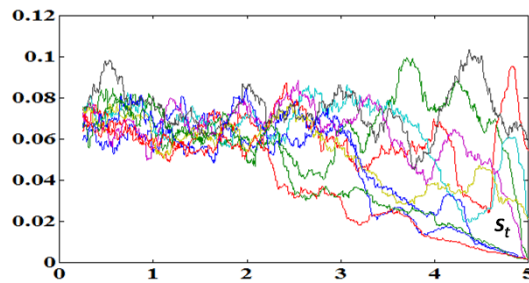


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## Synthetic risk indicator

Volatility of the product's potential returns

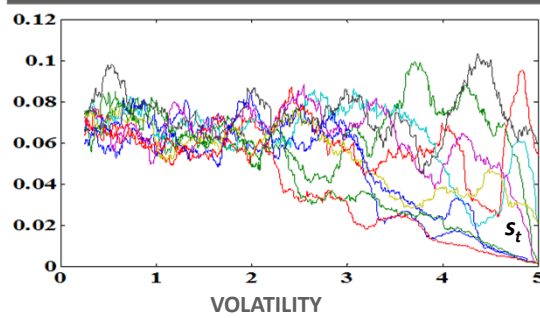


Volatility is the most immediate risk measure and it has a one-to-one relationship with whatever loss measure (VaR, ES, etc.)

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## Synthetic risk indicator

Volatility of the product's potential returns



e.g. : geometric brownian motion

$$dS_t = rS_t dt + \sigma S_t dW_t$$

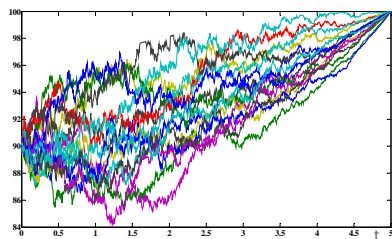
$$VaR_{\alpha, 1year} = e^{\sigma \Phi^{-1}(\alpha) + \left(r - \frac{\sigma^2}{2}\right)} - 1$$

$$ES_{\alpha, 1year} = \frac{1}{\alpha} e^{\left(\Phi^{-1}(\alpha) - \sigma\right)}$$

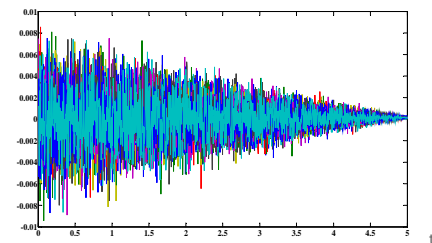
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# EXAMPLES

Simulation of the trajectories (Price)

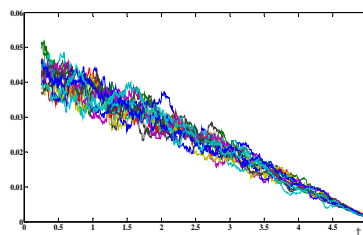


Simulation of the trajectories (Returns)



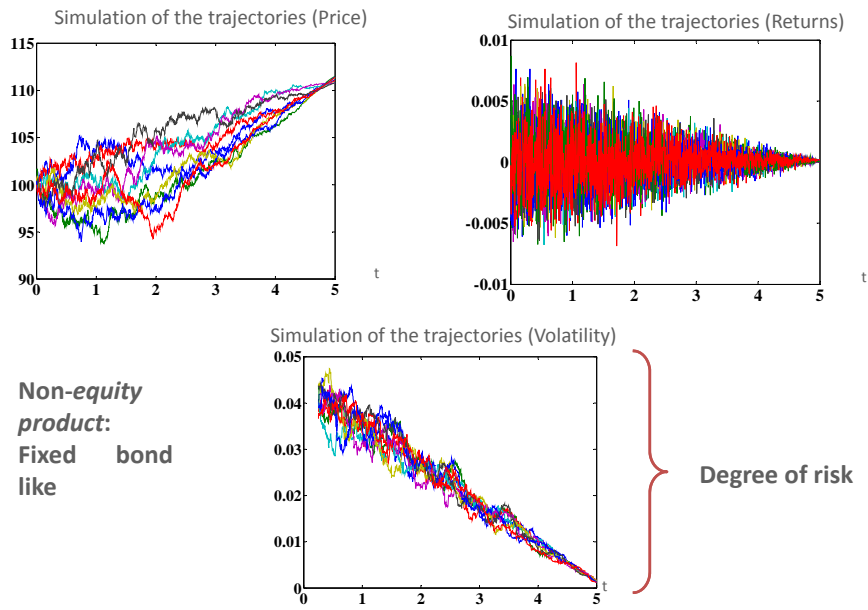
Simulation of the trajectories (Volatility)

**Non-equity  
product: Zero  
Coupon Bond**

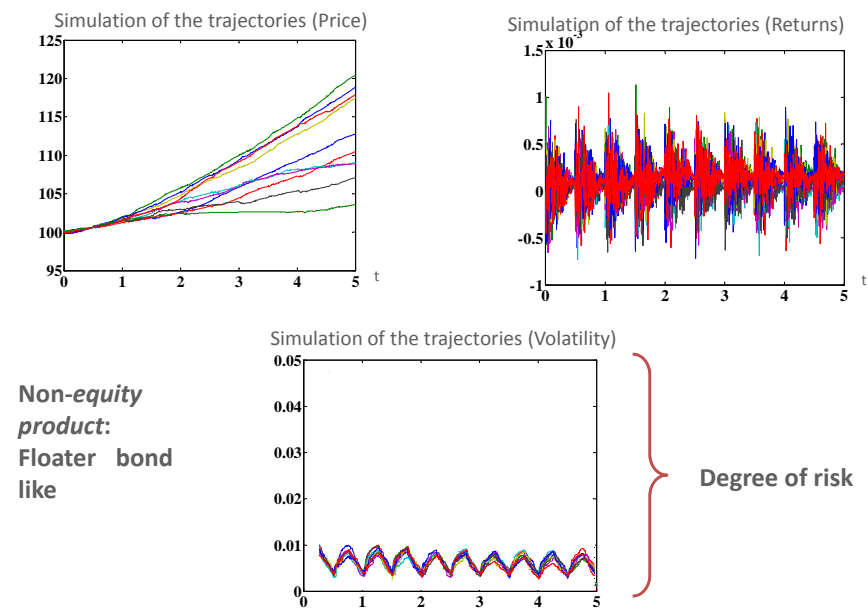


**Degree of risk**

### Synthetic risk indicator

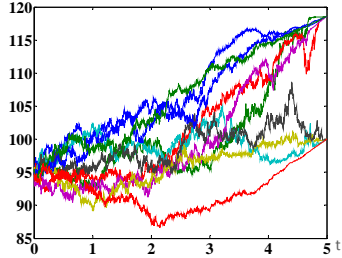


### Synthetic risk indicator

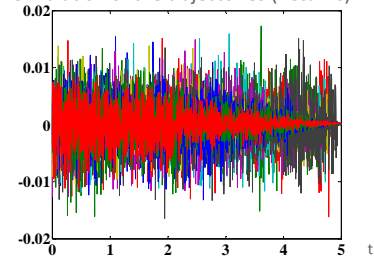


## Synthetic risk indicator

Simulation of the trajectories (Price)

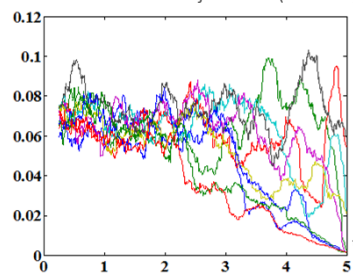


Simulation of the trajectories (Returns)



**Non-equity  
product:  
Structured  
Bond**

Simulation of the trajectories (Volatility)



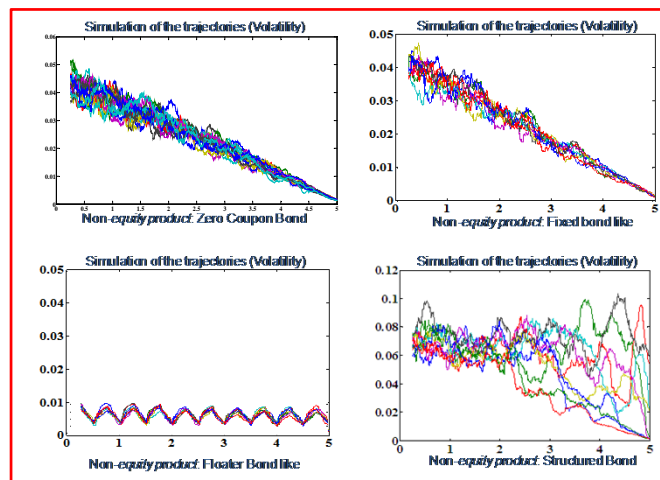
**Degree of risk**

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## Synthetic risk indicator

### COMPLEXITY FOR RETAIL INVESTORS

The volatility patterns are abstract objects that an average investor cannot handle.



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## Synthetic risk indicator

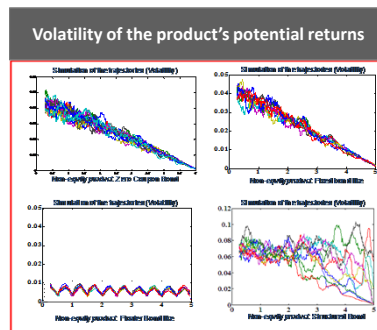
Conversely, a table with qualitative labels that characterizes the risk classes is very easy to understand

Risk Classes
Very Low
Low
Medium-Low
Medium
Medium-High
High
Very High

The assignment of the degree of risk is made according to a quantitative criterion that maps coherently any volatility interval into a corresponding qualitative risk class

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## Synthetic risk indicator



**MEASUREMENT:**  
product's positioning inside  
a grid of  $n$  volatility intervals

**REPRESENTATION:**  
mapping of any volatility interval into a  
corresponding qualitative risk class

**DEGREE OF RISK**

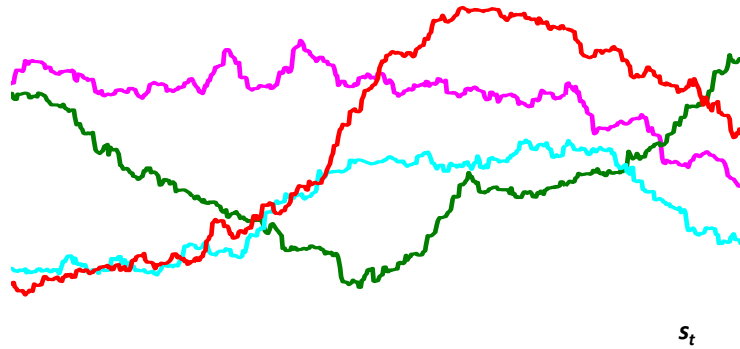


Risk Classes	Volatility Intervals	
Very Low	$\sigma_{1,min}$	$\sigma_{1,max}$
Low	$\sigma_{2,min}$	$\sigma_{2,max}$
Medium-Low	$\sigma_{3,min}$	$\sigma_{3,max}$
Medium	$\sigma_{4,min}$	$\sigma_{4,max}$
Medium-High	$\sigma_{5,min}$	$\sigma_{5,max}$
High	$\sigma_{6,min}$	$\sigma_{6,max}$
Very High	$\sigma_{7,min}$	$\sigma_{7,max}$

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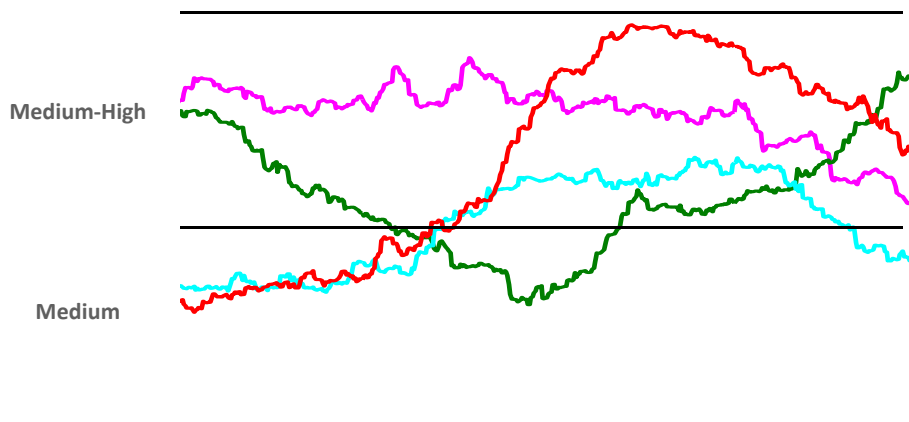
Synthetic risk indicator

Products with the same risk budget  
must have the same degree of risk



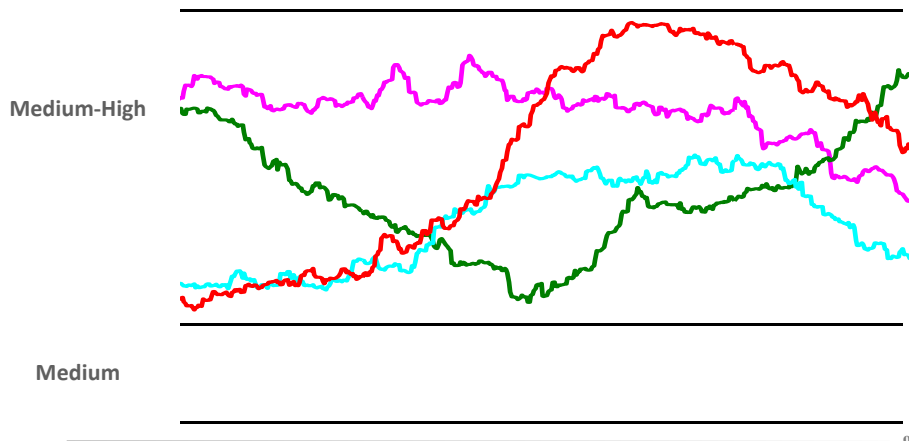
Synthetic risk indicator

Volatility intervals have to be suitably calibrated  
in order to avoid wrong risk representations



Synthetic risk indicator

Volatility intervals have to be suitably calibrated  
in order to avoid wrong risk representations



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Synthetic risk indicator

Volatility intervals have to be suitably calibrated  
in order to avoid wrong risk representations

**THE ISSUE**

Defining suitable requirements to partition the volatility space  $[0, +\infty)$  into an optimal number  $n^*$  of subsequent intervals with optima extrema



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Synthetic risk indicator

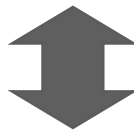
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Volatility intervals have to be suitably calibrated  
in order to avoid wrong risk representations

*Requirement n.1*

the **optimal grid** of volatility intervals  
has to be **consistent** with the **principle**:

**+ RISK + LOSSES**



**VOLATILITY INTERVALS MUST HAVE  
AN INCREASING WIDTH IN ABSOLUTE TERMS**

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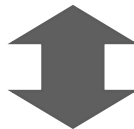
Synthetic risk indicator

---

Volatility intervals have to be suitably calibrated  
in order to avoid wrong risk representations

*Requirement n.2*

the optimal grid of volatility intervals must be  
**market feasible**



**REALIZED VOLATILITY CONSISTENT WITH MARKET  
EXPECTATIONS OF FUTURE VOLATILITY**  
(UNLESS FOR SIGNIFICANT SUDDEN SHOCKS)

---

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## Synthetic risk indicator

### Realized volatility

Any product on the markets reflects specific/different asset management policies

Historical data can be "dirty"



#### 1<sup>st</sup> INTUITION

It has to be studied a theoretical product managed by an automatic asset manager who has a specific risk budget, identified by a given volatility interval

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## Synthetic risk indicator

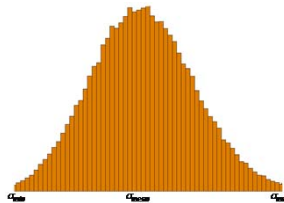
#### 1<sup>st</sup> INTUITION



#### **AUTOMATIC ASSET MANAGER:**

described by a stochastic volatility model with:

- mean reversion
- symmetry w.r.t. to a given risk budget
- ex ante minimization of the migration risk



$$dS_t = rS_t dt + \sigma_t S_t dW_t^{(1)}$$

$$d\sigma_t^2 = \kappa(\vartheta - \sigma_t^2) dt + v_t \sigma_t dW_t^{(2)}$$

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### Market expectations of future volatility

future volatility is predicted by exploiting information embedded in recently observed data



### 2<sup>nd</sup> INTUITION

Market expectation is given by volatility prediction intervals based on proper diffusive models

### 2<sup>nd</sup> INTUITION

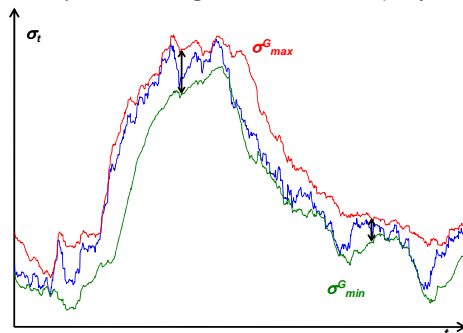


#### VOLATILITY PREDICTION INTERVALS:

obtained by the diffusion limit of a multiplicative GARCH model

$$d \ln \sigma_t^2 = (\beta_0 + 2\beta_1 E(\ln|Z_t|) + (\beta_1 - 1) \ln \sigma_t^2) dt + 2|\beta_1| \sqrt{\text{Var}(\ln|Z_t|)} dW_t^*$$

- well-known distributional properties
- immediate update according to new information (**adaptivity**)



### Assessing *market feasibility*

putting together the two ingredients



### 3<sup>rd</sup> INTUITION

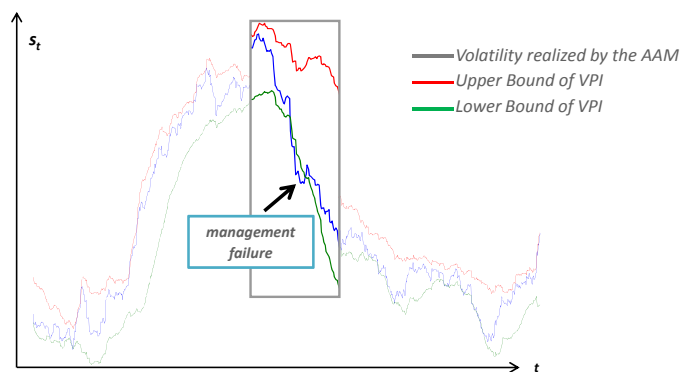
It requires to study when the realized volatility of the automatic asset manager is outside the Garch-based volatility prediction interval:

**MANAGEMENT FAILURES**

### 3<sup>rd</sup> INTUITION



**MANAGEMENT FAILURES:**



Synthetic risk indicator

3<sup>rd</sup> INTUITION

MANAGEMENT FAILURES:

Market feasibility regards all  $n$  intervals of the optimal grid at the same time:

NOT ABNORMALITY

HOMOGENEITY

low number of failures

equal number of failures

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Synthetic risk indicator

Solving for the optimal grid



in a reduced volatility space  $[\sigma_0, \sigma_n]$

NOT ABNORMALITY

same relative width:

&



$$\lambda = \left( \frac{\sigma_n}{\sigma_0} \right)^{\frac{1}{n}}$$

HOMOGENEITY

for all intervals

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## Synthetic risk indicator

### The optimal grid

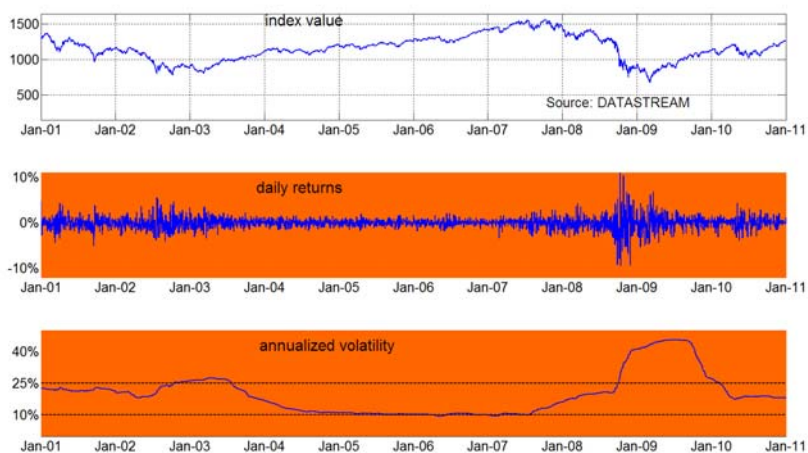
Risk Classes	Volatility Intervals	
	$\sigma_{min}$	$\sigma_{max}$
Very Low	0.01%	0.24%
Low	0.25%	0.63%
Medium-Low	0.64%	1.59%
Medium	1.60%	3.99%
Medium-High	4.00%	9.99%
High	10.00%	24.99%
Very High	25.00%	>25.00%

The optimal grid is consistent with the 1<sup>st</sup> requirement: + RISK + LOSSES

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## Synthetic risk indicator

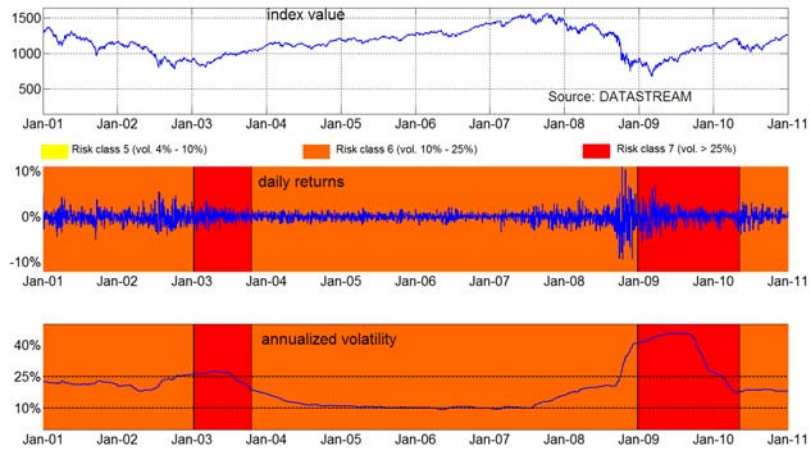
### Application to a market index: the optimal grid



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## Synthetic risk indicator

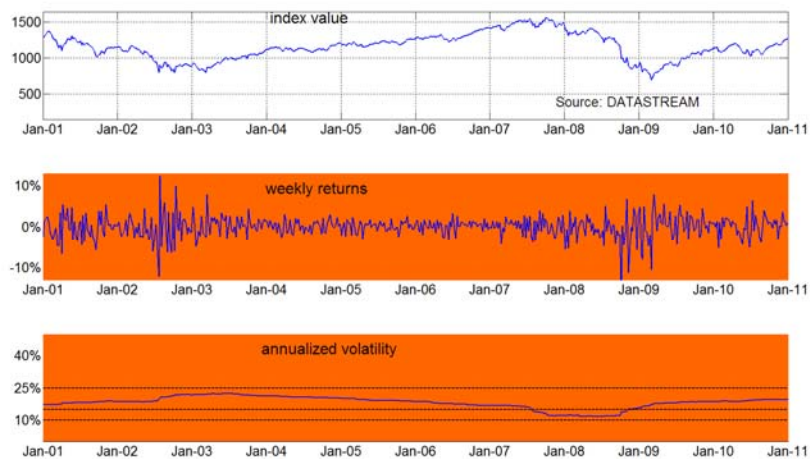
### Application to a market index: the optimal grid



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## Synthetic risk indicator

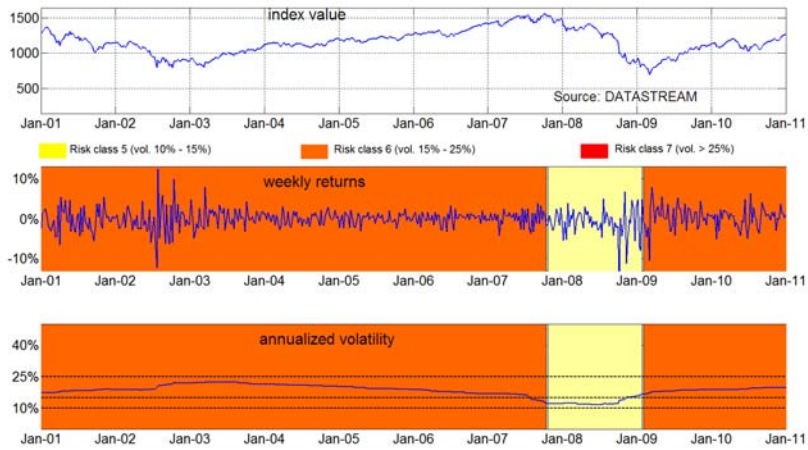
### Application to a market index: a non-optimal grid



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## Synthetic risk indicator

### Application to a market index: a non-optimal grid



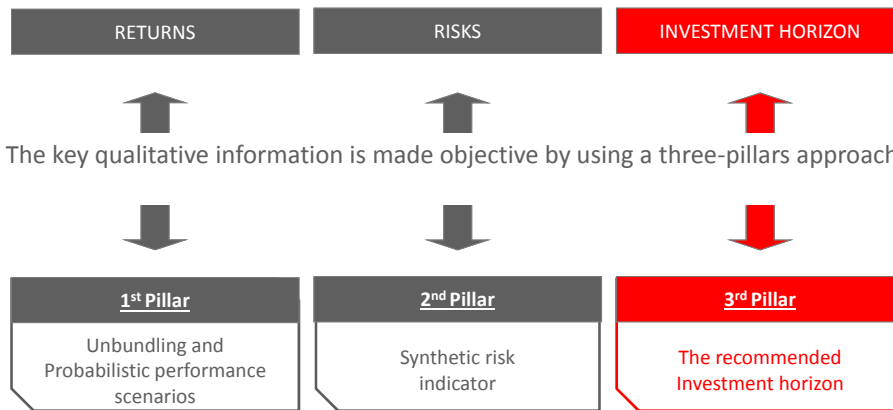
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## Syllabus

- Preliminaries: the three pillars
- Unbundling and Probabilistic performance scenarios
- Synthetic risk indicator
- The optimal time horizon
- An Application of the methodology

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## The recommended Investment horizon



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## The recommended Investment horizon



### The recommended investment time horizon

The event to study from a probabilistic point of view is related to possible exit strategies after having recovered all the costs of the product :

The investment recovers the initial costs and off-sets the running costs at least once

that can be calculated through the concept of

First Passage Time

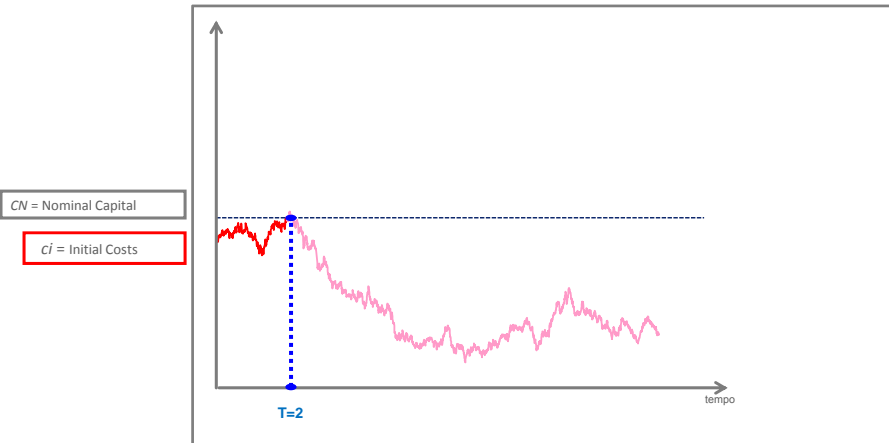
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## The recommended Investment horizon

### First Passage Time:

First time (expressed in years) such that the value of the Invested Capital (CI) recovers the initial costs and off-sets the running costs.

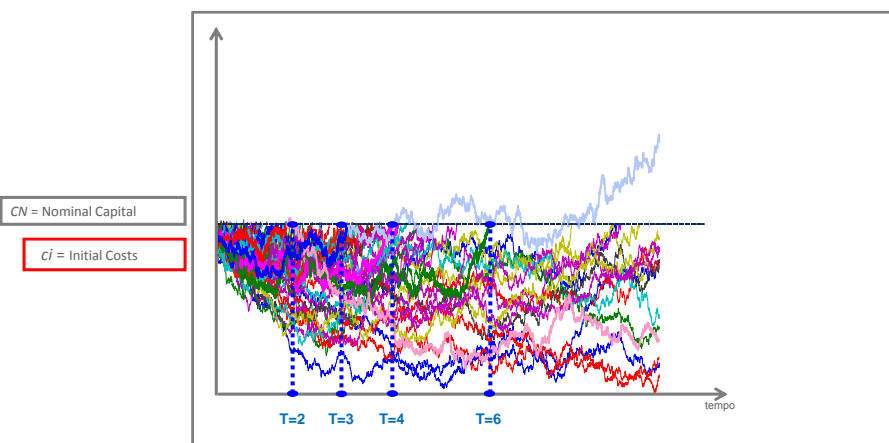


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## The recommended Investment horizon

### First Passage Time:

First time (expressed in years) such that the value of the Invested Capital (CI) recovers the initial costs and off-sets the running costs.



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## The recommended Investment horizon

The probability of the event:

The investment recovers the initial costs and off-sets the running costs at least once

given a confidence level  $\alpha$ , uniquely identifies a time  $T^*$  on the cumulative distribution function of the first passage times, i.e.:

$$T^* = \left\{ T \in \mathcal{R}^+ : \mathbb{P}[t^* \leq T] = \alpha \right\}$$

where

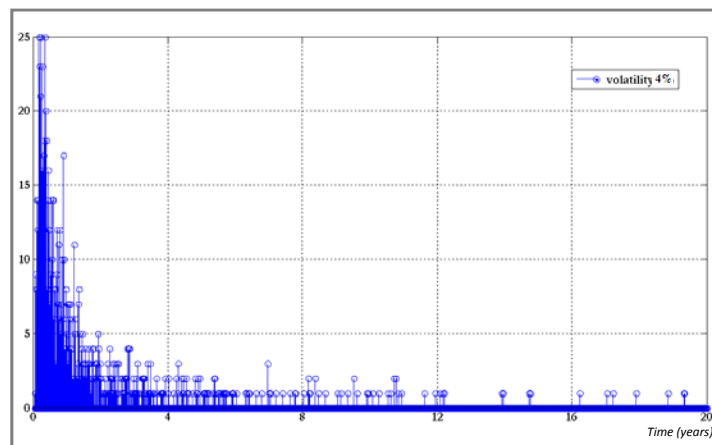
$$t^* = \inf \left[ t \in \mathcal{R}^+ : CI_t > CN \right]$$

is the first passage time

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## The recommended Investment horizon

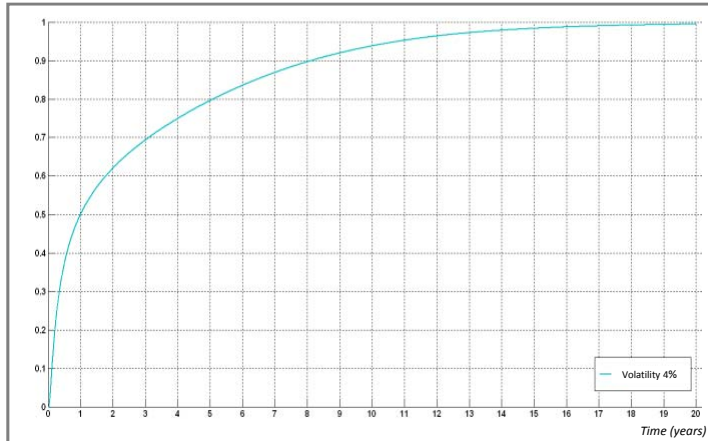
### 1. Calculation of the probability distribution of the first passage times:



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## The recommended Investment horizon

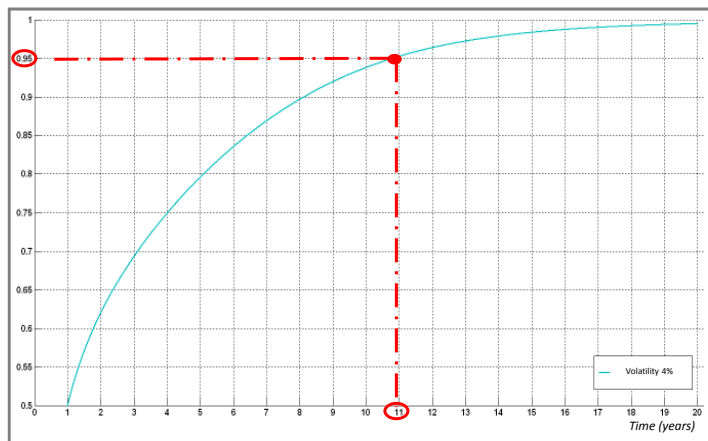
2. Derivation of the cumulative distribution function of the first passage times:



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## The recommended Investment horizon

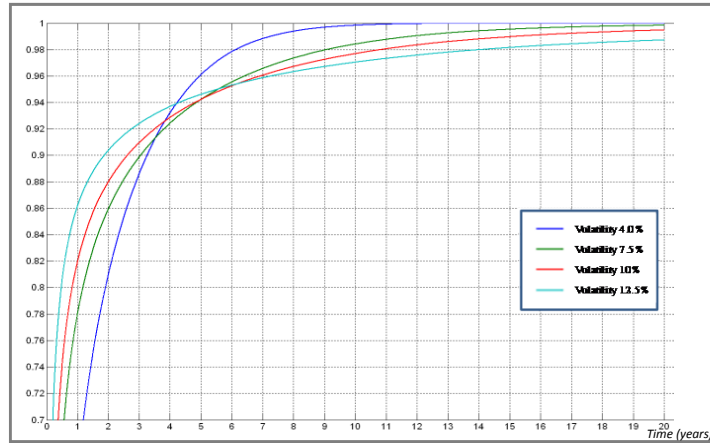
3. The confidence level  $\alpha$  uniquely identifies  $T^*$  on the cumulative distribution function of the first passage times:



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## The recommended Investment horizon

When many probability distribution functions are considered, letting varying volatilities and costs, the problem of correctly identifying a set of minimum thresholds arises:



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## The recommended Investment horizon

Anyway, the recommended minimum investment time horizon...

$$T^* = \left\{ T \in \mathcal{R}^+ : \mathbb{P}[t^* \leq T] = \alpha \right\}$$

.... Must be coherent with the principle

+ VOLATILITY + TIME HORIZON

The correct way to solve the problem is to set up an operative procedure to select properly each threshold according to the above principle

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## The recommended Investment horizon

Connection between probability, volatility and costs

First passage times for the break-even barrier are monitored at infinitesimal time intervals:



$dt \rightarrow 0$

$$T^* = \left\{ T \in \mathfrak{R}^+ : \mathbb{P}[t^* \leq T] = \alpha \right\}$$

$$\mathbb{P}[t^* \leq T] = N\left(d_2\left(\frac{CI_0}{CN}\right)\right) + \left(\frac{CN}{CI_0}\right)^{\frac{2(\bar{r}-cr)}{\sigma^2}-1} \cdot N\left(-d_2\left(\frac{CN}{CI_0}\right)\right)$$

$$d_2(x) = \frac{\log x + \left(\bar{r} - cr - \frac{1}{2}\sigma^2\right)T}{\sigma\sqrt{T}}$$

$$N(x) = \int_{-\infty}^x \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}z^2} dz$$

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## The recommended Investment horizon

Connection between probability, volatility and costs

Asymptotic properties:  $T \rightarrow \infty$

$cr$ : recurrent costs  
as a fixed %

$$\lim_{T \rightarrow \infty} \mathbb{P}[t^* \leq T] = \begin{cases} 1 & \text{if } (\bar{r} - cr) \geq \frac{1}{2}\sigma^2 \\ \left(\frac{CN}{CI_0}\right)^{\frac{2(\bar{r}-cr)}{\sigma^2}-1} & \text{if } (\bar{r} - cr) < \frac{1}{2}\sigma^2 \end{cases}$$

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## The recommended Investment horizon

Connection between probability, volatility and costs

Under our assumptions:

$$\lim_{T \rightarrow \infty} P[t^* \leq T] = \begin{cases} 1 & \text{if } (\bar{r} - cr) \geq \frac{1}{2} \sigma^2 \\ \left( \frac{CN}{CI_0} \right)^{\frac{2(\bar{r} - cr)}{\sigma^2} - 1} & \text{if } (\bar{r} - cr) < \frac{1}{2} \sigma^2 \end{cases}$$

For a given level of costs, it is possible to analytically derive the connection between volatility and time horizon

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## The recommended Investment horizon

Connection between probability, volatility and costs

$T \rightarrow \infty, dt \rightarrow 0$

FIRST ORDER  
SENSITIVITY  
ANALYSIS

$$\frac{dP}{d\sigma} = \left( -4 \frac{(\bar{r} - cr)}{\sigma^3} \ln \left( \frac{CN}{CI_0} \right) \left( \frac{CN}{CI_0} \right)^{\frac{2(\bar{r} - cr)}{\sigma^2} - 1} \right)$$

FIRST ORDER  
ASYMPTOTIC CONDITION

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## The recommended Investment horizon

Connection between probability, volatility and costs

$$T \rightarrow \infty, dt \rightarrow 0$$

$$\frac{dP}{d\sigma} = \left( -4 \frac{(\bar{r} - cr)}{\sigma^3} \ln \left( \frac{CN}{CI_0} \right) \left( \frac{CN}{CI_0} \right)^{\frac{2(\bar{r} - cr)}{\sigma^2} - 1} \right)$$

1.  $(\bar{r} - cr) > 0 \Leftrightarrow \frac{dP}{d\sigma} < 0$
2.  $(\bar{r} - cr) \leq 0 \Leftrightarrow \frac{dP}{d\sigma} \geq 0$

The existence of two alternative states of nature requires to verify whether both of them make sense in financial terms under the risk-neutral measure.

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## The recommended Investment horizon

Connection between probability, volatility and costs

$$T \rightarrow \infty, dt \rightarrow 0$$

$$\frac{dP}{d\sigma} = \left( -4 \frac{\bar{r}}{\sigma^3} \ln \left( \frac{CN}{CI_0} \right) \left( \frac{CN}{CI_0} \right)^{\frac{2\bar{r}}{\sigma^2} - 1} \right)$$

1.  $\bar{r} > 0 \Leftrightarrow \frac{dP}{d\sigma} < 0$
2.  $\bar{r} \leq 0 \Leftrightarrow \frac{dP}{d\sigma} \geq 0$

$cr = 0$

Being running costs a specific feature of any financial product they would interfere with the task of identifying which of the two conditions has a sound financial meaning. Therefore, they will be temporarily neglected.

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## The recommended Investment horizon

Connection between probability, volatility and costs

$$T \rightarrow \infty, dt \rightarrow 0$$

$$\frac{dP}{d\sigma} = \left( -4 \frac{\bar{r}}{\sigma^3} \ln \left( \frac{CN}{CI_0} \right) \left( \frac{CN}{CI_0} \right)^{\frac{2\bar{r}}{\sigma^2} - 1} \right)$$

1.  $\bar{r} > 0 \Leftrightarrow \frac{dP}{d\sigma} < 0$

2.  ~~$\bar{r} \leq 0 \Leftrightarrow \frac{dP}{d\sigma} \geq 0$~~

$cr = 0$

Since it is safe to assume a positive interest rate  $r$  in financial markets, only condition 1. correctly captures the connection between volatility and time horizon.

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## The recommended Investment horizon

Connection between probability, volatility and costs

$$T \rightarrow \infty, dt \rightarrow 0$$

$$\frac{dP}{d\sigma} = \left( -4 \frac{\bar{r}}{\sigma^3} \ln \left( \frac{CN}{CI_0} \right) \left( \frac{CN}{CI_0} \right)^{\frac{2\bar{r}}{\sigma^2} - 1} \right)$$

1.  $\bar{r} > 0 \Leftrightarrow \frac{dP}{d\sigma} < 0$

2.  ~~$\bar{r} \leq 0 \Leftrightarrow \frac{dP}{d\sigma} \geq 0$~~

$cr = 0$

As  $T \rightarrow \infty$  condition 1. implies that the cumulative distribution function  $P$  is a strictly decreasing function of the volatility, i.e.:

$$\forall \sigma_i, \sigma_j \in \mathfrak{R}^+, \sigma_j > \sigma_i \Rightarrow P(\sigma_j) < P(\sigma_i)$$

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## The recommended Investment horizon

Connection between probability, volatility and costs

$$T \rightarrow \infty, dt \rightarrow 0$$

$$\frac{dP}{d\sigma} = \left( -4 \frac{\bar{r}}{\sigma^3} \ln\left(\frac{CN}{CI_0}\right) \left(\frac{CN}{CI_0}\right)^{\frac{2\bar{r}}{\sigma^2}-1} \right)$$

1.  $\bar{r} > 0 \Leftrightarrow \frac{dP}{d\sigma} < 0$

2.  $\bar{r} \leq 0 \Leftrightarrow \frac{dP}{d\sigma} \geq 0$

$cr = 0$

In other words, for a given a confidence level, as the volatility grows, the recommended investment time horizon increases as well:

+VOLATILITY + RECOMMENDED INVESTMENT TIME HORIZON

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## The recommended Investment horizon

Connection between probability, volatility and costs

$$T \rightarrow \infty, dt \rightarrow 0$$

$$\frac{d^2P}{d\sigma^2} = \frac{4}{\sigma^4} (\bar{r} - cr) \ln\left(\frac{CN}{CI_0}\right) \left(\frac{CN}{CI_0}\right)^{\frac{2(\bar{r}-cr)}{\sigma^2}-1} \cdot \left[ 1 + \frac{4(\bar{r}-cr)}{\sigma^2} \ln\left(\frac{CN}{CI_0}\right) \right]$$

$(\bar{r} - cr) > 0 \Rightarrow \frac{d^2P}{d\sigma^2} > 0$

➔

SECOND ORDER ASYMPTOTIC  
CONDITION

Second Order  
Sensitivity  
Analysis

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## The recommended Investment horizon

Connection between probability, volatility and costs

$$T \rightarrow \infty, dt \rightarrow 0$$

$$1. \begin{cases} (\bar{r} - cr) > 0 \Leftrightarrow \frac{dP}{d\sigma} < 0 \\ (\bar{r} - cr) > 0 \Rightarrow \frac{d^2P}{d\sigma^2} > 0 \end{cases}$$

$$\exists T^* \in [0, \infty[ : \frac{dP}{d\sigma} = 0$$

~~$$2. (\bar{r} - cr) \leq 0 \Leftrightarrow \frac{dP}{d\sigma} \geq 0$$~~

Summarizing the results of the asymptotic analysis in continuous time:

- As  $T \rightarrow \infty$ , for given a confidence level, more volatility implies a larger recommended investment time horizon
- It is always possible to find a minimum and finite time  $T^*$ , beyond which the strong condition  
+VOLATILITY + RECOMMENDED INVESTMENT TIME HORIZON

holds

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## The recommended Investment horizon

### DETERMINATION OF THE INVESTMENT TIME HORIZON

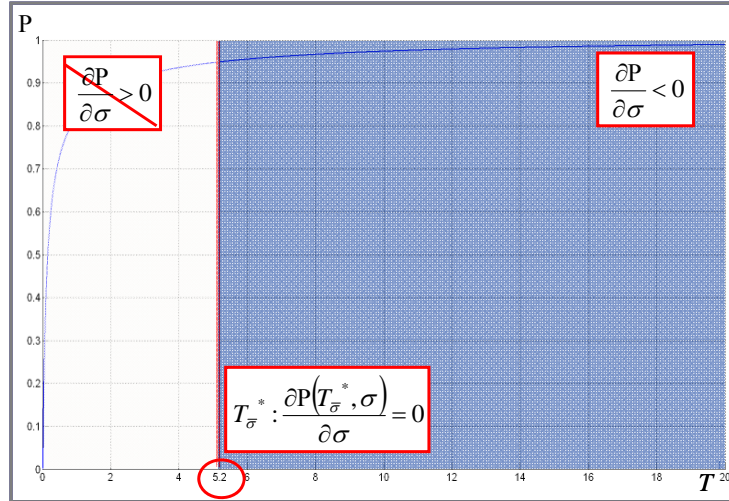
$\left\{ \begin{array}{l} T \rightarrow \infty \\ dt \rightarrow 0 \\ P(\infty, \sigma) \\ 1. \begin{cases} (\bar{r} - cr) > 0 \Leftrightarrow \frac{dP}{d\sigma} < 0 \\ (\bar{r} - cr) > 0 \Rightarrow \frac{d^2P}{d\sigma^2} > 0 \end{cases} \end{array} \right.$	$\Rightarrow$	<p>General Framework:</p> $\left\{ \begin{array}{l} T \text{ finite} \\ dt \rightarrow 0 \\ P(T, \sigma) \\ (\bar{r} - cr) > 0 \Leftrightarrow \lim_{T \rightarrow \infty} \frac{\partial P(T, \sigma)}{\partial \sigma} < 0 \\ (\bar{r} - cr) > 0 \Rightarrow \lim_{T \rightarrow \infty} \frac{\partial^2 P(T, \sigma)}{\partial \sigma^2} > 0 \end{array} \right.$
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Everything shown above also holds with  $T$  finite!

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## The recommended Investment horizon

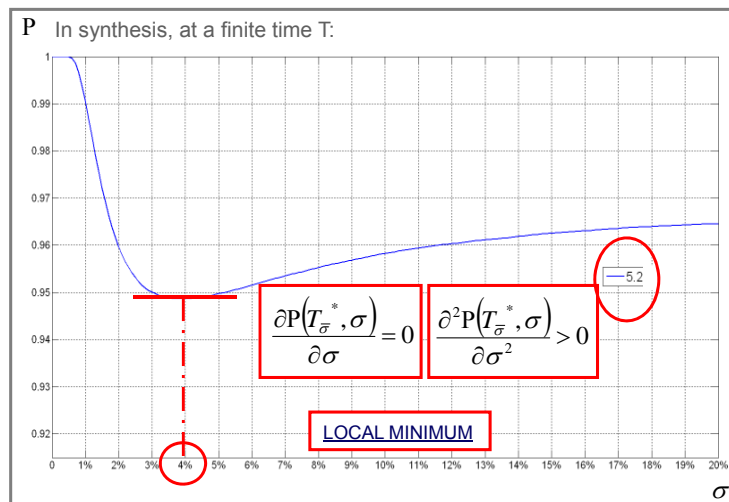
### DETERMINATION OF THE INVESTMENT TIME HORIZON



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## The recommended Investment horizon

### DETERMINATION OF THE INVESTMENT TIME HORIZON



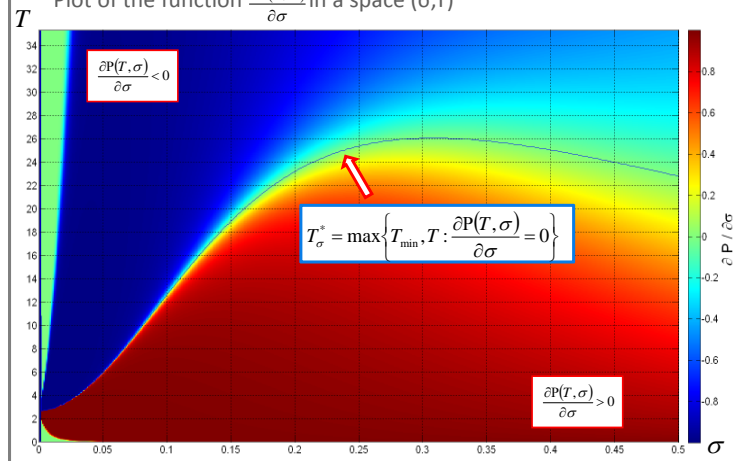
134

## The recommended Investment horizon

### DETERMINATION OF THE INVESTMENT TIME HORIZON

#### FIRST ORDER SENSITIVITY ANALYSIS

Plot of the function  $\frac{\partial P(T, \sigma)}{\partial \sigma}$  in a space  $(\sigma, T)$



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## The recommended Investment horizon

### DETERMINATION OF THE INVESTMENT TIME HORIZON

#### SECOND ORDER SENSITIVITY ANALYSIS

$$\frac{\partial^2 P(T, \sigma)}{\partial \sigma^2}$$

Given the monotonicity condition of the probability distribution with respect to volatility, i.e.:

$$\forall \sigma_i, \sigma_j \in \mathfrak{R}^+, \sigma_j > \sigma_i \Rightarrow P(\infty, \sigma_j) < P(\infty, \sigma_i)$$

In order to fulfill this condition, it's necessary to restrict the analysis in the region where the probability function is strictly increasing, i.e.:

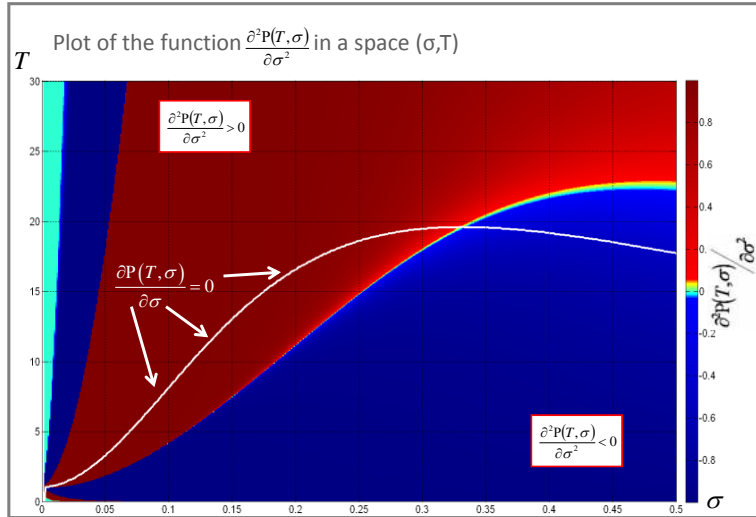
$$\left. \frac{\partial^2 P(T, \sigma)}{\partial \sigma^2} \right|_{T=T_{\sigma}^*} > 0 \Rightarrow T_{\sigma}^* \text{ increasing}$$

$$\left. \frac{\partial^2 P(T, \sigma)}{\partial \sigma^2} \right|_{T=T_{\sigma}^*} < 0 \Rightarrow T_{\sigma}^* \text{ decreasing}$$

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## The recommended Investment horizon

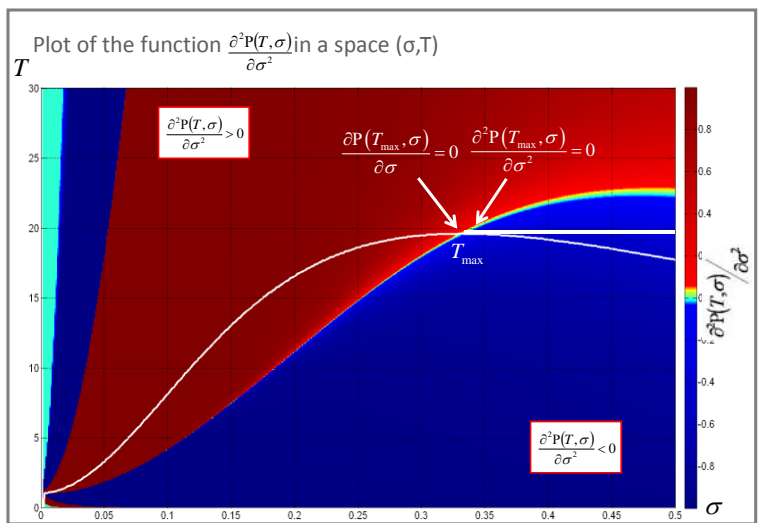
### DETERMINATION OF THE INVESTMENT TIME HORIZON



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## The recommended Investment horizon

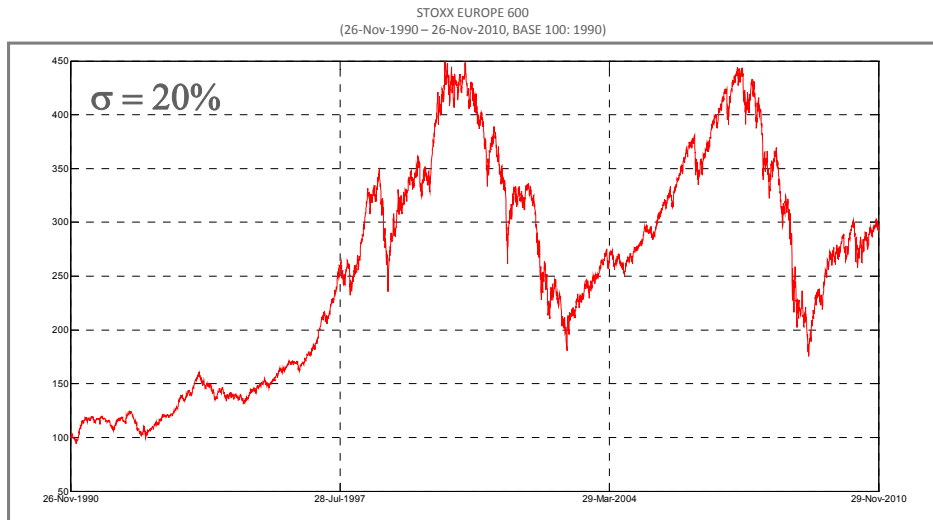
### DETERMINATION OF THE INVESTMENT TIME HORIZON



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## The recommended Investment horizon

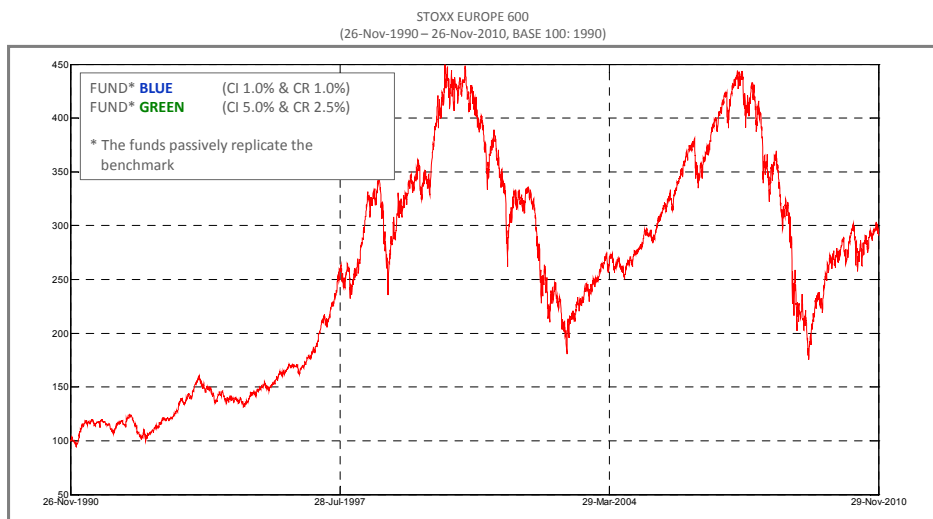
### DETERMINATION OF THE INVESTMENT TIME HORIZON



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## The recommended Investment horizon

### DETERMINATION OF THE INVESTMENT TIME HORIZON

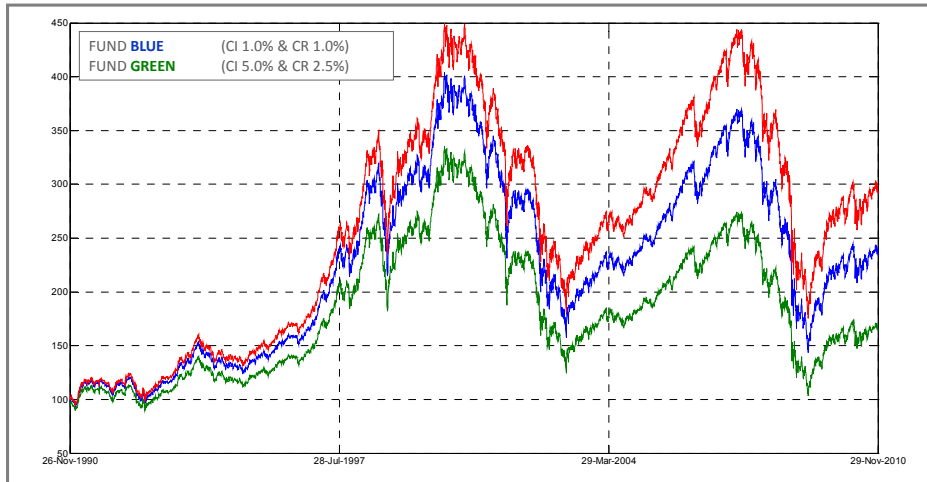


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## The recommended Investment horizon

### DETERMINAZIONE DELL'ORIZZONTE TEMPORALE

STOXX EUROPE 600  
(26/11/1990 – 26/11/2010, BASE 100: 1990)

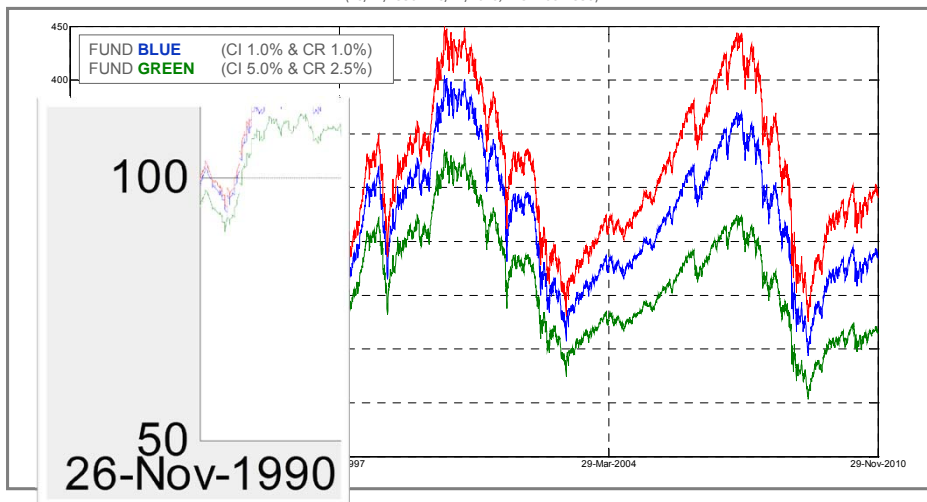


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## The recommended Investment horizon

### DETERMINAZIONE DELL'ORIZZONTE TEMPORALE

STOXX EUROPE 600  
(26/11/1990 – 26/11/2010, BASE 100: 1990)

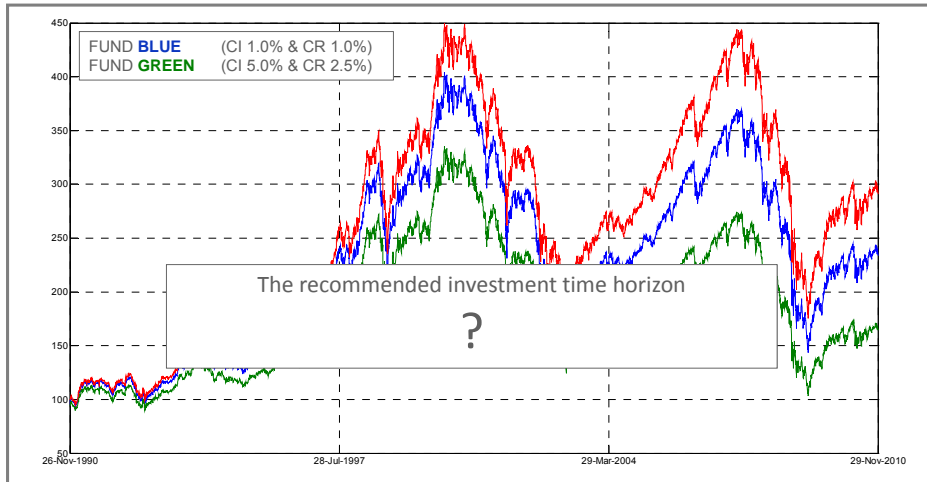


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## The recommended Investment horizon

### DETERMINAZIONE DELL'ORIZZONTE TEMPORALE

STOXX EUROPE 600  
(26/11/1990 – 26/11/2010, BASE 100: 1990)



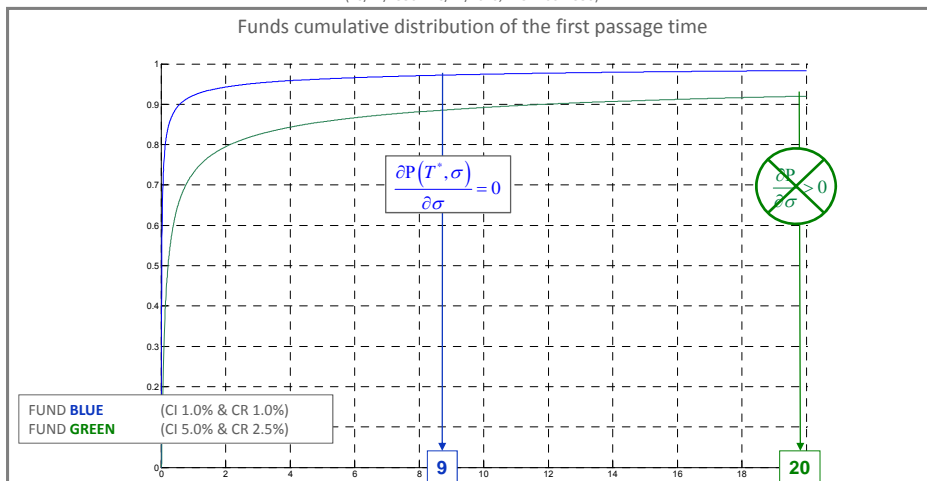
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## The recommended Investment horizon

### DETERMINAZIONE DELL'ORIZZONTE TEMPORALE

STOXX EUROPE 600  
(26/11/1990 – 26/11/2010, BASE 100: 1990)

Funds cumulative distribution of the first passage time



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## Syllabus

- Preliminaries: the three pillars
- Unbundling and Probabilistic performance scenarios
- Synthetic risk indicator
- The optimal time horizon
- An Application of the methodology

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## Examples

### DEFAULTABLE BOND

DESCRIPTION Senior bond with a 5 year maturity, paying bi-annual step-up coupons ranging from 4.7% to 5.30%.

**Financial investment table  
(Price Unbundling)**

A	Theoretical value of the Risk-Free component	91.3
B	Theoretical value of the Risky component	5
C = A + B	<i>Fair value</i>	96.3
D	Costs	3.7
E = C + D	Issue price	100

1st PILLAR

**Table of the probabilistic performance scenarios**

SCENARIOS	PROBABILITY	MEAN VALUES
The performance is <u>negative</u>	9.5%	49.3
The performance is <u>positive but lower</u> than the risk-free asset	0.0%	-
The performance is <u>positive and in line</u> with the risk-free asset	87.4%	115.6
The performance is <u>positive and higher</u> than the risk-free asset	3.1%	131.1

2nd PILLAR Degree of Risk: Medium-High

3rd PILLAR Recommended investment time horizon: 5 years

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## Examples



### VPPI PRODUCT

DESCRIPTION VPPI technique is aimed at protecting the initial value of the financial investment over a specified time horizon and obtaining possible gains by limited exposure to the equity markets.

1st PILLAR

#### Financial investment table (Price Unbundling)

A	Theoretical value of the Risk-Free component	90.1
B	Theoretical value of the Risky component	6.4
C = A + B	Fair value	96.5
D	Costs	3.5
E = C + D	Issue price	100

#### Table of the probabilistic performance scenarios

SCENARIOS	PROBABILITY	MEAN VALUES
The performance is <u>negative</u>	36.9%	96.9
The performance is <u>positive but lower</u> than the risk-free asset	18.5%	101
The performance is <u>positive and in line</u> with the risk-free asset	39.9%	107.1
The performance is <u>positive and higher</u> than the risk-free asset	4.7%	195.5

2nd PILLAR

Degree of Risk: Medium

3rd PILLAR

Recommended investment time horizon: 5 years

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## Examples



### INDEX LINKED CERTIFICATE

DESCRIPTION The index-linked certificate is characterised by a complex financial engineering that makes intensive use of diverse derivatives components. These derivatives link the performances of the product to the variability of an equity index.

1st PILLAR

#### Financial investment table (Price Unbundling)

A	Theoretical value of the Risk-Free component	86.2
B	Theoretical value of the Risky component	9.9
C = A + B	Fair value	96.1
D	Costs	3.9
E = C + D	Issue price	100

#### Table of the probabilistic performance scenarios

SCENARIOS	PROBABILITY	MEAN VALUES
The performance is <u>negative</u>	18.9%	49.1
The performance is <u>positive but lower</u> than the risk-free asset	0.0%	-
The performance is <u>positive and in line</u> with the risk-free asset	68.9%	120.9
The performance is <u>positive and higher</u> than the risk-free asset	12.2%	131.6

2nd PILLAR

Degree of Risk: High

3rd PILLAR

Recommended investment time horizon: 5 years

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## Testimonials

*"This book fills the gap that exists between the risk management tools available to industry insiders, and those available to investors. It is a welcome contribution that will be helpful to anyone who needs to assess the risk of non-equity products."*

**Jaksa Cvitanic, Professor of Mathematical Finance, Caltech**

*"Rigor and clarity characterize this methodology to assess the risk of every non-equity product. Well established stochastic techniques are applied in an original way to convey the key information on the time horizon, the degree of risk, the costs and potential returns of the investment and therefore to match the investor's preferences in terms of liquidity attitude, risk taking, desired returns and acceptable losses."*

**Prof. Svetlozar Rachev, Department of Statistics and Applied Probability, University of California at Santa Barbara**

*"I warmly welcome the publication of this book which describes a probabilistic framework for risk evaluation. The specific aim is that of providing financial institutions and regulators with tools and techniques for an objective and clear representation of key investor information. This shall help in orientating buyers through the difficult path of non-equity products selection."*

**Prof. Francesco Corielli, Department of Finance, Bocconi University**

*"This book constitutes an excellent collection of quantitative methods to the measurement and representation of the risks of non-equity products that comes from a simple but also winning intuition: the information needs of retail investors are not really different from those of financial institutions since they both want the upside gain by trying to contain the downside risk."*

**Prof. Hélyette Geman, School of Business, Economics and Informatics, Birkbeck, University of London**

*"This important book establishes a benchmark for a future financial regulation based on quantitative techniques. At the same time it casts a serious challenge to the financial industry on the need of quantitative disclosure, that will be the future of the financial system worldwide. Hope the challenge will be accepted."*

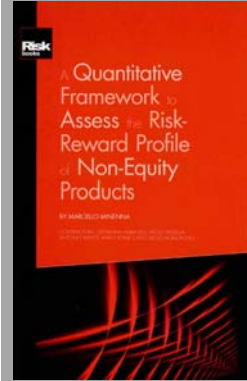
**Prof. Umberto Cherubini, Department of Mathematical Economics, University of Bologna**

*"This book contains a valid quantitative methodology to shed light on the risks embedded in any non-equity product. By answering the key questions of any investor about the potential performances, the risk rating and the optimal holding time of the product, the three "pillars" of the book are the best candidates to definitely remove the informative lack that worldwide regulators have recognized in the existing rules on risks disclosure. The adoption of these "pillars" would be the ideal completion of the regulatory reform undertaken by the European Authorities regarding the revision of the information contents for Packaged Retail Investment Products. Should the quantitative framework set forth in this work become the reference to update the regulatory framework on transparency, an authentic reversal of the traditional approaches to risks transparency would be realized with effective benefits for investors' comprehension and for allowing them to pick the product that best fits their needs."*

**Prof. Riccardo Cesari, Professor of Mathematical Methods for Economic and Financial Sciences, University of Bologna**

*"This innovative book sheds a light on the dark path of the financial risks intrinsic to non-equity financial products, which are often underestimated, or even poorly understood, by investors seeking higher returns. Mathematical finance techniques are here applied in an original and unconventional manner for the purpose of effectively disclosing these risks and properly assessing their impact on investments' returns."*

**Fabio Mercurio, Head of Quant Business Managers at Bloomberg LP and adjunct professor at NYU**



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