

# *Portfolio Construction*

# Agenda

- 1. Introduction to the “Portfolio Construction”**
- 2. Analysis of Financial Markets**
- 3. Strategic Asset Allocation: Naïve Portfolio Formation Rule**
- 4. Strategic Asset Allocation: A Quantitative Approach**
- 5. Naïve *versus* Markowitz**
- 6. “Putting Markowitz at work”**
- 7. Heuristic techniques**
- 8. Bayesian techniques**

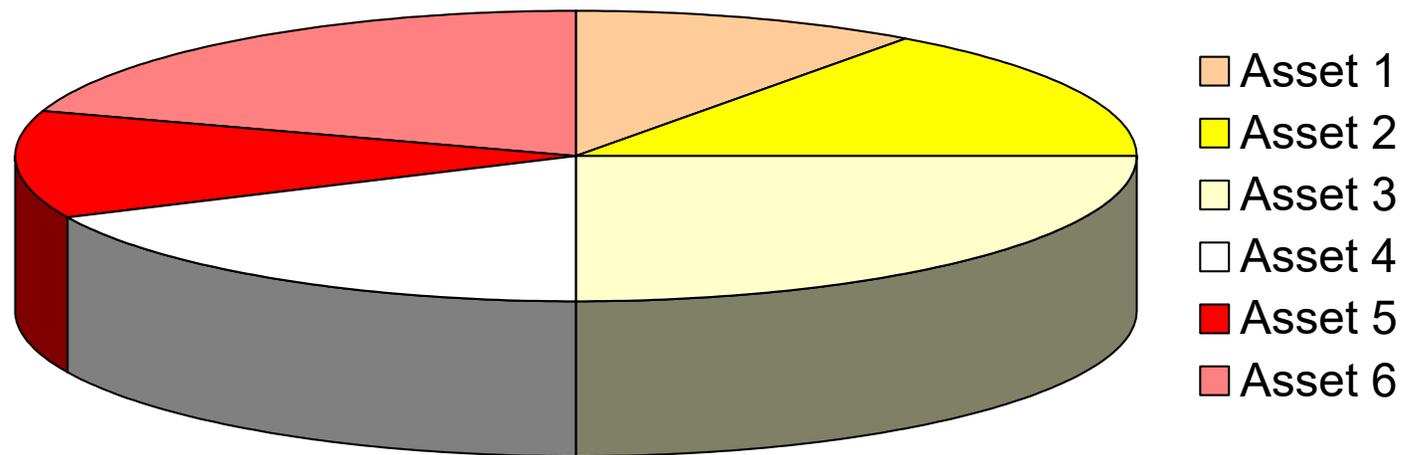
# Agenda

1. Introduction to the “Portfolio Construction”

# Portfolio construction: Is it easy?

From a superficial point of view,  
building a portfolio is **easy**:

- you have just to **aggregate different assets**



Portfolio construction: Is it easy?

The answer is **NO**

.....Unfortunately building a **GOOD** portfolio that is able to satisfy the needs/expectations of an investor is **HARD**.

In fact, in order to construct a good portfolio, you have to make many **difficult decisions**.

# Portfolio Construction: Problems to be solved

- Selection of Financial Markets where “investing money” (Money, Bond, Stock, etc...) : Which? How many?
- Estimation of future trend of the markets selected.
- Construction of an optimization model that returns the optimal portfolio (that returns the optimal weights of the Financial Market) in the long run.
- Development of an evaluation model that it is able to verify that the portfolio selected satisfies the needs of the investor.
- Development of a “market timing” model, useful in order to make tactical changes to the portfolio composition, in order to anticipate bull/bear trends.
- Selection of the best financial products for every financial market.

# A mistake? Very dangerous!

- Adverse selection of Financial Markets (too much risky or poorly performing markets)
- Error in Estimation of future trend of the markets selected or overconfidence in the ability of prediction.
- Construction of a weak optimization model.
- Incapacity to verify that the portfolio selected satisfies the needs of the investor.
- Errors in “market timing” decisions (increase of the equity weight at the beginning of a *bear trend*).
- Adverse selection of products for every financial market (poor performance or high costs).

# Don't under-estimate the process of portfolio construction

- Mistakes can be “deadly”.
- So, it is necessary to have:
  - skilled human resources;
  - good IT procedures;
  - consistent models of Portfolio Construction.

## Well-Organized procedures

- Institutional investors (pension funds, mutual funds, etc...) organize the process of portfolio construction on stages...
- ...where every phase is able to create (extra-performance) or destroy value (under-performance);
- The main stages are three.

# Stages of the “Portfolio Construction”

1. Strategic Asset Allocation

+

2. Tactical Asset Allocation

+

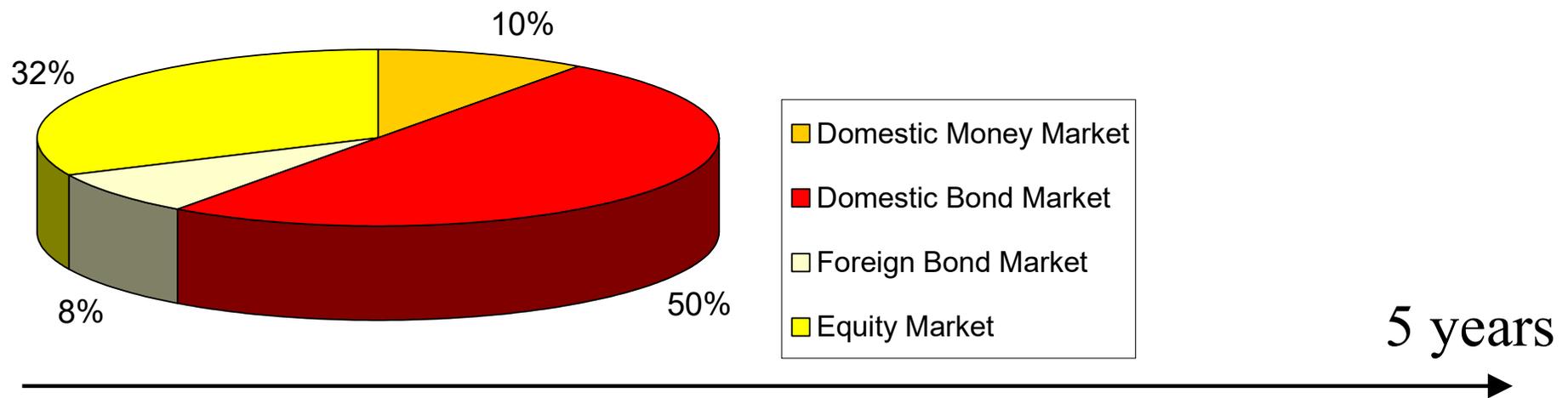
3. Stock-Bond/Fund Selection

# Stage 1: Strategic Asset Allocation (SAA)

- **Strategic Asset Allocation** is:
  - the portfolio composed by **financial markets** (or asset classes).....
  - ..... that the investor must hold **in the long run** (all the investment horizon).

# Strategic Asset Allocation: Example

- The investor has a 5-years investment horizon.
- The Asset Manager builds a portfolio, composed by financial markets (*that is supposed to be coherent with the risk tolerance of the investor*).



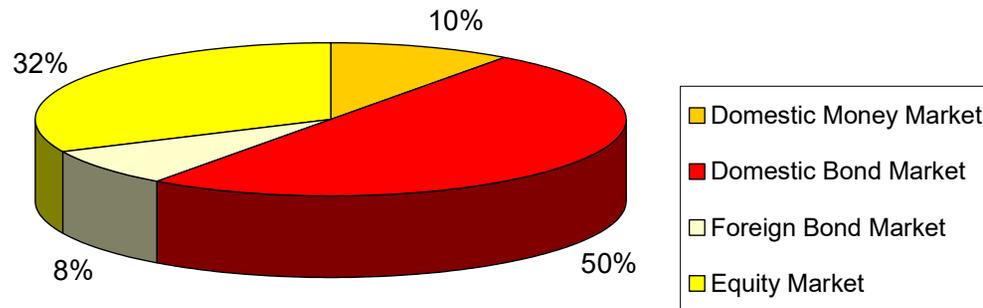
“On average” the portfolio composition is expected to be this one in the next five years.

## Stage 2: Tactical Asset Allocation (TAA)

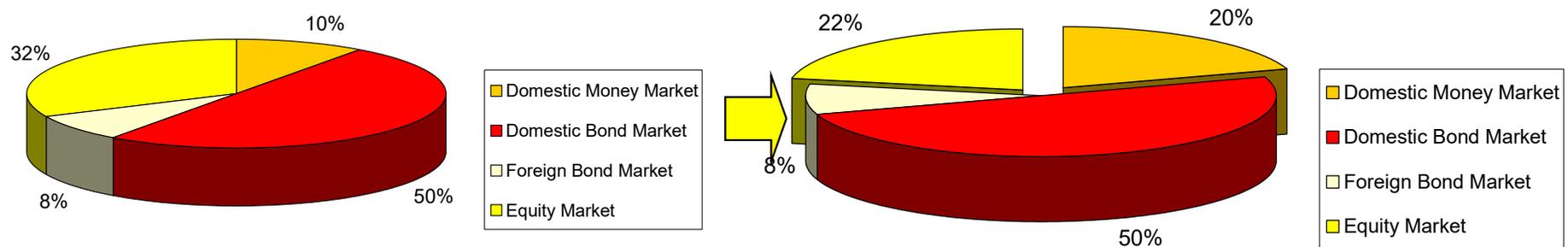
- **Tactical Asset Allocation is:**
  - The change made to the strategic composition in order to anticipate bull/bear trends.
  - ..... that the investor must hold **in the short run** (next 1-3 months).

# Tactical Asset Allocation: Example

- The SAA is the following:



- But the Asset Manager has the expectation that in the next 3 months the Equity Market will decrease.
- So, for the next 3 months he suggests the following changes in the portfolio composition:



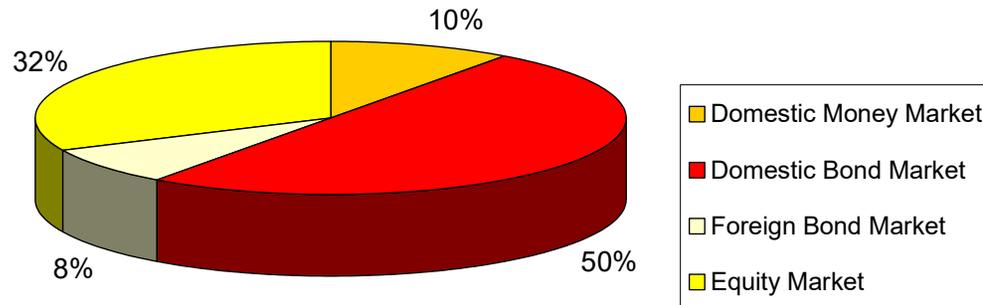
After three months, the tactical portfolio will be dismantled (and the strategic portfolio will be resumed).....may be we will create e new tactical solution.

# Stage 3: Stock-Bond/Fund Selection

- **Stock-Bond/Fund Selection** is:
  - the process to select the best product for every market in the portfolio.
  - You can (alternatively):
    - directly select stocks & bonds (**stock-bond selection**);
    - indirectly select stocks & bonds, identifying the best fund managers (**fund selection**).

# Stock-Bond/Fund Selection : Example (1/2)

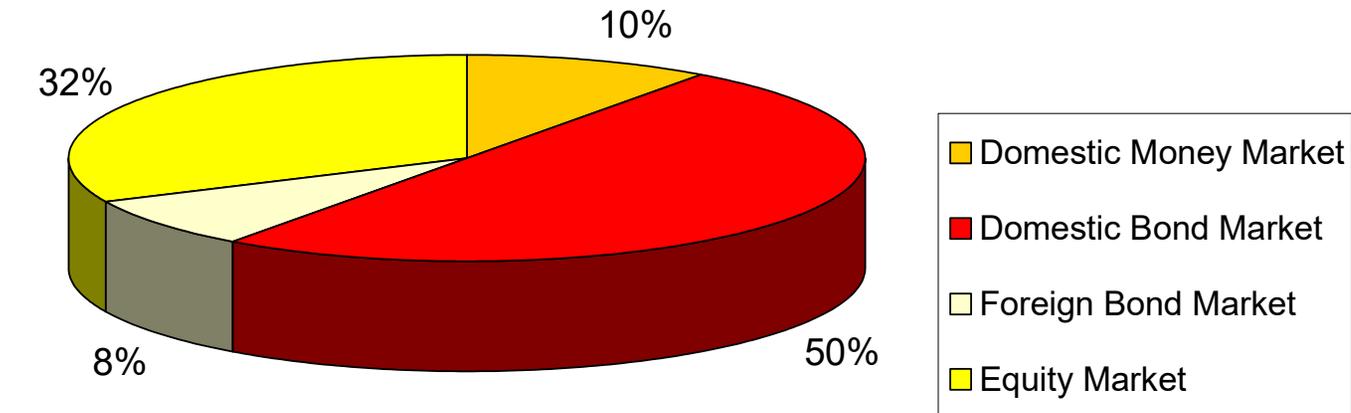
- A French Pension Fund has the following SAA:



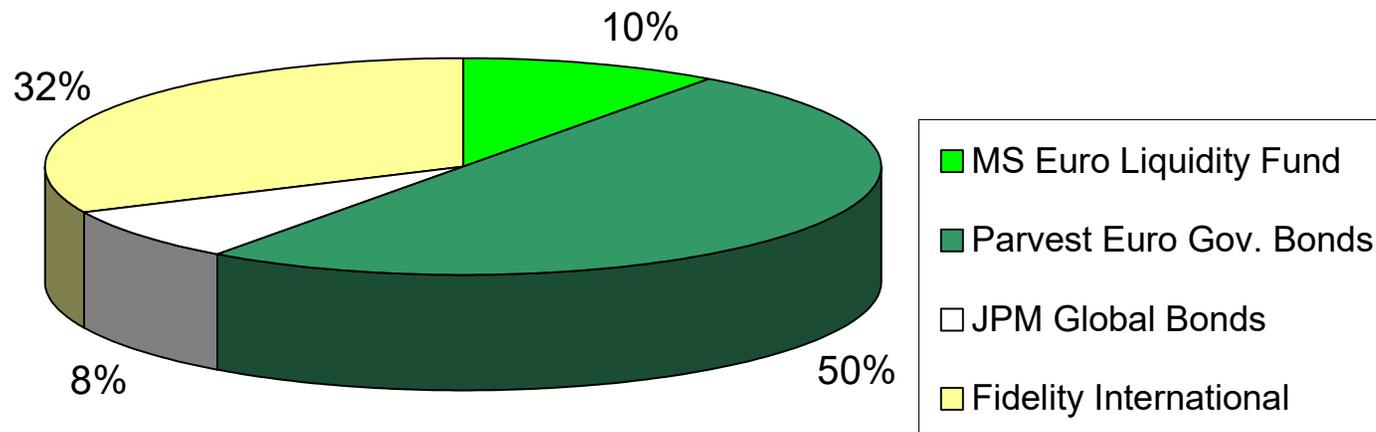
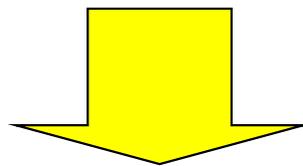
- The board of directors hasn't the ability to directly select the stocks/bonds....
- so the Pension Fund identifies, for every market, the fund managers that are supposed to be the best ones:

<b>Markets (Asset Classes)</b>	<b>Funds selected</b>
Domestic Money Market	MS Euro Liquidity Fund
Domestic Bond Market	Parvest Euro Gov. Bonds
Foreign Bond Market	JPM Global Bonds
Equity Market	Fidelity International

# Stock-Bond/Fund Selection : Example (2/2)



From  
Markets....



.....to  
Products.

# The “pillars” of Asset Allocation

Investor:

**Investor’s preferences**

Asset Manager:

**Asset Manager’s expectations  
about the future trend of  
Financial Markets**

**Optimization Model**

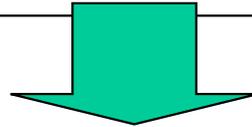
***OPTIMAL PORTFOLIO***

# Agenda

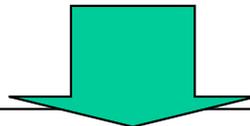
## 2. Analysis of Financial Markets

# From Investor's Preferences to Financial Markets Evaluation

- It is well known that Investors:
  - love return;
  - hate risk (are *risk adverse* ).



So, if we want to build a portfolio that best suit the investor's preferences, we need to know the *risk-return* profile of Financial Markets and Market Portfolio.



Risk-Return analysis of Financial Markets

# The Financial Markets

Analysis of the following Asset Classes.

## ASSET CLASSES:

- Money Market EMU
- Bond Market EMU
- Bond Market World
- Equity Market Europe
- Equity Market North America
- Equity Market Japan
- Equity Market Pacific ex Japan
- Equity Emerging Markets

## MARKET INDEXES:

- JPM Euro 3 Months
- Citygroup EMU Aggr. all maturities
- JPM Global
- MSCI Europe
- MSCI North America
- MSCI Japan
- MSCI Pacific ex Japan
- MSCI Emerging Markets

# Historical series of annual returns

Thanks to the market indexes we have a set of historical returns of financial markets:

	JPM Euro 3 months	Citygroup EMU All Maturities	JPM Global	MSCI Europe	MSCI North America	MSCI Japan	MSCI Pacific ex Japan	MSCI Emerging Markets
1988	7,28%	4,30%	17,89%	26,59%	25,56%	51,41%	40,98%	51,45%
1989	9,16%	1,40%	1,84%	19,57%	20,59%	-3,38%	6,05%	51,78%
1990	11,55%	3,10%	-0,96%	-17,12%	-17,05%	-43,67%	-24,16%	-23,58%
1991	10,41%	11,37%	17,13%	11,58%	27,60%	9,86%	32,86%	58,26%
1992	11,11%	12,80%	11,33%	-1,35%	9,64%	-17,04%	10,01%	16,11%
1993	9,03%	14,44%	20,41%	35,53%	15,27%	33,65%	88,14%	83,69%
1994	6,30%	-1,84%	-9,60%	-10,63%	-11,72%	7,76%	-25,11%	-18,48%
1995	6,58%	16,27%	10,18%	9,77%	23,41%	-7,64%	1,77%	-14,07%
1996	4,83%	7,29%	12,41%	27,59%	30,93%	-9,54%	26,84%	11,89%
1997	4,42%	6,16%	18,31%	41,85%	52,36%	-11,52%	-21,47%	1,03%
1998	4,46%	10,94%	6,82%	17,21%	17,75%	-3,41%	-16,21%	-32,85%
1999	3,15%	-2,97%	10,67%	33,06%	42,14%	87,20%	62,47%	90,86%
2000	4,32%	8,39%	10,49%	-2,46%	-5,84%	-22,85%	-10,91%	-26,37%
2001	4,74%	6,25%	4,75%	-16,83%	-8,77%	-25,97%	-7,26%	0,40%
2002	3,53%	8,49%	0,31%	-32,86%	-35,81%	-25,17%	-23,53%	-22,66%
2003	2,54%	3,77%	-4,92%	11,92%	6,09%	11,76%	17,29%	25,87%
2004	2,18%	7,56%	2,09%	9,28%	1,40%	6,38%	15,57%	13,54%
2005	2,20%	5,67%	7,93%	23,01%	21,23%	43,29%	27,27%	50,45%
2006	3,02%	-0,28%	-5,11%	16,65%	1,53%	-5,87%	14,68%	15,72%
2007	4,42%	0,97%	-0,86%	-0,73%	-5,46%	-15,38%	13,77%	22,10%
2008	5,75%	9,97%	18,47%	-45,21%	-35,73%	-26,51%	-49,45%	-51,85%

(in Euro)

➤ **Investors aim to maximise the return (the final value) of their investments**

# Average Return (1/2)

Investors love financial markets with higher average returns:



$$\bar{R} = \frac{(R_1 + R_2 + \dots + R_{T-1} + R_T)}{T}$$



$$\bar{R} = \frac{\sum_{t=1}^T R_t}{T}$$

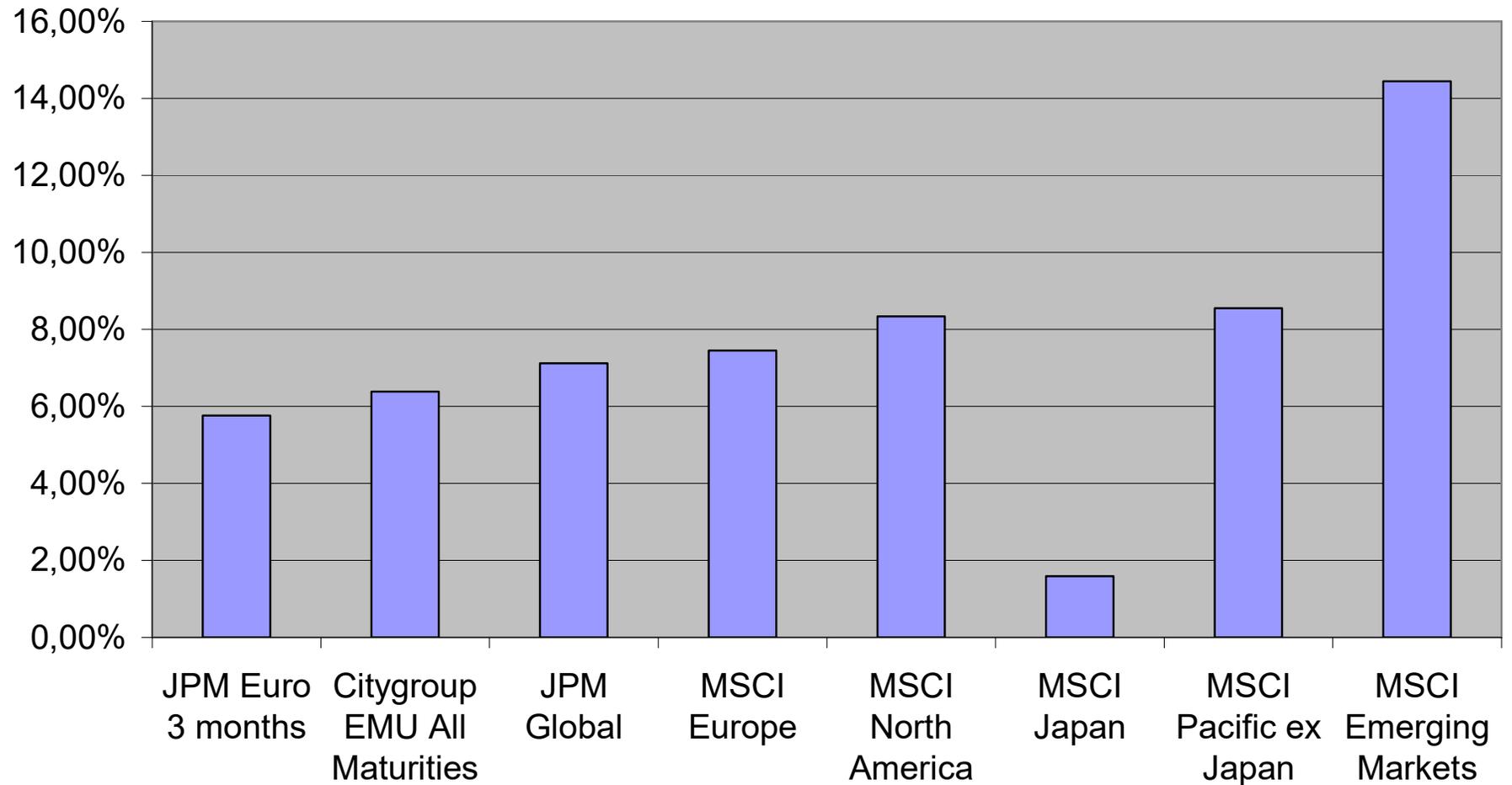
Excel:

=average(Historical series)

(1988-2008)	JPM Euro 3 months	Citygroup EMU All Maturities	JPM Global	MSCI Europe	MSCI North America	MSCI Japan	MSCI Pacific ex Japan	MSCI Emerging Markets
Average of Annual Returns	5,76%	6,38%	7,12%	7,45%	8,34%	1,59%	8,55%	14,44%

# Average of Annual Returns (2/2)

## Average of Annual Returns (1988-2008)



# Average Return of a Portfolio (1/3)

If we know:

- the Portfolio Weight of each market ( $w_i$ );
- the Average Returns of each market ( $\bar{R}_i$ )

The estimation of the Portfolio Average Return is straightforward:

$$\bar{R}_{Port} = \sum_{i=1}^k w_i \times \bar{R}_i$$

Excel:

=sumproduct(Weights, Average Returns)

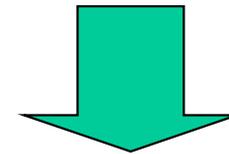
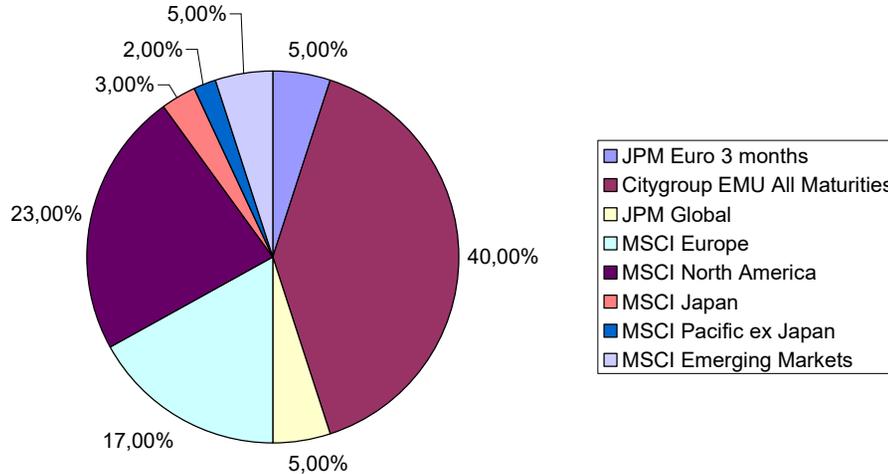
## Average Return of a Portfolio (2/3)

Using Matrices:

$$\bar{R}_{Port} = \begin{bmatrix} w_1 & w_2 & \cdots & \cdots & w_i & \cdots & w_k \end{bmatrix} \times \begin{bmatrix} \bar{R}_1 \\ \bar{R}_2 \\ \vdots \\ \vdots \\ \bar{R}_i \\ \vdots \\ \bar{R}_k \end{bmatrix}$$

# Average Return of a Portfolio (3/3)

	JPM Euro 3 months	Citygroup EMU All Maturities	JPM Global	MSCI Europe	MSCI North America	MSCI Japan	MSCI Pacific ex Japan	MSCI Emerging Markets
<b>Weights</b>	5,00%	40,00%	5,00%	17,00%	23,00%	3,00%	2,00%	5,00%
<b>Average of Annual Returns</b>	5,76%	6,38%	7,12%	7,45%	8,34%	1,59%	8,55%	14,44%



$$\bar{R}_{Port} = \sum_{i=1}^k w_i \times \bar{R}_i = 7.32\%$$

Investors want to maximise the average (or expected) return of the portfolio.

➤ **Investors are “risk adverse”**: given a targeted return they try to minimise the risk.

# What is “risk”?

(1/2)

The financial literature has formulated many mathematical & statistical indicators useful in order to “capture” risk.

Examples:

- Standard Deviation;
- Semi- Standard Deviation;
- Downside risk;
- Beta;
- Duration/Modified Duration;
- Value at Risk (VaR);
- Shortfall probability;
- Tracking Error Volatility (TEV).

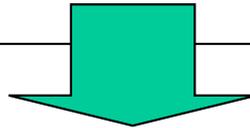


Many indicators.....maybe a phenomenon difficult to measure.

# What is “risk”?

(2/2)

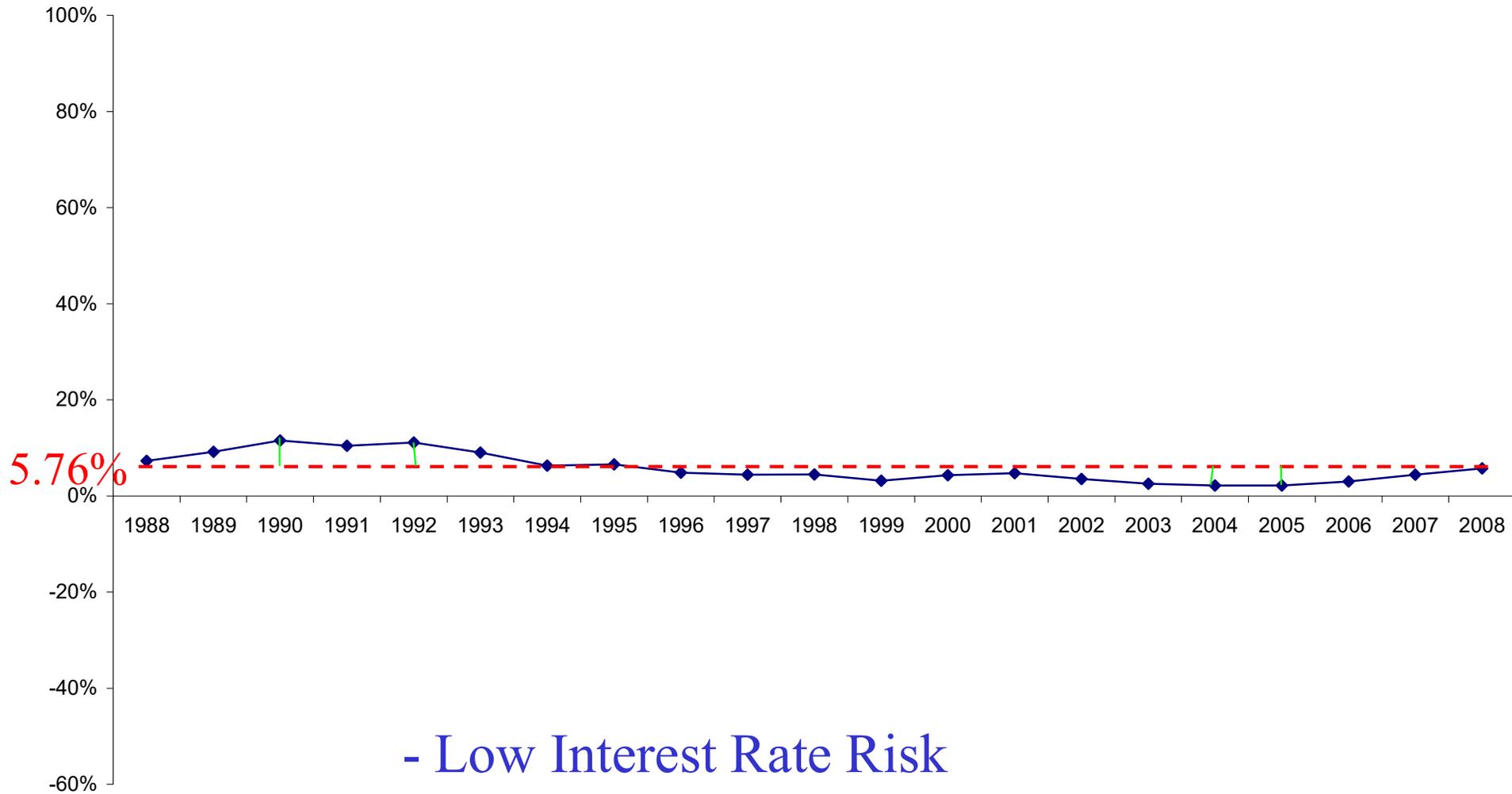
- Commonly in the financial environment risk is interpreted as the “uncertainty of returns”;
- So markets with **volatile, unstable** returns are considered risky.



Graphical evidences

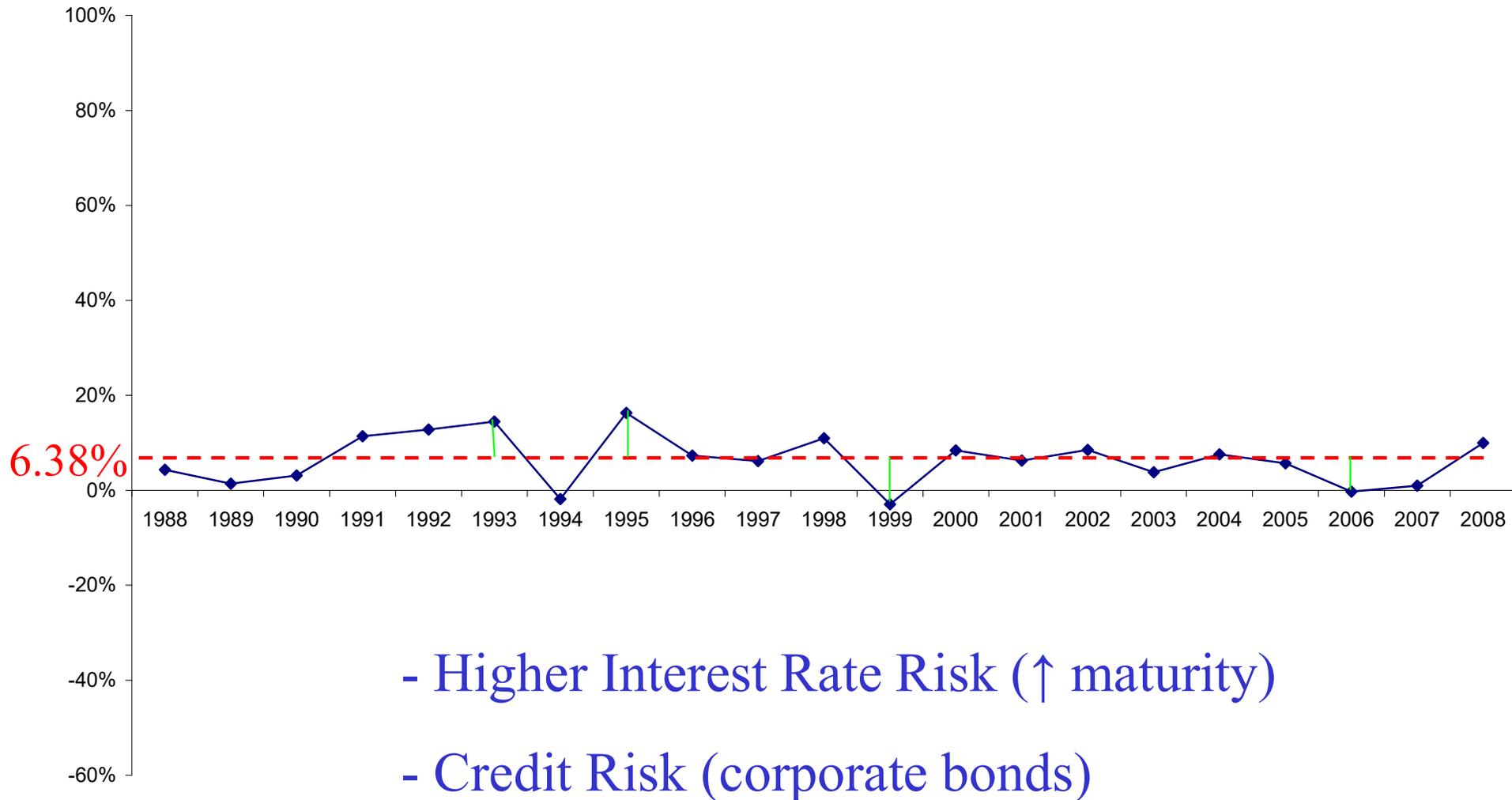
# Very Low volatility: Money Market EMU

Annual Return of JPM Euro 3 months (Money Market EMU) [1988-2008]



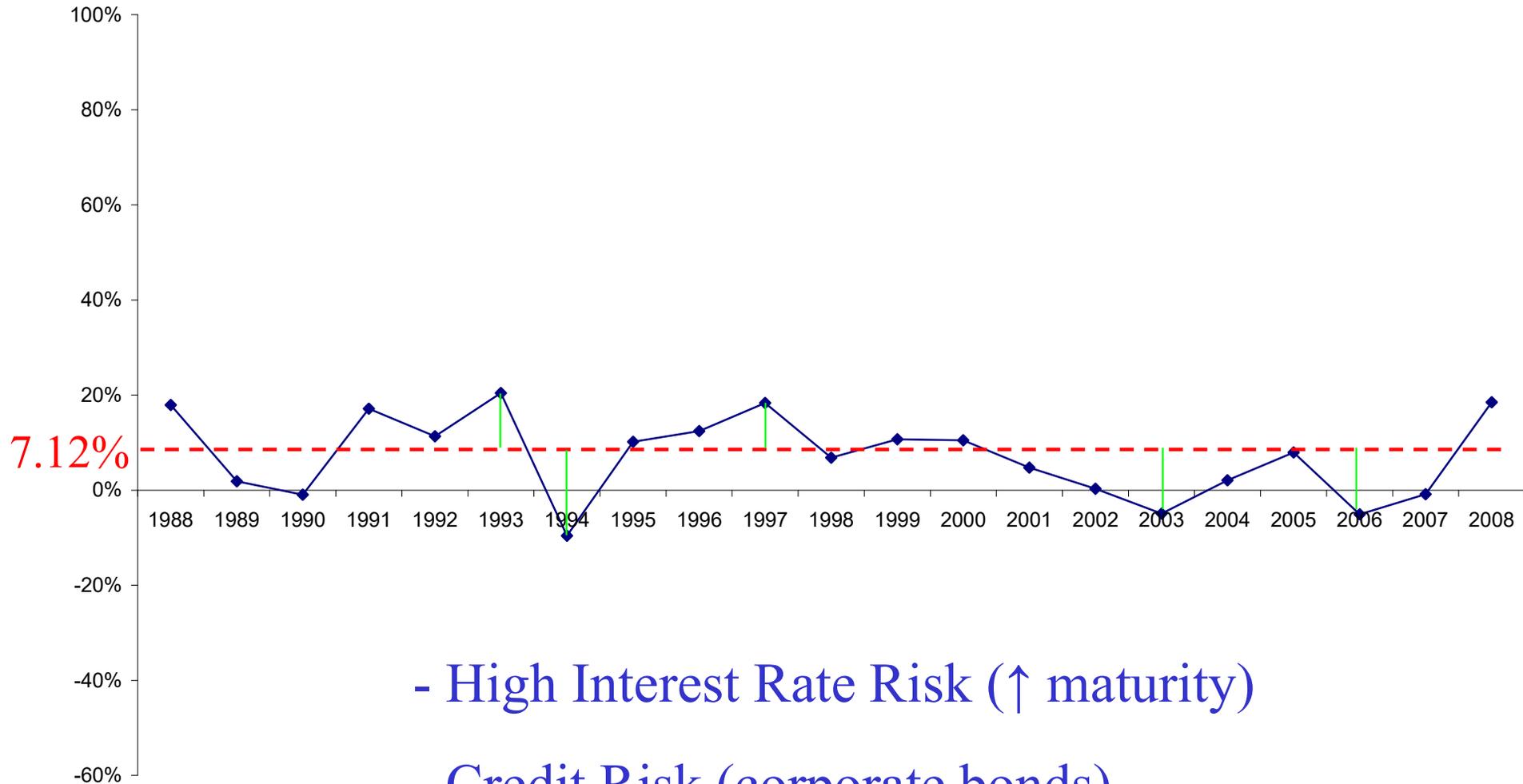
# Low volatility: Bond Market EMU

Annual Return of JCitygroup EMU All Maturities (Bond Market EMU) [1988-2008]



# Middle volatility: International Bond Market

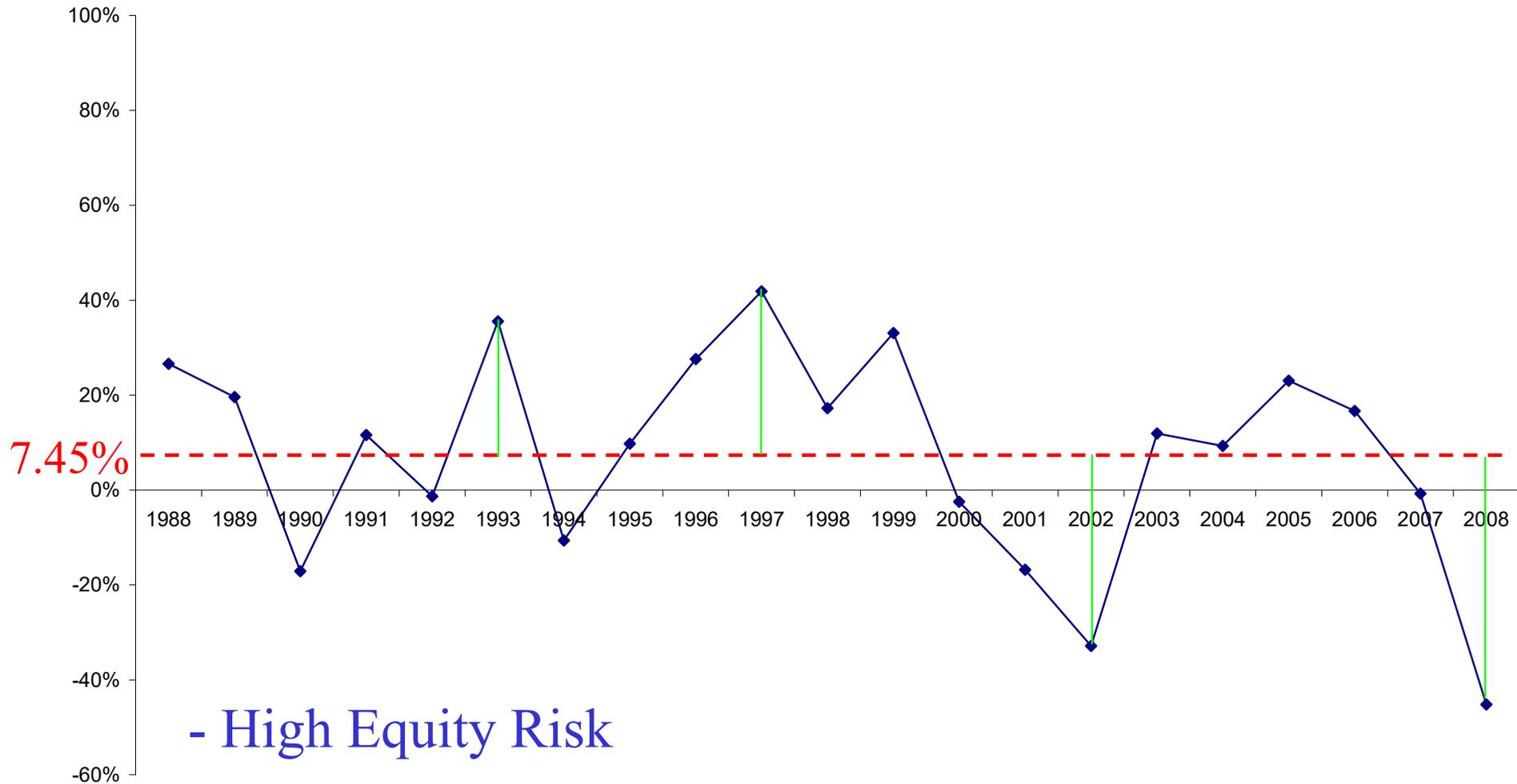
Annual Return of JPM Global (International Bond Market) [1988-2008]



- High Interest Rate Risk (↑ maturity)
- Credit Risk (corporate bonds)
- Exchange Risk

# High volatility: European Equity Market

Annual Return of MSCI Europe (European Equity Market) [1988-2008]

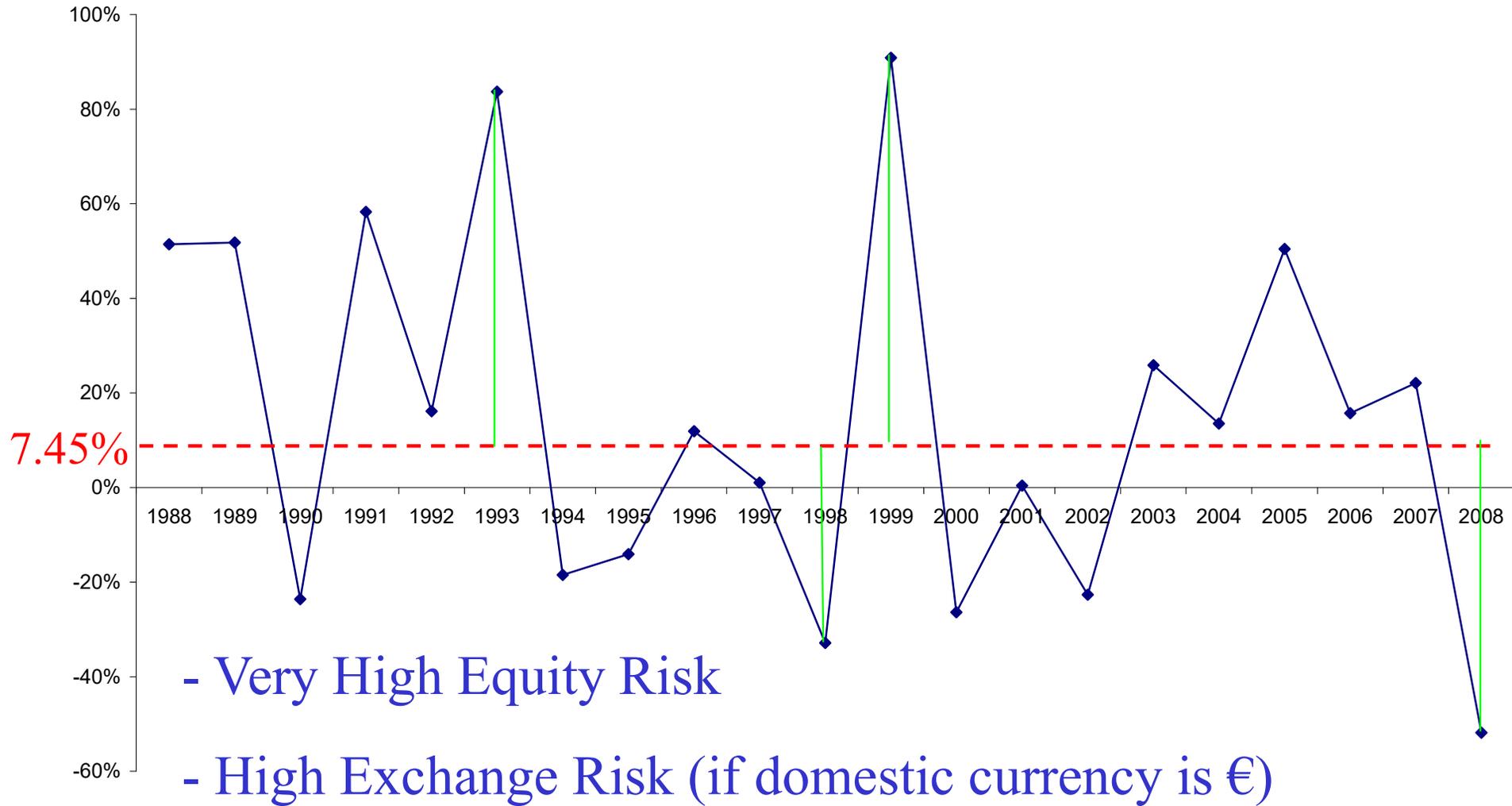


- High Equity Risk

- Very Low Exchange Risk (if domestic currency is €)

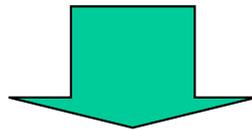
# Very High volatility: Emerg. Mkts Equity

Annual Return of MSCI EM (Emerg. Mkts Equity) [1988-2008]



# Standard Deviation of Returns

- Finally, we need a statistical indicator able to synthesise the volatility.
- The most common parameter is the:



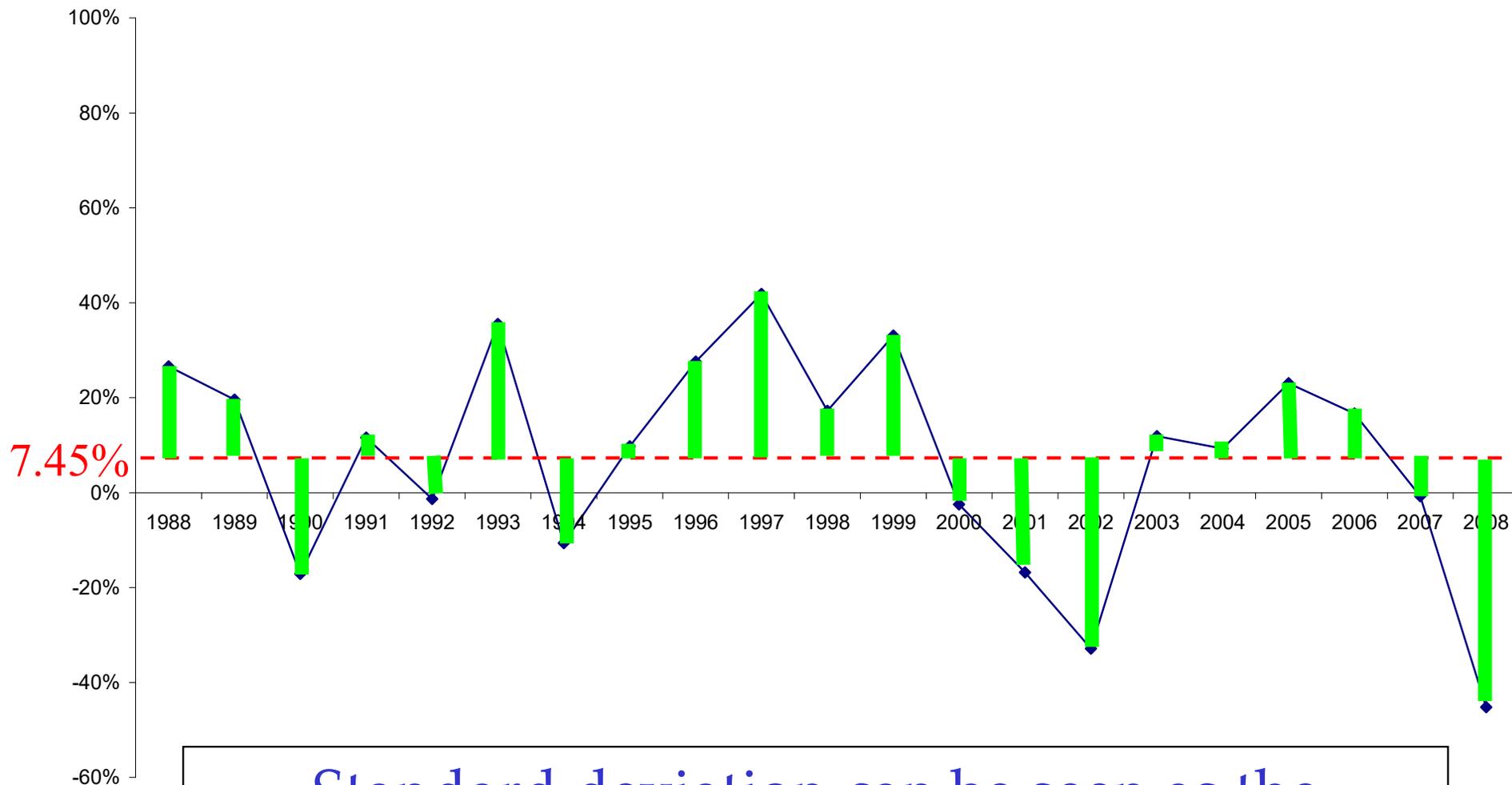
Standard deviation ( $\sigma$ )

$$\sigma = \sqrt{\frac{\sum_{i=1}^T (R_i - \bar{R})^2}{T - 1}}$$

Excel:  
=stdev(Historical series)

# $\sigma$ : an easy interpretation (1/2)

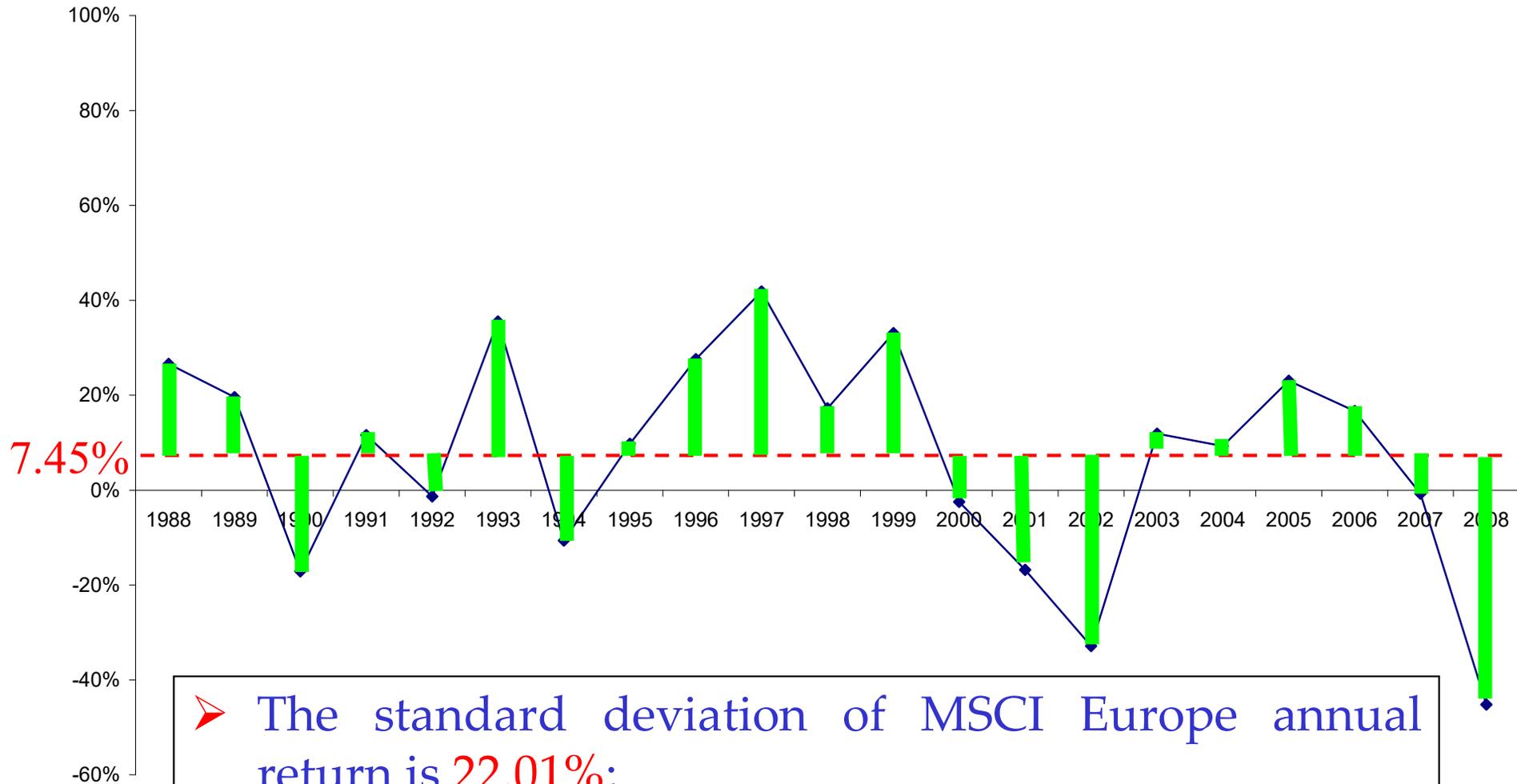
Annual Return of MSCI Europe (European Equity Market) [1988-2008]



Standard deviation can be seen as the average of “gaps” between the average return and every annual return.

# $\sigma$ : an easy interpretation (2/2)

Annual Return of MSCI Europe (European Equity Market) [1988-2008]

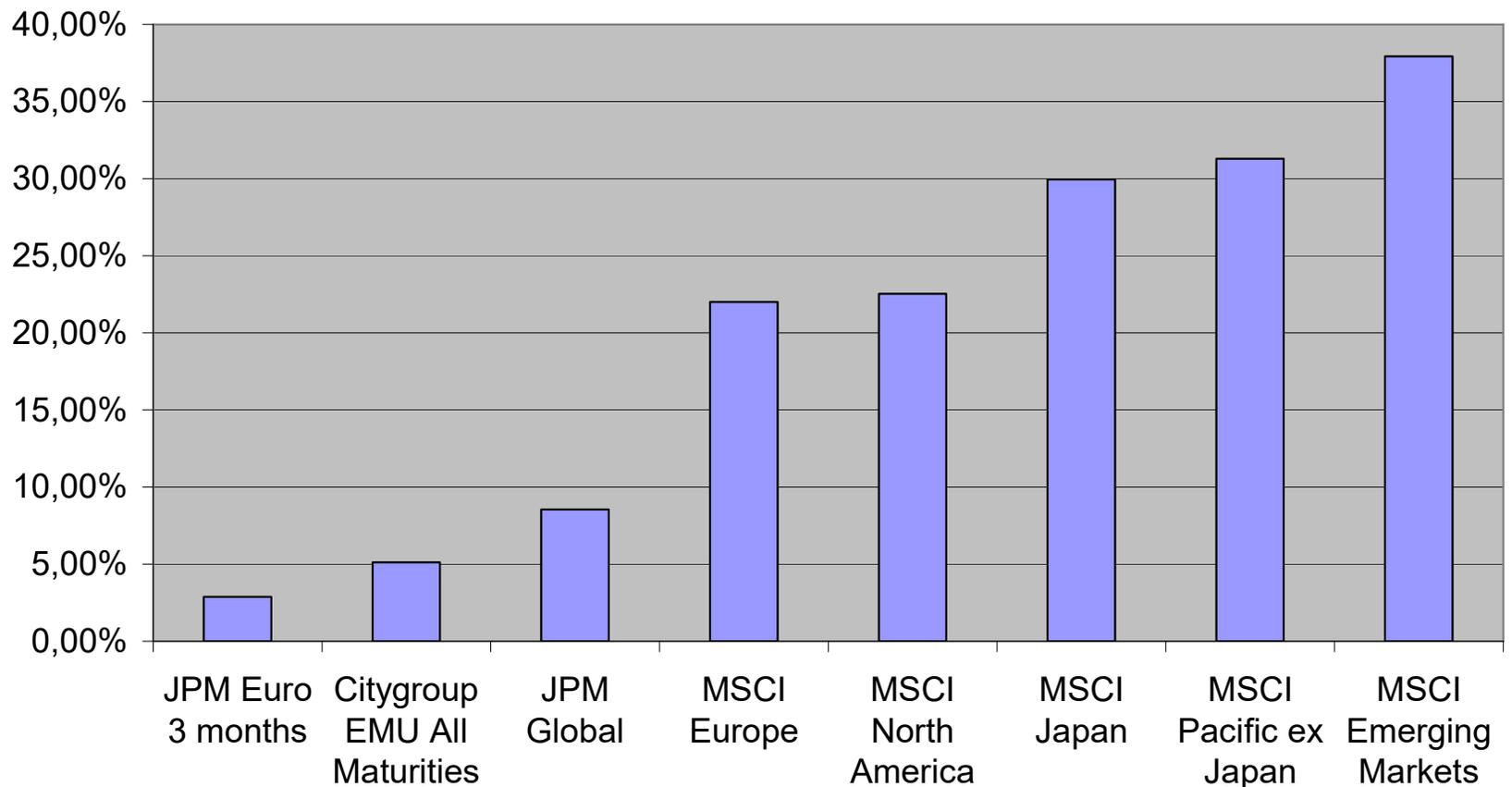


- The standard deviation of MSCI Europe annual return is **22.01%**;
- We can say that the annual return is likely to have an average deviation from the average return of **22.01%**.

# Standard Deviations of Asset Classes

(1988-2008)	JPM Euro 3 months	Citygroup EMU All Maturities	JPM Global	MSCI Europe	MSCI North America	MSCI Japan	MSCI Pacific ex Japan	MSCI Emerging Markets
$\sigma$ of Annual Returns	2,88%	5,11%	8,55%	22,01%	22,53%	29,95%	31,29%	37,93%
RANKING	1	2	3	4	5	6	7	8

**Standard Deviation of Annual Returns (1988-2008)**

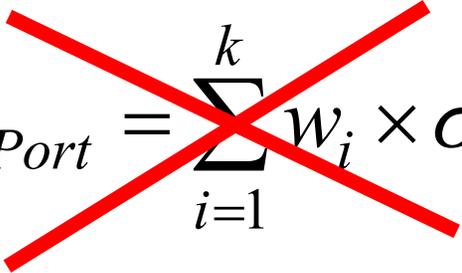


# Standard deviation of a Portfolio: NOT a weighted average (1/2)

If we know:

- the Portfolio Weight of each market ( $w_i$ )
- the standard deviations of each market ( $\sigma_i$ )

The estimation of the Portfolio standard deviation is **NOT** the following:

$$\sigma_{Port} = \sum_{i=1}^k w_i \times \sigma_i$$


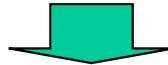
That is, The portfolio standard deviation is NOT the weighted average of the standard deviation of the markets.



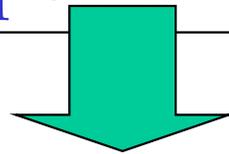
# NOT a weighted average: 1<sup>st</sup> empirical evidence

Using historical series of MSCI market indices on the time horizon 2001-2008, we measure the standard deviation of the following equity market sectors:

- MSCI Europe Pharmaceutical,  $\sigma_{\text{Pharm}} = 12,54\%$ ;
- MSCI Europe Biotechnology,  $\sigma_{\text{Biotech}} = 30,32\%$ ;



Which is the standard deviation of the MSCI Europe Pharma/Biotech?

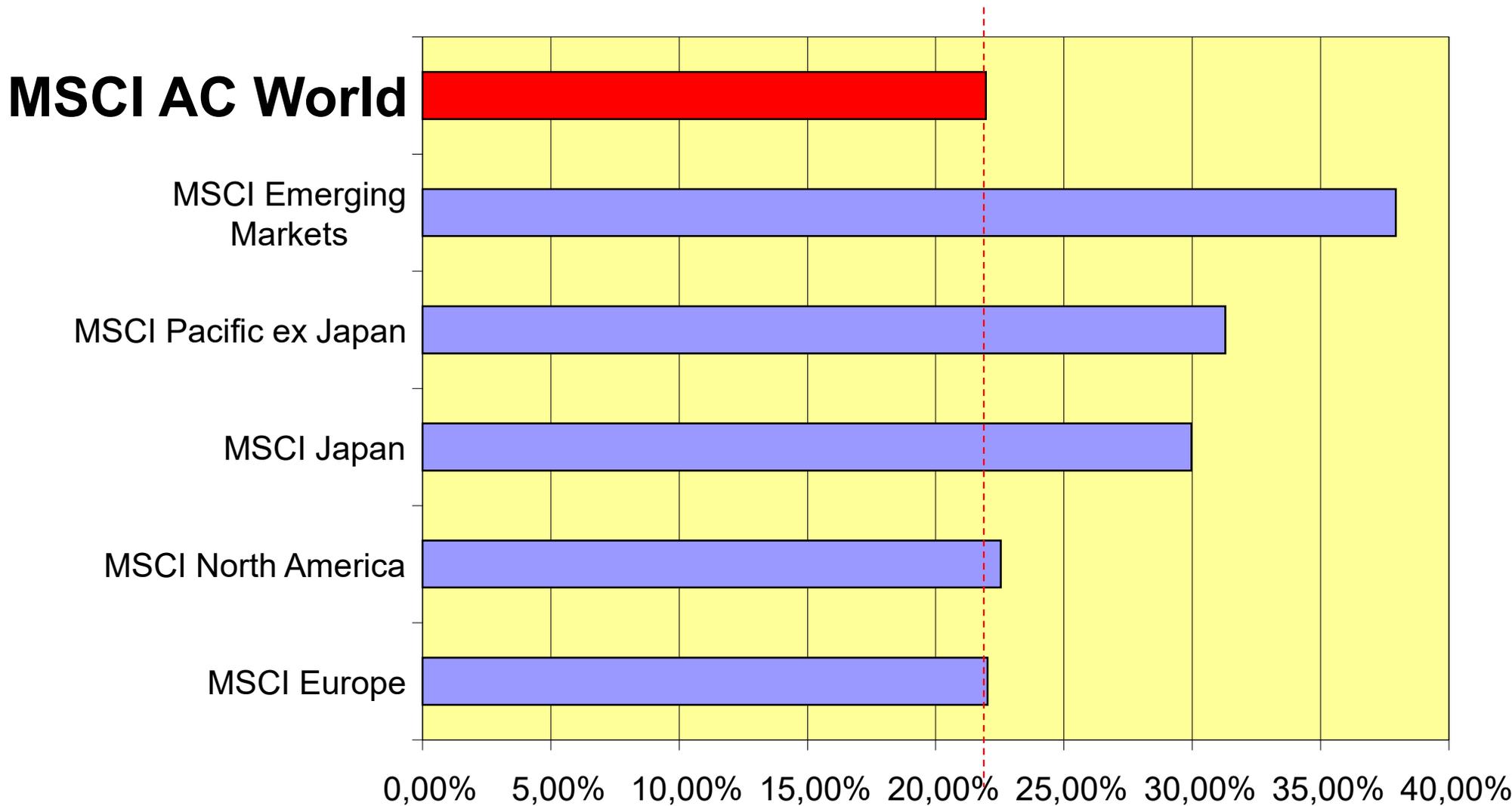


$$\sigma_{\text{Pharma / Biotech}} = 12.44\%$$

It can't be the weighted average!

# NOT a weighted average: 2<sup>nd</sup> empirical evidence

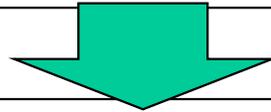
## Standard Deviation of Annual Returns (1988-2008)



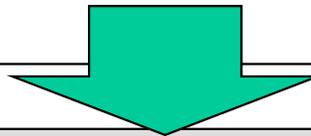
**The world equity market has a standard deviation of returns that is lower than the standard deviation of all the country markets. Again, risk can't be the weighted average.**

# The diversification effect

- Since 1952 is well known that it is possible to reduce risk avoiding concentration.
- Proverb: *"Don't put your eggs in the same basket"*



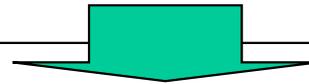
- Financial history shows that markets have the tendency to move one each other in a different way:
  - year 1998: MSCI Europe (+17.21%) vs MSCI EM (-32.85%)
  - year 1995: MSCI North America (+23.41%) vs MSCI Japan (+1.77%)



Thanks to the diversified behaviour of financial markets, the portfolio standard deviation is lower than the weighted average.

# We need to “Capture” the diversification effect

- In order to measure the diversification effect (that is, the power of diversification in reducing risk) we must measure:



The Correlation ( $\rho$ )

## Correlation ( $\rho$ ): characteristics (1/2)

- The correlation is calculated for a couple of markets;
- $-1 \leq \rho \leq +1$
- If  $\rho > 0$ , markets move in the same direction (*both gain or both lose*)
- If  $\rho = +1$ , markets **perfectly** move in the **same direction**
- If  $\rho < 0$ , markets move in opposite direction (*one gains, the other loses*)
- If  $\rho = -1$ , markets **perfectly** move **in opposite direction** (*they move perfectly synchronised, but in opposite direction*)
- If  $\rho = 0$ , markets are independent (*no tendency to move in the same or in the opposite direction*) *(follows)*

## Correlation ( $\rho$ ): characteristics (2/2)

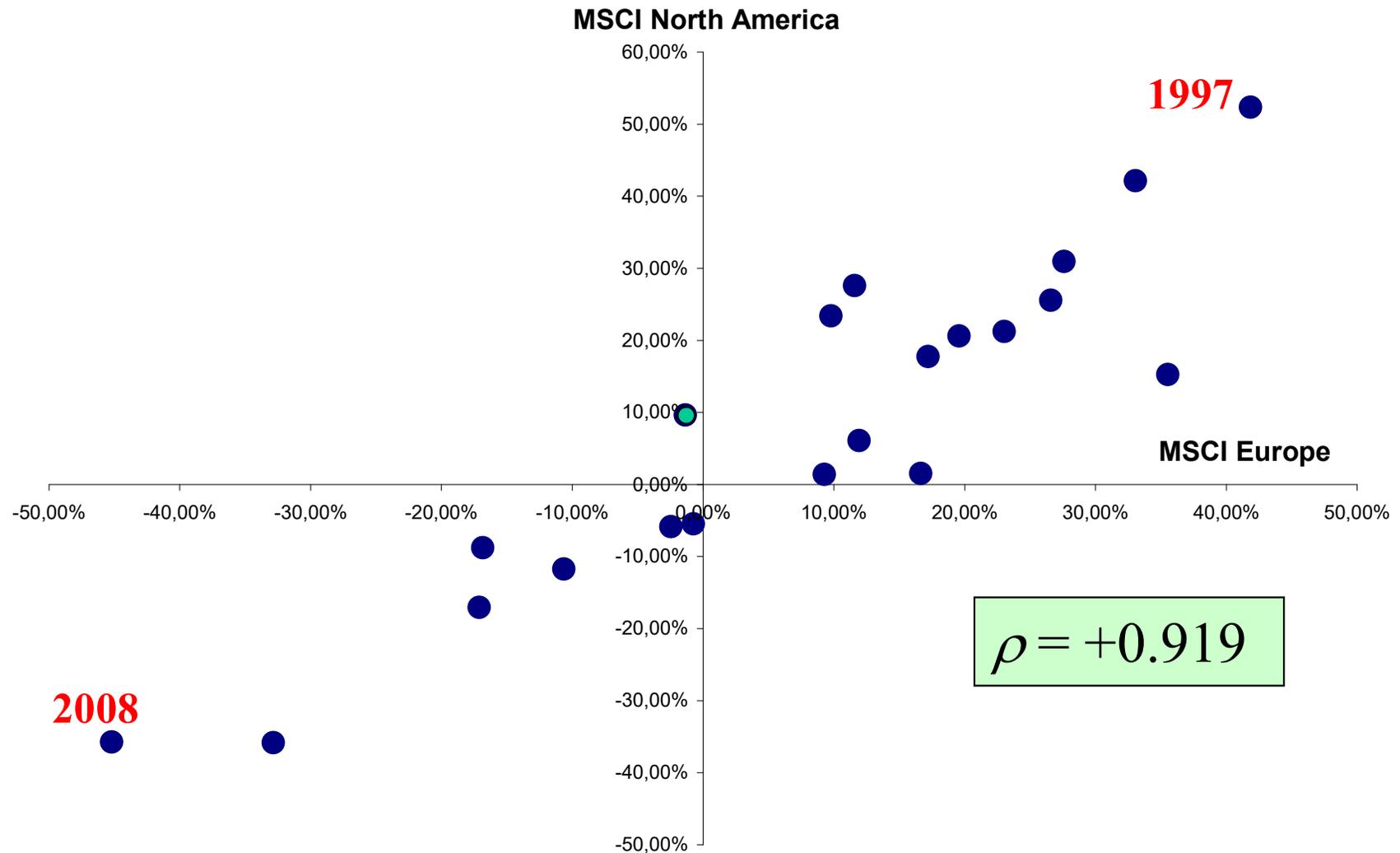
- o If  $\rho = +1$ , **no** diversification
- o If  $\rho < +1$ , **yes** diversification
- o The lower the correlation, the higher the diversification (the risk reduction)

Excel:

`=correl(Historical series mkt1, Historical series mkt2)`

# Correlation: the scatter graph

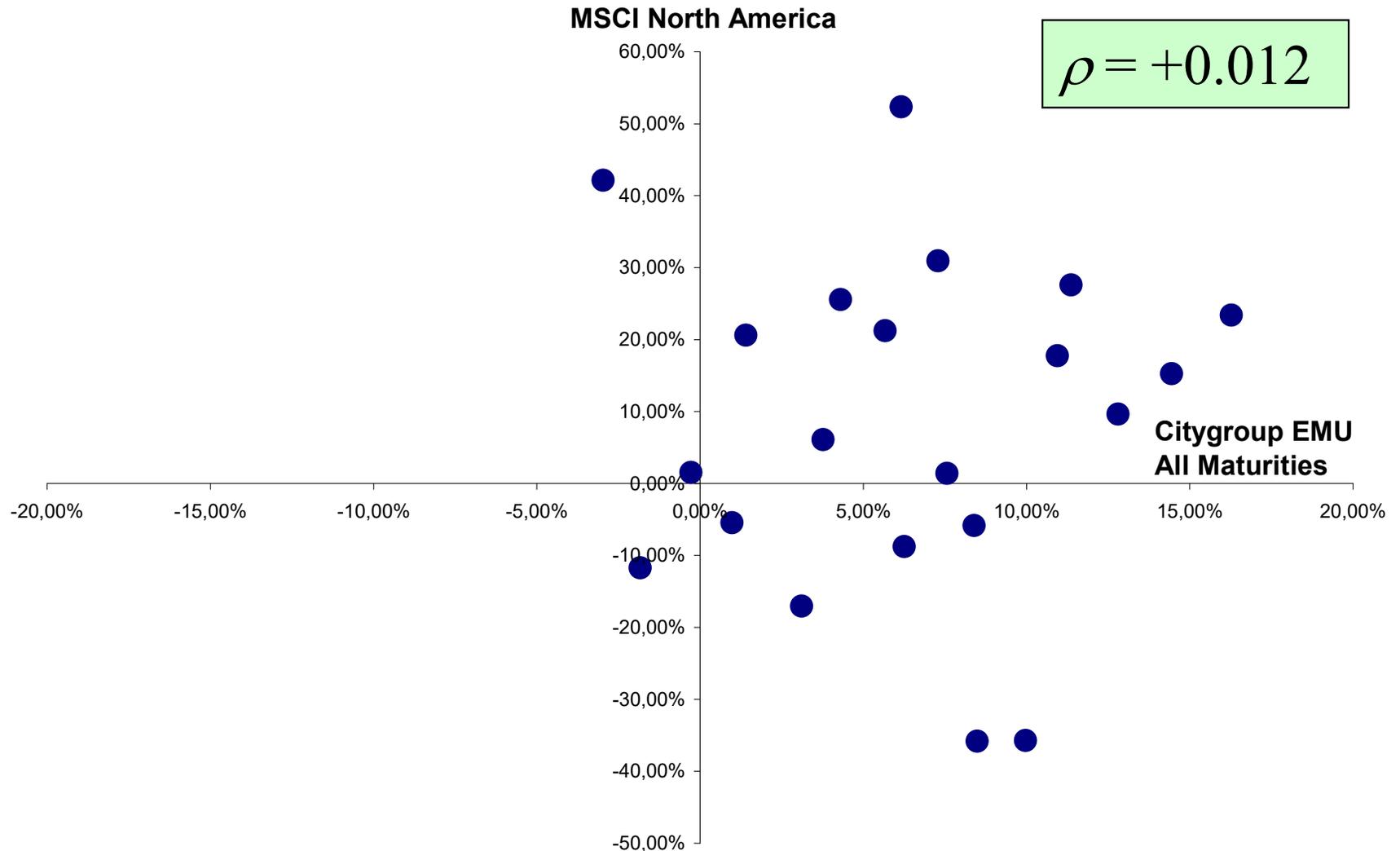
## Case 1: Positive correlation



Strong tendency to move in the same direction (20 times on 21)

# Correlation: the scatter graph

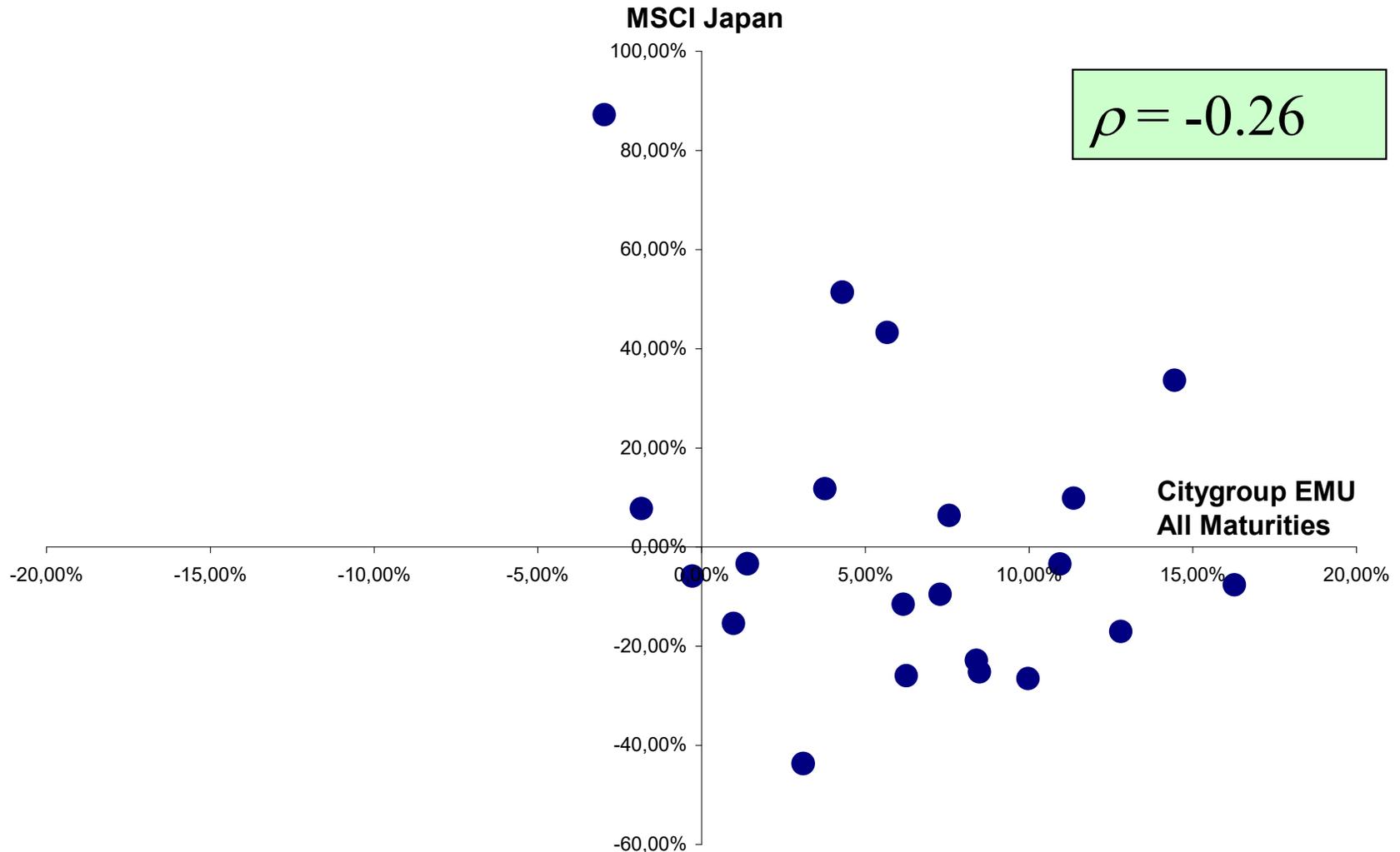
## Case 2: Zero correlation



No tendency (12 times in the same direction - 9 times in opposite direction)

# Correlation: the scatter graph

## Case 3: Negative correlation



(7 times in the same direction - 14 times in opposite direction)

# Correlation Matrix

➤ The Correlation Matrix shows the correlations between all the couples of markets:

(1988-2008)

<i>Correlations</i>	JPM Euro 3 months	Citygroup EMU All Maturities	JPM Global	MSCI Europe	MSCI North America	MSCI Japan	MSCI Pacific ex Japan	MSCI Emerging Markets
JPM Euro 3 months	1							
Citygroup EMU All Maturities	0,30	1						
JPM Global	0,27	0,58	1					
MSCI Europe	-0,09	-0,07	0,31	1				
MSCI North America	0,02	0,01	0,44	0,92	1			
MSCI Japan	-0,21	-0,26	0,24	0,63	0,57	1		
MSCI Pacific ex Japan	0,05	0,01	0,31	0,70	0,55	0,75	1	
MSCI Emerging Markets	0,10	-0,20	0,25	0,68	0,60	0,78	0,91	1

# Correlation Matrix with Excel

The image illustrates the steps to create a correlation matrix in Excel. It shows the 'Analisi dati' (Data Analysis) dialog box with 'Correlazione' (Correlation) selected. Below it, the 'Correlazione' (Correlation) dialog box is shown with the 'Intervallo di input' (Input range) set to '\$B\$1:\$L\$121'. A callout box points to this field with the text 'Insert here the return series of all the markets'. The 'Dati raggruppati per' (Grouped by) section has 'Colonne' (Columns) selected, and 'Etichette nella prima riga' (Labels in the first row) is checked. The 'Opzioni di output' (Output options) section has 'Nuovo foglio di lavoro' (New worksheet) selected.

**Analisi dati**

Strumenti di analisi

- Analisi varianza: ad un fattore
- Analisi varianza: a due fattori con replica
- Analisi varianza: a due fattori senza replica
- Correlazione**
- Covarianza
- Statistica descrittiva
- Smorzamento esponenziale
- Test F a due campioni per varianze
- Analisi di Fourier
- Istogramma

OK  
Annulla  
?

**Correlazione**

Input

Intervallo di input:

Dati raggruppati per:

- Colonne
- Righe

Etichette nella prima riga

Opzioni di output

- Intervallo di output:
- Nuovo foglio di lavoro:
- Nuova cartella di lavoro

OK  
Annulla  
?

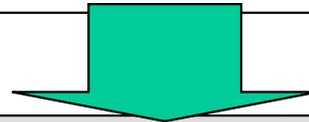
Insert here the return series of all the markets

# The “Gift” of globalization

- As showed by the correlation matrix, globalization has strongly increased the correlation amoung equity markets.

	MSCI Europe	MSCI North America	MSCI Japan	MSCI Pacific ex Japan	MSCI Emerging Markets
MSCI Europe	1				
MSCI North America	0,92	1			
MSCI Japan	0,63	0,57	1		
MSCI Pacific ex Japan	0,70	0,55	0,75	1	
MSCI Emerging Markets	0,68	0,60	0,78	0,91	1

- Today traditional risky assets are not able to produce big benefits of diversification.



This is the main reason why many institutional investors suggest **not to limit** the investment to the classical asset classes (bonds and listed stocks).....

They suggest to invest money also in “**alternative investments**”:

- Hedge funds;
- Commodities;
- Private Equity;
- Real Estate.

**Pay attention: They do not show negative correlation!**

# Standard Deviation of a Portfolio (1/2)

If we know:

- the portfolio weight of each market ( $w_i$ )
- the standard deviations of each market ( $\sigma_i$ )
- the correlations between couples of markets ( $\rho_{i,j}$ )

The estimation of the Portfolio standard deviation is the following:

$$\sigma_{Port} = \sqrt{\sum_{i=1}^k \sum_{j=1}^k w_i \cdot w_j \cdot \sigma_i \cdot \sigma_j \cdot \rho_{i,j}} = \sqrt{\sum_{i=1}^k \sum_{j=1}^k w_i \cdot w_j \cdot \sigma_{i,j}}$$


A “two markets” portfolio:

$$\sigma_{port} = \sqrt{(w_1 \cdot \sigma_1)^2 + (w_2 \cdot \sigma_2)^2 + 2 \cdot w_1 \cdot w_2 \cdot \sigma_1 \cdot \sigma_2 \cdot \rho_{12}}$$

# Standard Deviation of a Portfolio (2/2)

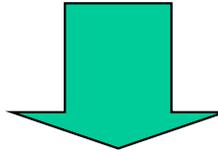
Using Matrices:

$$\sigma_{Port} = \sqrt{\begin{bmatrix} w_1\sigma_1 & w_2\sigma_2 & \cdots & \cdots & w_i\sigma_i & \cdots & w_k\sigma_k \end{bmatrix} \times \begin{bmatrix} \rho_{1,1} & \rho_{1,2} & \rho_{1,3} & \cdots & \rho_{1,j} & \cdots & \rho_{1,k} \\ \rho_{2,1} & \rho_{2,2} & \rho_{2,3} & \cdots & \rho_{2,j} & \cdots & \rho_{2,k} \\ \cdots & \cdots & \cdots & \cdots & \cdots & \cdots & \cdots \\ \cdots & \cdots & \cdots & \cdots & \cdots & \cdots & \cdots \\ \rho_{i,1} & \rho_{i,2} & \rho_{i,3} & \cdots & \rho_{i,j} & \cdots & \rho_{i,k} \\ \cdots & \cdots & \cdots & \cdots & \cdots & \cdots & \cdots \\ \rho_{k,1} & \rho_{k,2} & \rho_{k,3} & \cdots & \rho_{k,j} & \cdots & \rho_{k,k} \end{bmatrix} \times \begin{bmatrix} w_1\sigma_1 \\ w_2\sigma_2 \\ \vdots \\ \vdots \\ w_i\sigma_i \\ \vdots \\ w_k\sigma_k \end{bmatrix}}$$

**Corr**

# Standard deviation of a Portfolio: Numerical example

	JPM Euro 3 months	Citygroup EMU All Maturities	JPM Global	MSCI Europe	MSCI North America	MSCI Japan	MSCI Pacific ex Japan	MSCI Emerging Markets
$\sigma$ of Annual Returns	2,88%	5,11%	8,55%	22,01%	22,53%	29,95%	31,29%	37,93%
Weights	5,00%	40,00%	5,00%	17,00%	23,00%	3,00%	2,00%	5,00%

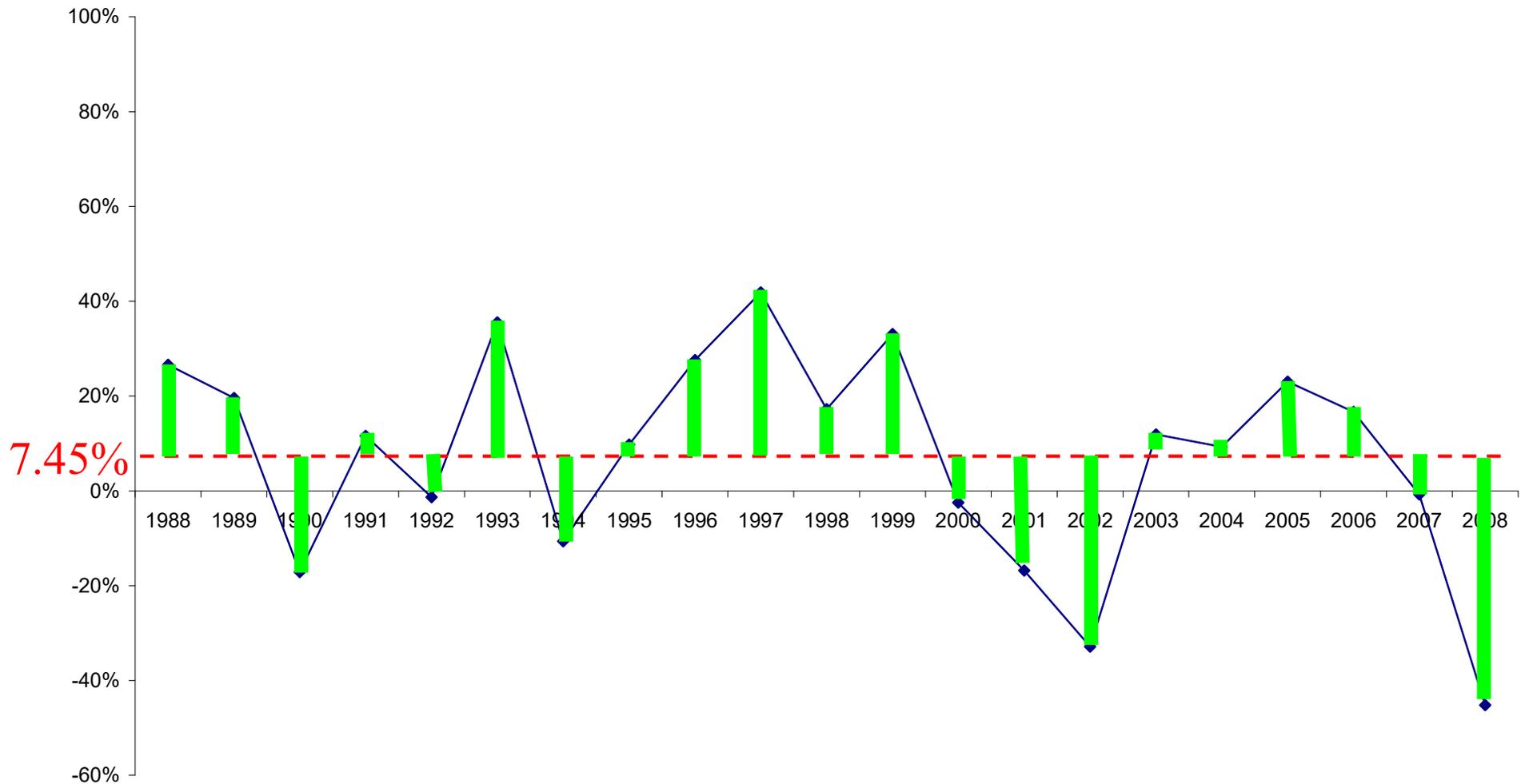


$$\sigma_{Port} = \sqrt{\sum_{i=1}^k \sum_{j=1}^k w_i \cdot w_j \cdot \sigma_i \cdot \sigma_j \cdot \rho_{i,j}} = 11.44\%$$

~~$$\sigma_{Port} = \sum_{i=1}^k w_i \cdot \sigma_i = 14.96\%$$~~

# Standard deviation: is it a good measure of risk?

Annual Return of MSCI Europe (European Equity Market) [1988-2008]



Risk means “bad returns”, so we should focus only on volatility that has negative consequences.

# Semi-standard deviation (semi- $\sigma$ )

This statistical indicator is perfect when you want to measure the downside risk of a market

But this measure is rarely used.

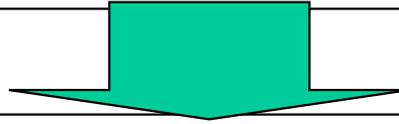
Why?

- It is difficult to measure the semi- $\sigma$  of a portfolio
- It is difficult to build a model of portfolio optimization in which the risk is measured using semi- $\sigma$

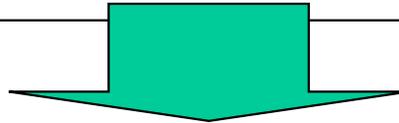
Refusing the semi- $\sigma$  is a convenient solution!

# From volatility to *potential loss*

Is it easy to interpret a measure of volatility?



- Financial experience suggests that investors are not able to interpret the meaning of standard deviation.
- For investors risk means losses, not volatility.....
- so it can be useful to “capture” risk estimating the portfolio potential loss.



Value at Risk (VaR)

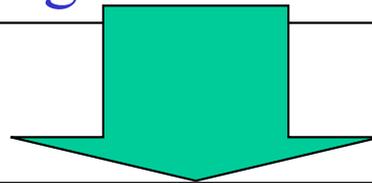
# Introduction to VaR models

*An investor want to invest money in the European Equity Market. His holding period is 1 year.*

*He wants to know which is the risk of this equity market.*

*The standard deviation of annual returns is 22.01%.*

Therefore, the annual return is likely to have an average deviation from the average return of 22.01%.



This statistical indicator is not able to tell him which is the risk to incur in heavy and exceptional losses (year:2008)



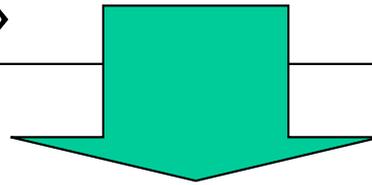
If he want to explore the “darkest side” of the risk, he need a VaR methodology.

## VaR models: the aim

VaR models are able to estimate the potential losses.

For example, thanks to them we can say:

*«Do you want to invest 100,000€ on European Equity Market?  
Well, you have to know that in case of terrible financial events  
you can lose 35,000€!»*



A capital loss of 35% is very easy to understand!

## VaR models: definition

Given a **time horizon (=1 year)**, Value at Risk is the **potential loss (= -35%)** where the **confidence level (=98%)** means that the probability of higher losses is “1-conf.level” (=2%).

### Key elements:

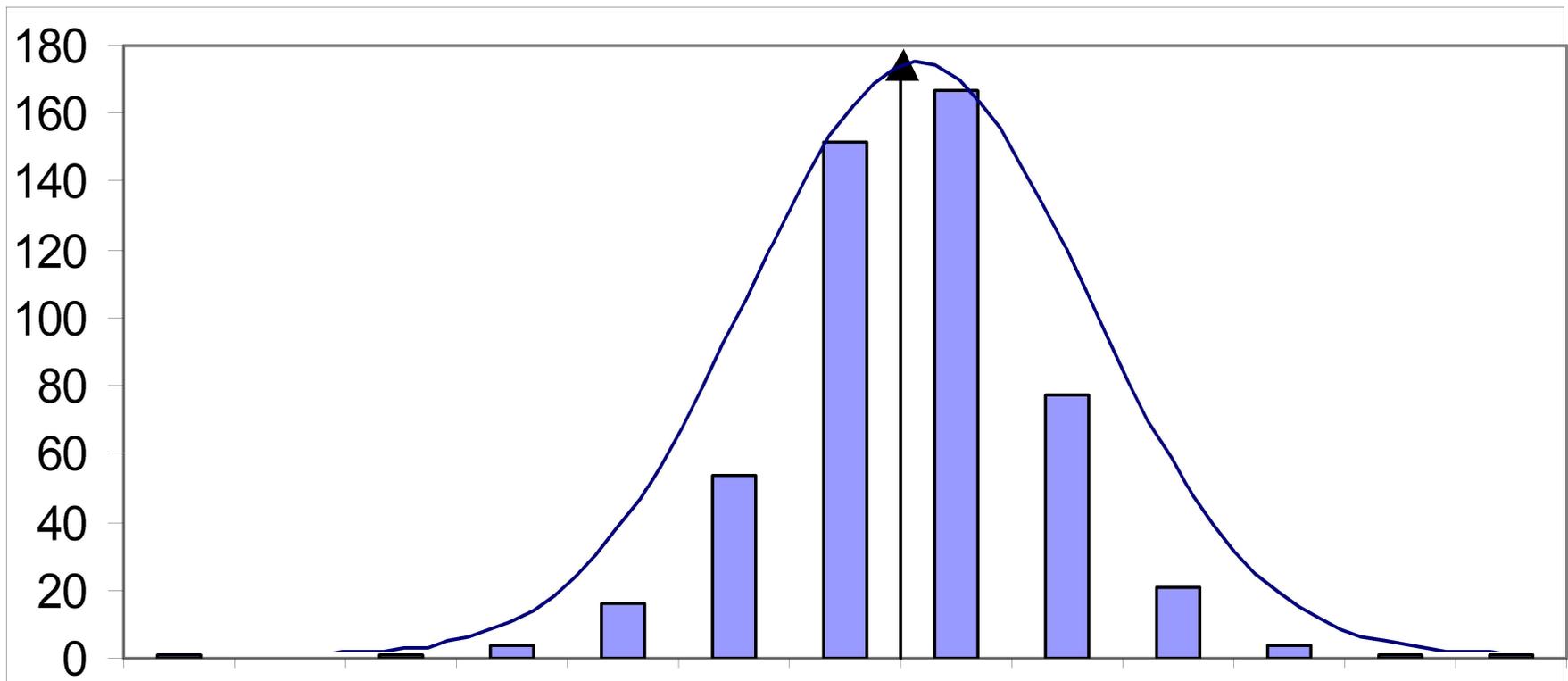
1. Time horizon;
2. Potential loss (not maximum loss);
3. Confidence level (1-c.l. is the prob. of higher losses).

# VaR models: Calculation (1/2)

We analyse a parametrical methodology named “*variance-covariance*”.

The statistical assumption of this method is the following:

*“Returns are normally distributed”*



# VaR models: Calculation (2/2)

Given this assumption, VaR is estimated as follows:

$$VaR = \bar{R} - k \times \sigma$$

AVERAGE RETURN

STANDARD DEVIATION

**THE “K” VALUE IS RELATED TO THE CONFIDENCE LEVEL WE CHOOSE:**

- IF c.l = 95% → k = 1.65
- IF c.l = 98% → k = 2.05
- IF c.l = 99% → k = 2.33

Statistical Tables

# VaR models: Example

1 year VaR of European Equity Market

$$\bar{R}_{Equ.Europe} = 7.45\%$$

$$\sigma_{Equ.Europe} = 22.01\%$$

$$l.c. = 98\% \Rightarrow k = 2.05$$

$$VaR = \bar{R} - k \times \sigma = 7.45\% - 2.05 \times 22.01\% = -37.7\%$$

“Given a 1 year time horizon, the potential loss is -37.7%.  
The probability of higher losses is 2%”.

*If an investor wants to invest money on European Equity  
Market he has to tolerate a -37.7% annual loss.*

In the following analysis we make the hypothesis to be an Asset Management Committee engaged in Strategic Asset Allocation process.

The purpose is to identify a good model to build a SAA.

# Agenda

## 3. Strategic Asset Allocation: Naïve Portfolio Formation Rule

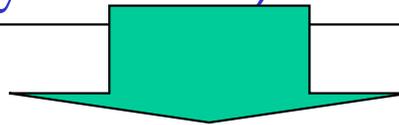
This first solution follows a qualitative approach

Naïve portfolios refuses mathematical solutions.

# Naïve Portfolio Formation Rule: A primitive approach

Naïve strategies:

- are mathematics/statistics free;
- don't need optimization models;
- don't need numerical-quantitative estimations; estimations can be qualitative-judgemental (European Equity Market will beat the Japanese Equity Market).



Naïve strategies:

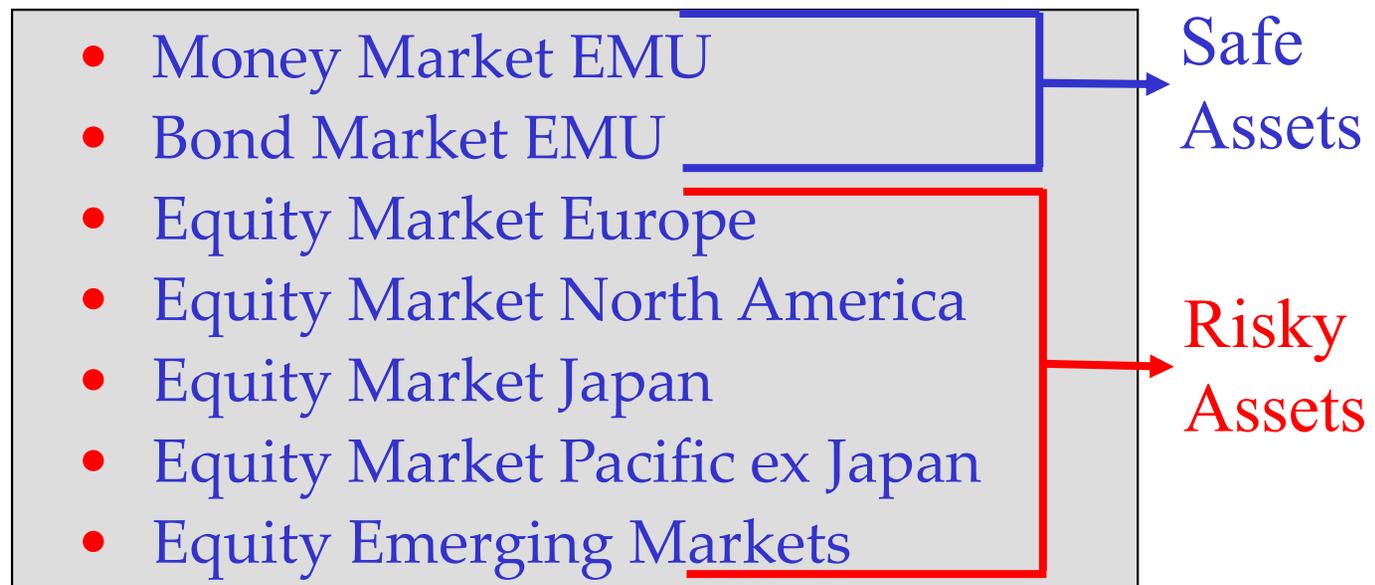
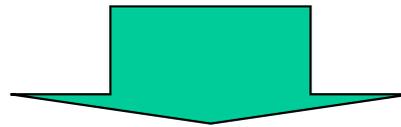
- are easy to put into practice;
- can generate good solution, never optimal ones;
- generate portfolios that are usually diversified and reasonable.

# Naïve Portfolio Formation Rule: Example (1/7)

We need to identify the SAA of a pension fund.

As members of the Asset Allocation Committee, we need to manage a procedure able to identify the portfolio of asset classes.

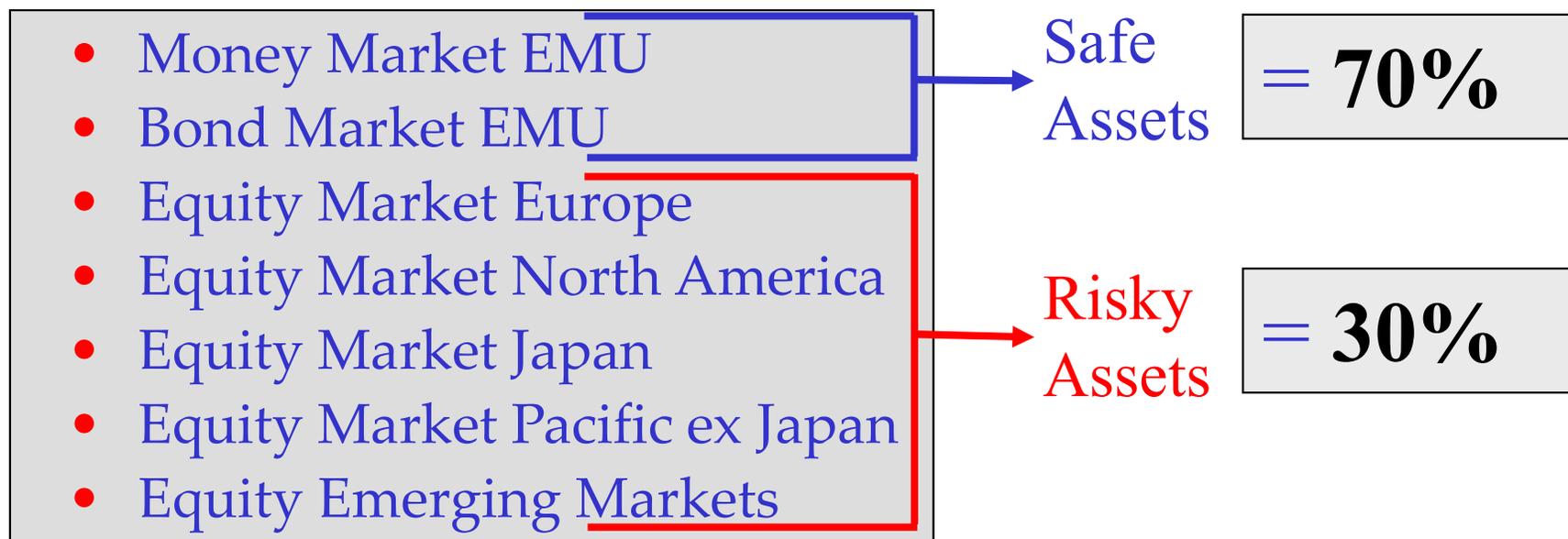
1. First, we select the asset classes where “putting money”:



# Naïve Portfolio Formation Rule: Example (2/7)

2. We identify the risk profile of the portfolio selected:

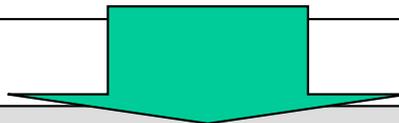
Given the expected risk tolerance of investors that will put money in the pension fund we make the following decision:



Risk profile can be “captured” with quantitative methodology (see Cucurachi). But Naïve portfolios refuses quantitative approach.

# Naïve Portfolio Formation Rule: Example (3/7)

3. We select the weights inside the “Safe-Assets Group”



In the time horizon of the investment we forecast a **general increase of interest rates** in EMU area.

So, in order to maximise the expected return, we need to **reduce the maturity** of the Safe-Assets Group.

Safe-Assets	Weights
Money Market EMU	55%
Bond Market EMU	15%
<b>Sum</b>	<b>70%</b>

# Naïve Portfolio Formation Rule:

## Example (4/7)

4. We select the weights inside the “Risky-Assets Group”

If we don't have views about the future trend of Stock Markets, we should replicate the composition of the World Market. This solution is named Market Neutral: it is “loyal” to the Market.

Risky-Assets	Equity Markets Capitalisation		Portfolio Weights
Equity Market Europe	31%	➡	$(31\% \times 30\%) = 9.3\%$
Equity Market North America	48%	➡	$(48\% \times 30\%) = 14.4\%$
Equity Market Japan	10%	➡	$(10\% \times 30\%) = 3.0\%$
Equity Market Pacific ex Japan	4%	➡	$(4\% \times 30\%) = 1.2\%$
Equity Emerging Markets	7%	➡	$(7\% \times 30\%) = 2.1\%$
<b>Sum</b>	<b>100%</b>		<b>30%</b>

No value-added for investors: we just replicate the market!

# Naïve Portfolio Formation Rule:

## Example (5/7)

4. We select the weights inside the “Risky-Assets Group”

But we try to beat the market, so we “depict” the future:

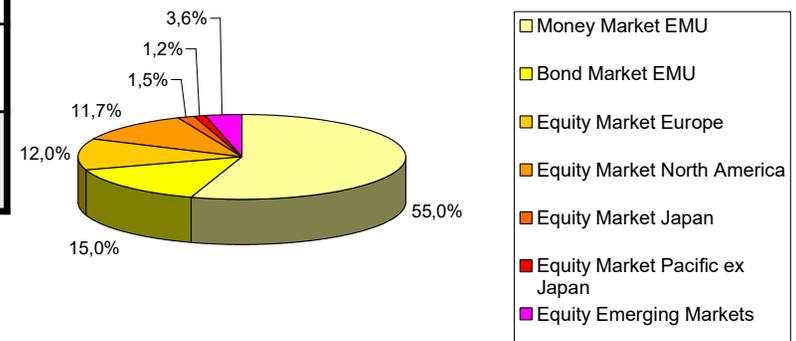
- Europe will over perform North America
- EM will over perform Japan
- Pacific ex Japan NEUTRAL

Risky-Assets	Equity Markets Capitalisation	New Group Weights	Portfolio Weights
Europe	abandoned	31% →	40% → $(40% \times 30%) = 12.0%$
North America		48% →	39% → $(39% \times 30%) = 11.7%$
Japan		10% →	5% → $(5% \times 30%) = 1.5%$
Pacific ex Japan		4% →	4% → $(4% \times 30%) = 1.2%$
Em. Mkts		7% →	12% → $(12% \times 30%) = 3.6%$
<b>Sum</b>	<b>100%</b>		<b>30%</b>

# Naïve Portfolio Formation Rule: Example (6/7)

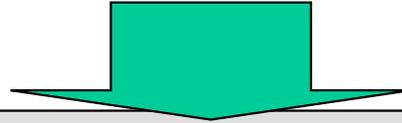
## The final portfolio

Assets	Portfolio Weights
😊 Money Market EMU	55.0%
😞 Bond Market EMU	15.0%
😊 Equity Market Europe	12.0%
😞 Equity Market North America	11.7%
😞 Equity Market Japan	1.5%
😞 Equity Market Pacific ex Japan	1.2%
😊 Equity Emerging Markets	3.6%
<b>Sum</b>	<b>100.0%</b>



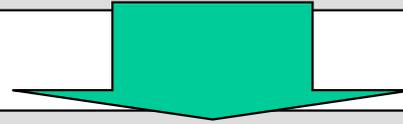
# Naïve Portfolio Formation Rule:

## Example (7/7)



The final naïve portfolio:

- is diversified;
- has a reasonable composition.



.....but at its best:

- it is a good solution....
- it is not the optimal one.

*If you want more, you need*

***MODERN PORTFOLIO THEORY (MPT)***

# Agenda

## 4. Strategic Asset Allocation: A Quantitative Approach

# Quantitative Approach: The Markowitz Model

Harry Markowitz's *Portfolio selection* is the "father" of portfolio optimization.....

.....and his model (even if it is 50 years old) is widely used in portfolio construction.

No doubt, there are other mathematical approaches. But no one has the Markowitz's model aptitude to be:

- rigorous from a mathematical point of view;
- easy to be implemented.

# The Markowitz Model: The hypotheses

Given a unique time horizon.....

.....investors want to maximise the expected return (*“They love returns”*).

Investors are risk adverse (*“They hate risk”*)

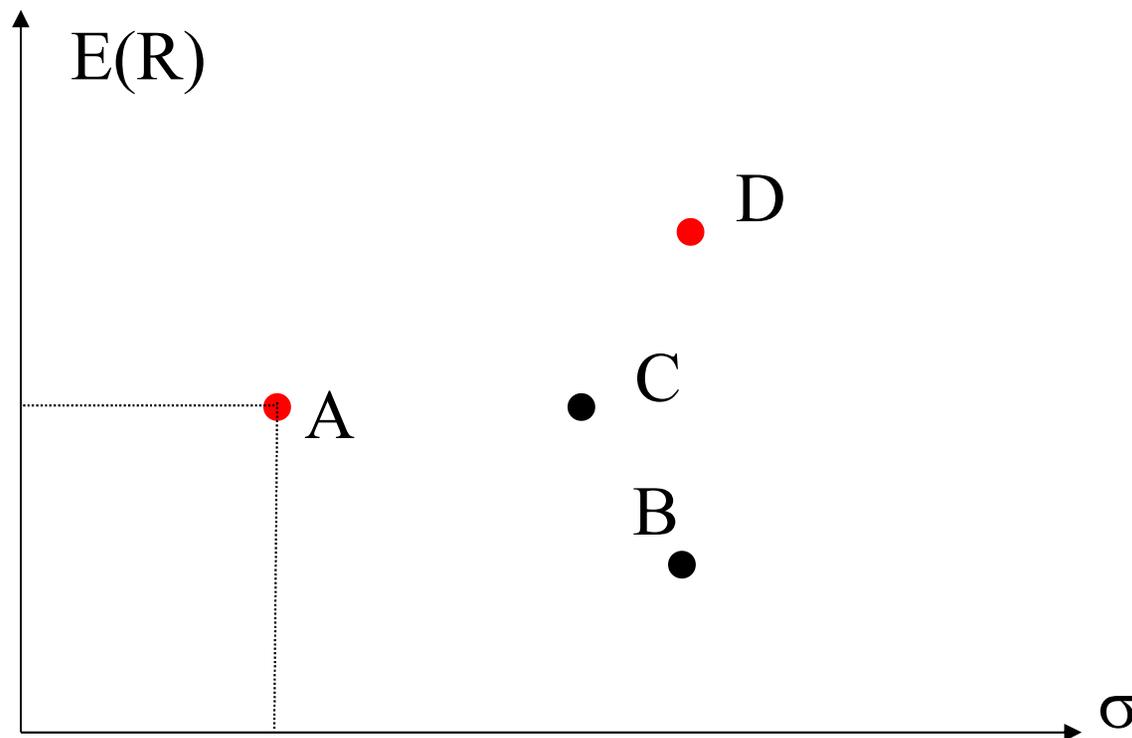
The statistical parameter used to measure risk is the standard deviation.

«One of the measures considered, the semi-standard deviation, produces efficient portfolios some what preferable to those of the standard deviation. Those produced by the standard deviation are satisfactory, however, and the standard deviation itself is **easier to use, more familiar to many, and perhaps easier to interpret** than the semi-standard deviation». (Markowitz, 1959).

# The “Expected Return – Standard Deviation” Principle

Risk is *bad variable*:

Therefore, investors are willing to increase risk only if higher risk produces higher return.



Solutions B & C are **inefficient**

Solution D is **efficient**: it is an optimal solution for “*high risk tolerance*” investors



# The Markowitz Model: A few remarks

1. 5 stages to be performed.
2. The process is time-expensive: you need to estimate many parameters.
3. For example: with 8 asset classes selected, it is necessary to estimate:
  - 8 expected return;
  - 8 standard deviation;
  - 28 correlation.
4. Unfortunately asset manager don't like to produce quantitative and numerical estimation ("*European equity market is expected to perform 7.0%*")......they prefer to produce qualitative estimation ("*European equity market will beat North American equity market*").
5. No way, if we want to use the Markowitz model, quantitative estimations are required.

# Stage 1: Selection of Asset Classes (1/2)

## Asset Classes

MKT1

MKT2

MKT3

MKT4

MKT5

MKT6

MKT7

MKT8

MKT9

MKT10

MKT11

MKT12

- From a theoretical point of view, we shouldn't narrow the investment opportunities.
- So we could select dozens of asset classes (may be hundreds).
- But this theoretical position cannot be performed because of many practical problems:
  - Increase of parameters to be estimated;
  - Reduction of Asset Under Management (AUM) for every asset class → **Increase of management fees**

# Stage 1: Selection of Asset Classes (2/2)

## Asset Classes

MKT1

MKT2

MKT3

MKT4

MKT5

MKT6

MKT7

MKT8

MKT9

MKT10

MKT11

MKT12

Asset Managers usually select not more than 10-12 asset classes

- “Marginal players” (ex: Japanese Money Market ) are ignored;
- Similar (highly positive correlated) market are aggregated.

## Stage 2: Expected Returns [E(R)] (1/2)

### Exp. Returns

MKT1

MKT2

MKT3

MKT4

MKT5

MKT6

MKT7

MKT8

MKT9

MKT10

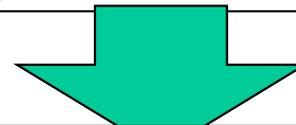
MKT11

MKT12

### MY SUGGESTION:

Expected Returns shouldn't be the "historical average returns".

- Empirical studies say that the "*Rear-view mirror*" strategy doesn't work.
- Future is different from the past (returns probability distribution are not stationary).



### A wrong belief:

In the financial environment is widely spread the idea that Harry Markowitz suggested to use historical estimators. This is wrong:

*"The procedures, I believe, should combine statistical techniques and the judgment of practical men. [...] One suggestion is to use the observed parameters for some period of the past. I believe that better methods, which take into account more information, can be found» (Markowitz, 1952)*

## Stage 2: Expected Returns [E(R)] (2/2)

### Exp. Returns

MKT1

MKT2

MKT3

MKT4

MKT5

MKT6

MKT7

MKT8

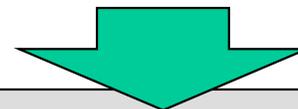
MKT9

MKT10

MKT11

MKT12

- Empirical studies suggest that historical average return **are not good** predictor of the future return.
- Empirical studies suggest that estimation error in E(R) are “deadly”.
- Asset Managers must forecast the future, not trust the predictive power of the past.
- Expected returns must be forward looking, not backward looking.



Statistical techniques can be useful:

- **Macroeconomic models:** based on the connection between future return and macroeconomics factor;
- **Autoregressive models:** based on the study of trend of the historical series of returns.

## Stage 3: Standard Deviations ( $\sigma$ ) (1/2)

$\sigma$

MKT1

MKT2

MKT3

MKT4

MKT5

MKT6

MKT7

MKT8

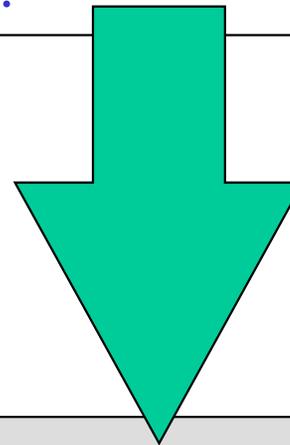
MKT9

MKT10

MKT11

MKT12

- Empirical studies suggest that historical standard deviations are **good** predictor of the future standard deviation.
- Estimation error in standard deviation are not “deadly”.



### MY SUGGESTION:

You can apply the “classical rule”, using the “*observed  $\sigma$  for some period of the past*”.

We save time and focus our efforts on Expected Return prediction.

## Stage 3: Standard Deviations ( $\sigma$ ) (2/2)

$\sigma$

MKT1

MKT2

MKT3

MKT4

MKT5

MKT6

MKT7

MKT8

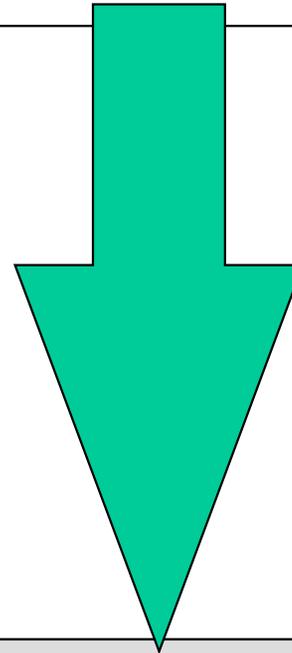
MKT9

MKT10

MKT11

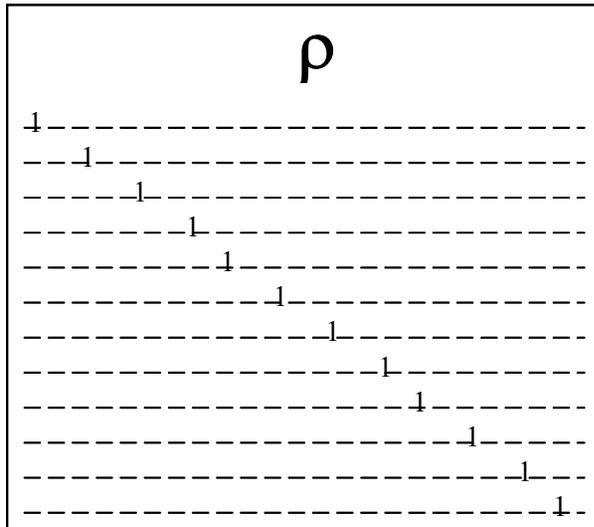
MKT12

➤ If you want, you can use more sophisticated technical models:

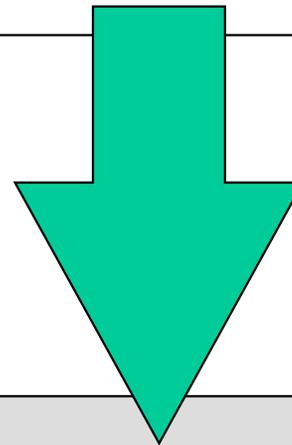


- Implied volatility;
- Econometric models (ARCH, GARCH).

## Stage 4: Correlations ( $\rho$ ) (1/2)



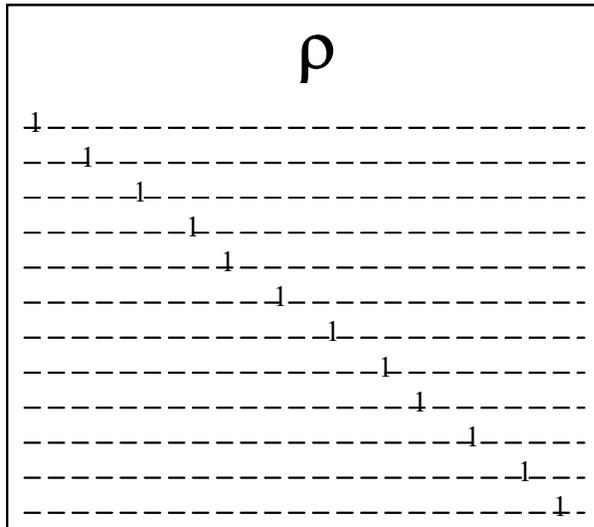
- Empirical studies suggest that historical correlations are **good** predictor of the future correlation.
- Estimation error in correlation are not “deadly”.



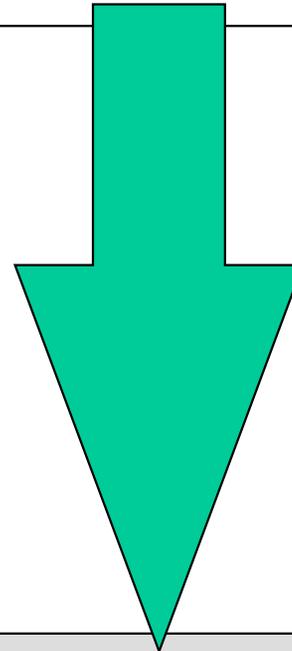
### MY SUGGESTION:

You can apply the “classical rule”, using the “*observed  $\rho$  for some period of the past*”.

## Stage 5: Correlations ( $\rho$ ) (2/2)

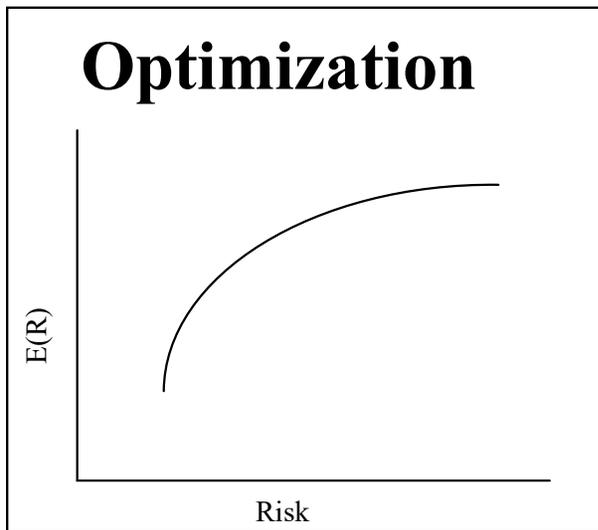


➤ If you want, you can use more sophisticated technical models:

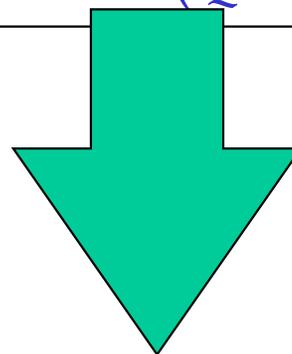


➤ Econometric models (ARCH, GARCH).

# Final Stage: Optimization (1/3)



- If we have: **Asset Classes**,  $E(r)$ ,  $\sigma$  and  $\rho$ ...
- We can optimize (Quadratic Programming).



**Find the weights ( $w_i$ ) able to:**

*Objective function* →

**MIN  $\sigma_{\text{Portfoglio}}$**

*Constraints:*

*1<sup>st</sup> constraint:* →

**Exp. Return =  $E(R)^*$**

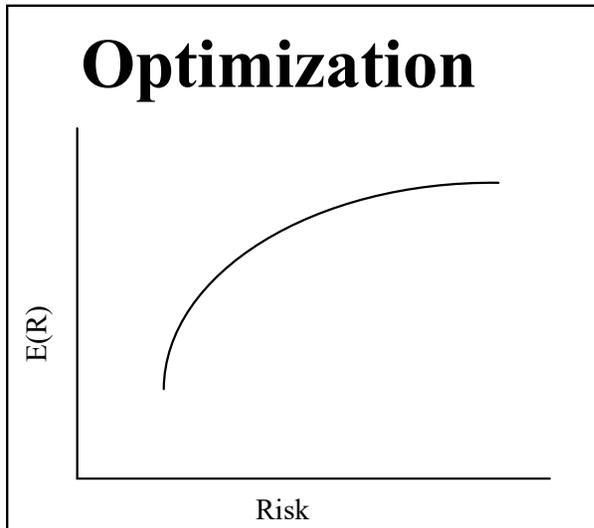
*2<sup>nd</sup> constraint:* →

**$w_1 + \dots + w_i + \dots + w_n = 1$**

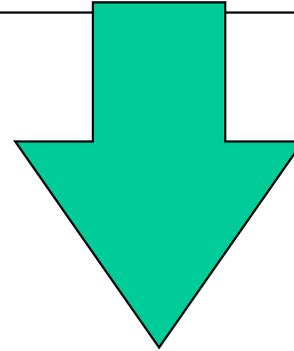
*3<sup>rd</sup> constraint:* →

**$w_i \geq 0$**

# Final Stage: Optimization (2/3)



➤ Mathematical structure of the Markowitz optimization:



$$\underset{W}{Min} \quad \sigma_{Port}^2$$

Constraints :

$$\sum_{i=1}^k w_i E(R_i) = E(R^*)$$

$$\sum_{i=1}^n w_i = 1$$

$$w_i \geq 0 \quad \text{with } i = 1, \dots, k$$

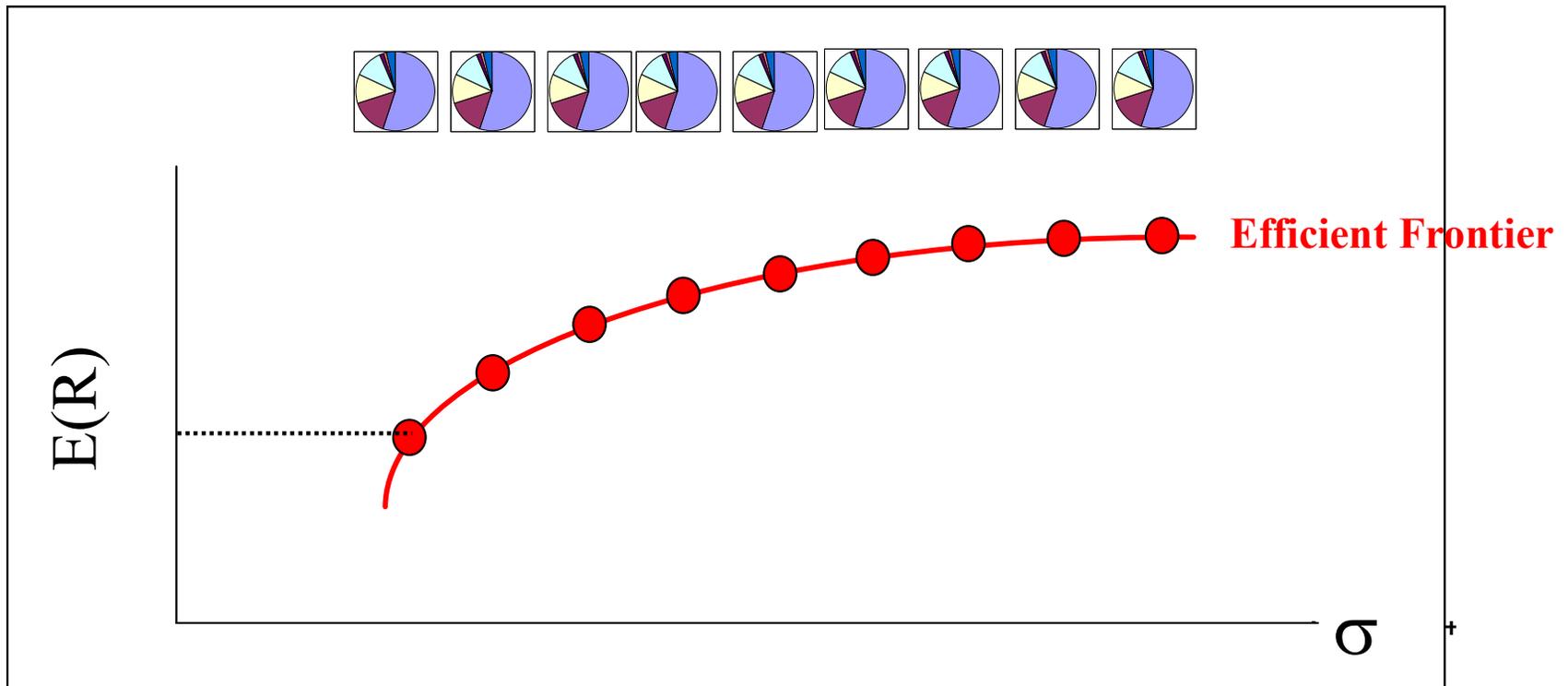
# Final Stage: Optimization (3/3)

$$\text{Min}_w \sigma_{Port}^2$$

Constraints:

$$\sum_{i=1}^k w_i E(R_i) = E(R^*)$$
$$\sum_{i=1}^k w_i = 1$$
$$w_i \geq 0 \text{ with } i = 1, \dots, k$$

- We run this optimization for a targeted expected return  $[E(R)^*]$ .....
- The optimization returns:
  - the portfolio composition.....
  - that is efficient as, given the targeted  $E(R)$ , it is able to minimise the standard deviation
- Running the optimization for different targeted  $E(R)$  we obtain a range of efficient portfolio



# Markowitz Optimization: An application (1/8)

## Asset Classes selected:

Benchmark selezionati

Nome
JPM Euro 3 mesi
JPM EMU Aggregate Tutte le Scadenze
MSCI Europa
MSCI Nord America
MSCI Giappone
MSCI Pacifico ex Giappone
MSCI Emerging Market Free

# Markowitz Optimization: An application (2/8)

Expected Returns estimated:

	Rendimento atteso %
JPM EMU Aggregate Tutte le	3,2
JPM Euro 3 mesi	2,8
MSCI Giappone	4,5
MSCI Pacifico ex Giappone	6
MSCI Nord America	6
MSCI Europa	7
MSCI Emerging Market Free	8

# Markowitz Optimization: An application (3/8)

Standard deviations estimated:

	Rischio atteso %
JPM EMU Aggregate Tutte le	1,8
JPM Euro 3 mesi	4,2
MSCI Giappone	22,8
MSCI Pacifico ex Giappone	23
MSCI Nord America	21
MSCI Europa	20
MSCI Emerging Market Free	29

# Markowitz Optimization:

## An application (4/8)

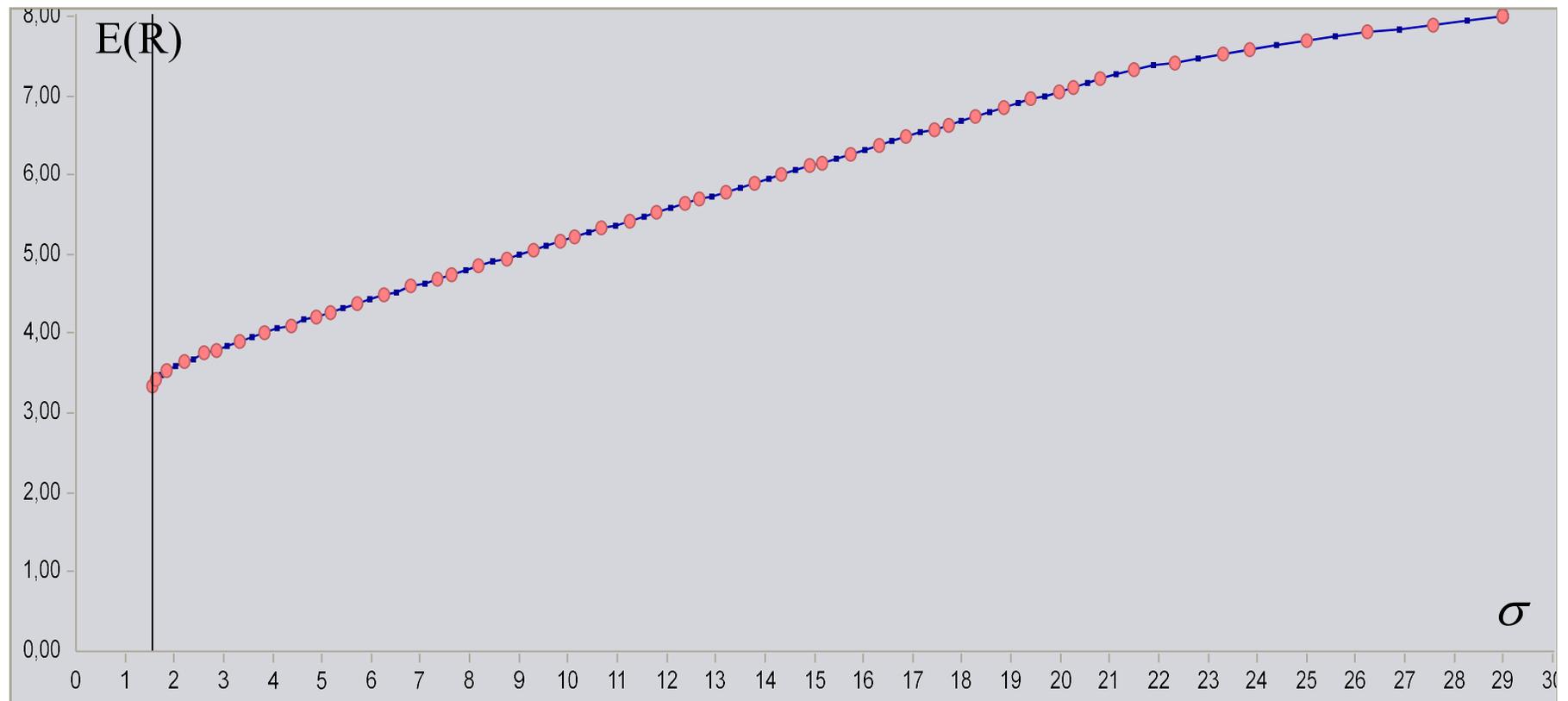
### Correlations estimated:

Matrice delle correlazioni

	JPM EMU Aggregate Tutte le	JPM Euro 3 mesi	MSCI Giappone	MSCI Pacifico ex Giappone	MSCI Nord America	MSCI Europa	MSCI Emerging Market Free
JPM EMU Aggregate	1	Storico	Storico	Storico	Storico	Storico	Storico
JPM Euro 3 mesi	0,27	1	Storico	Storico	Storico	Storico	Storico
MSCI Giappone	-0,26	-0,27	1	Storico	Storico	Storico	Storico
MSCI Pacifico ex	-0,27	-0,15	0,61	1	Storico	Storico	Storico
MSCI Nord America	-0,35	-0,13	0,61	0,74	1	Storico	Storico
MSCI Europa	-0,4	-0,2	0,53	0,72	0,85	1	Storico
MSCI Emerging Market	-0,34	-0,18	0,62	0,87	0,76	0,76	1

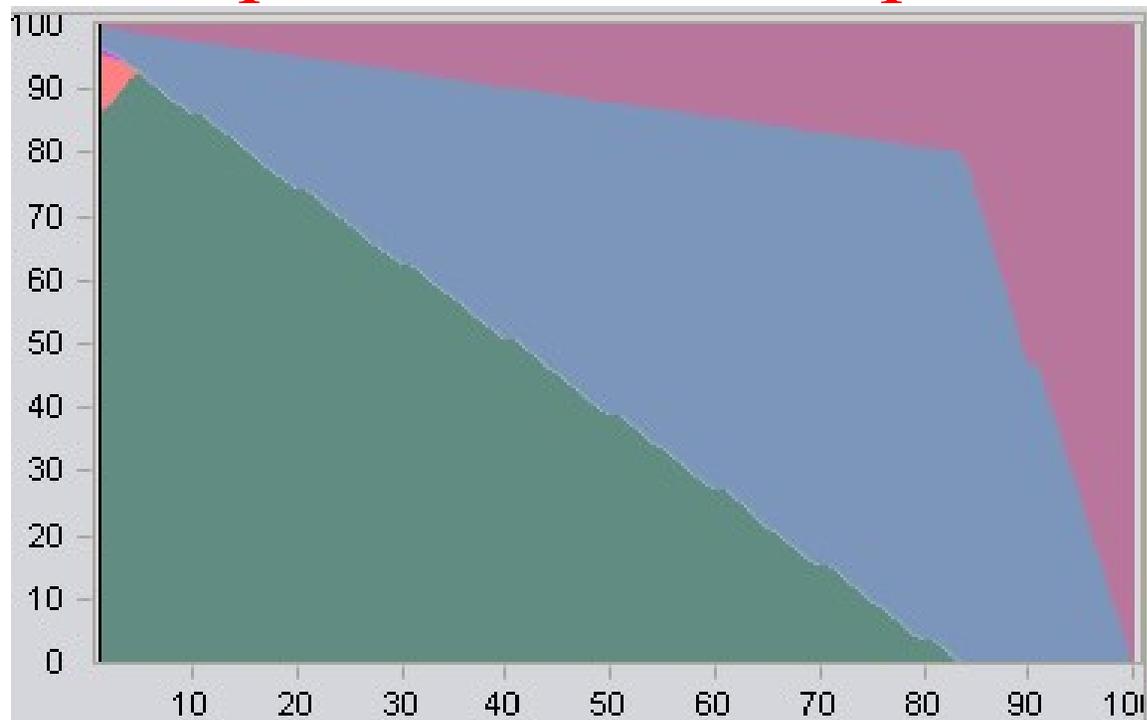
# Markowitz Optimization: An application (5/8)

## Output: Efficient Frontier



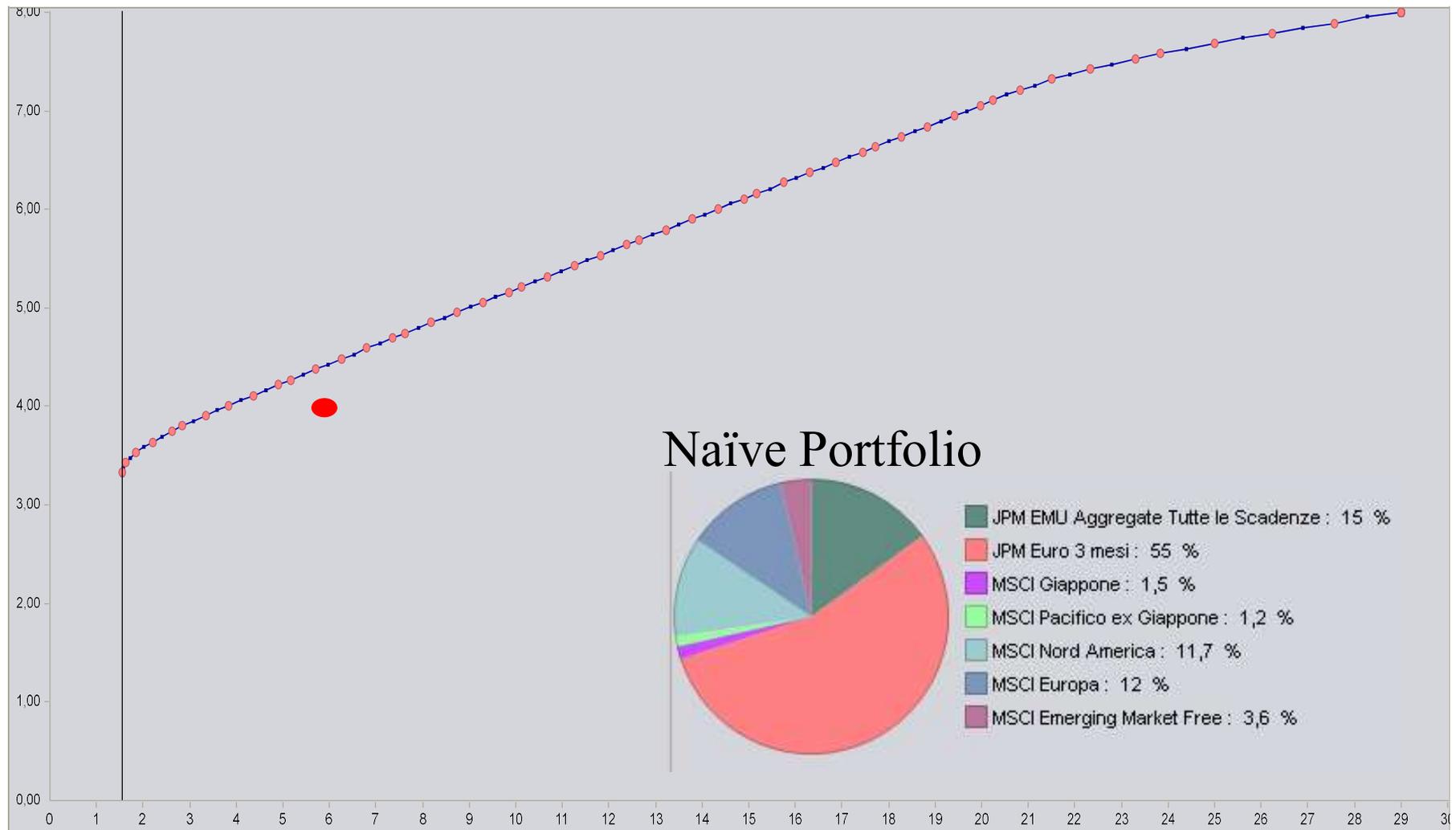
# Markowitz Optimization: An application (6/8)

## Output: Portfolio composition



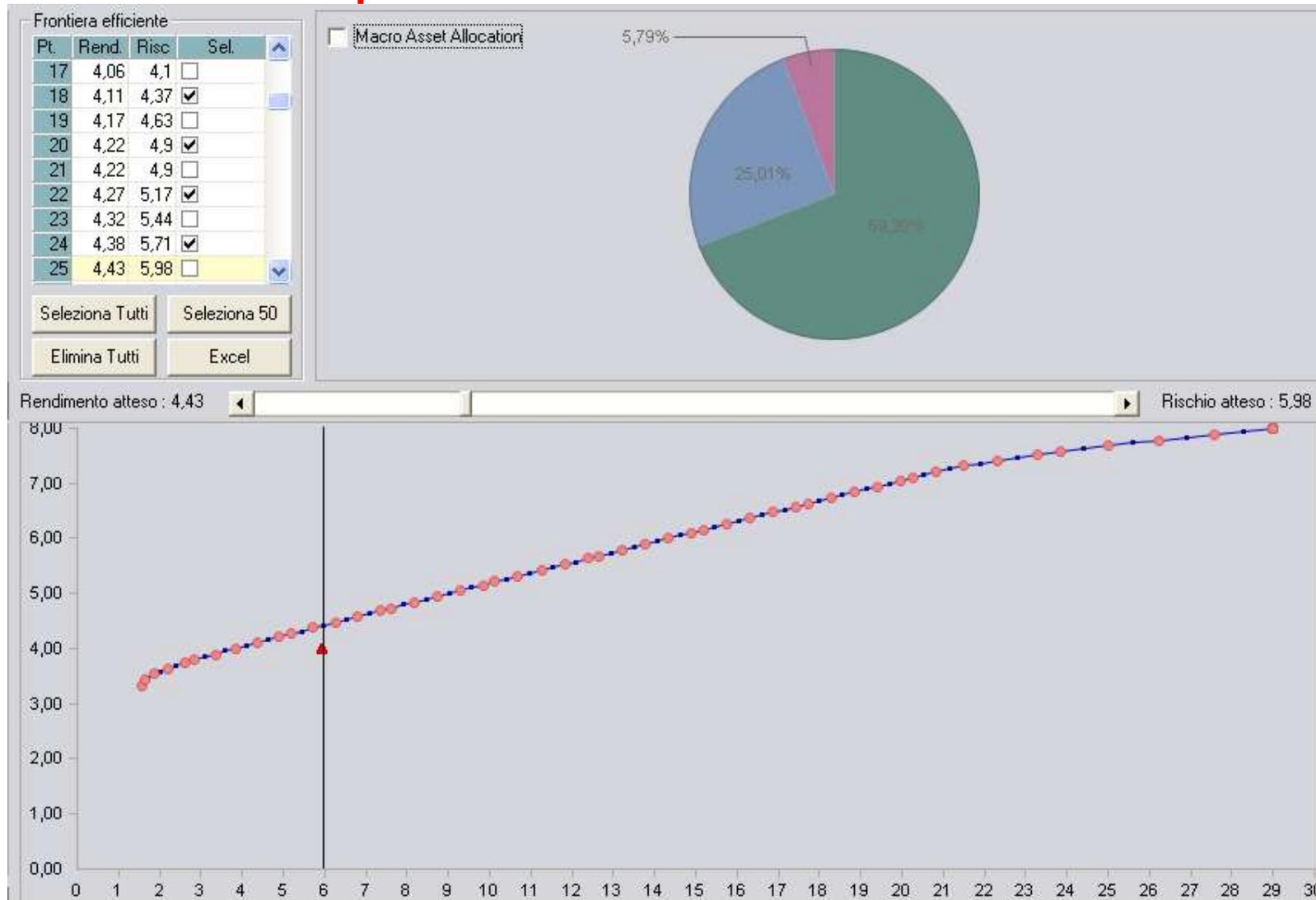
# Markowitz Optimization: An application (7/8)

All the other portfolios are inefficient



# Markowitz Optimization: An application (8/8)

A better portfolio exists:



# Markowitz Optimization: Excel

(1/2)

Markowitz Optimization can be easily processed using Excel

	A	B	C	D	E	F	G	H	I	J	K	L	M	N
1	<b>Asset class</b>	<b>E(R)</b>	<b><math>\sigma</math></b>	<b>Weights</b>		<b>Correl</b>								<b><math>W_i \cdot \sigma_i</math></b>
2	JPM Euro 3 mesi	2,80%	1,80%	0,00%		1,00	0,27	-0,20	-0,13	-0,27	-0,15	-0,18		0,00%
3	JPM EMU Aggregate	3,20%	4,20%	69,80%		0,27	1,00	-0,40	-0,35	-0,26	-0,27	-0,34		2,93%
4	MSCI Europa	7,00%	20,00%	24,95%		-0,20	-0,40	1,00	0,85	0,53	0,72	0,76		4,99%
5	MSCI Nord America	6,00%	21,00%	0,00%		-0,13	-0,35	0,85	1,00	0,61	0,74	0,76		0,00%
6	MSCI Giappone	4,50%	22,80%	0,00%		-0,27	-0,26	0,53	0,61	1,00	0,61	0,62		0,00%
7	MSCI Pacifico ex Giappone	6,00%	23,00%	0,00%		-0,15	-0,27	0,72	0,74	0,61	1,00	0,87		0,00%
8	MSCI Emerging Market Free	8,00%	29,00%	5,24%		-0,18	-0,34	0,76	0,76	0,62	0,87	1,00		1,52%
9	<b>Portafoglio</b>	<b>4,40%</b>	<b>5,71%</b>	<b>100,00%</b>										
10														

`=SUMPRODUCT(B2:B8;D2:D8)`

`=SUM (D2:D8)`

`=SQRT(MMULT(MMULT(TRANSPOSE(N2:N8);F2:L8);(N2:N8)))`

Then click all together:  
Ctrl+Shift+Enter

# Markowitz Optimization: Excel

(2/2)



**Parametri del Risolutore**

Imposta cella obiettivo:  

Uguale a:    Max    Min    Valore di:

Cambiando le celle:

Vincoli:

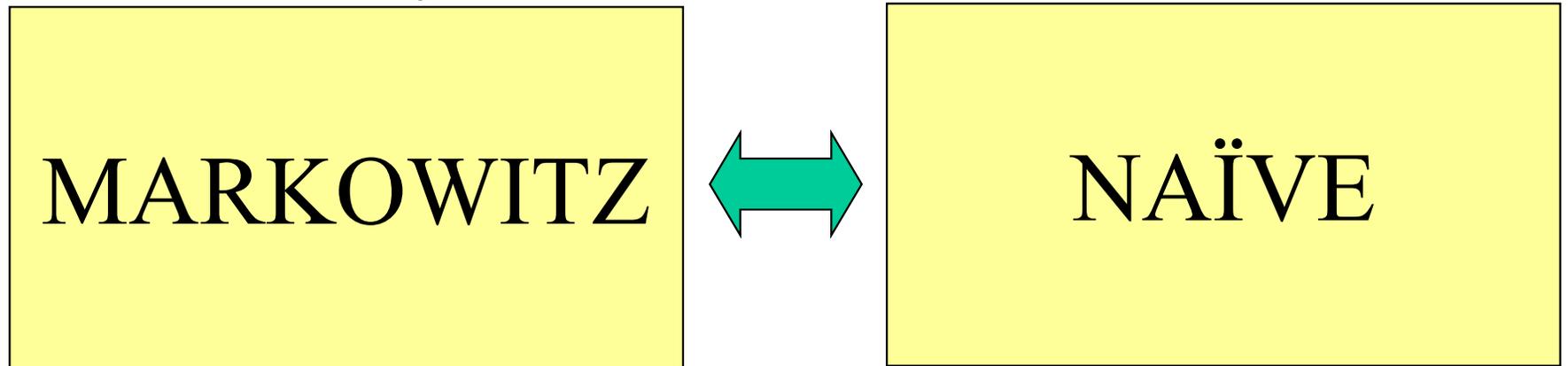
**Constraints**

**Weights**

**Risk**

# Markowitz *versus* Naïve

Which is your choice to build a SAA?



# Agenda

## 5. Naïve *versus* Markowitz

# It glitters but.....

Markowitz optimization seems to be the best solution.

Nevertheless financial literature has showed that this model has some problems:

1. Efficient portfolios are often **unreasonable** (Portfolios highly concentrated and/or big weights to “marginal markets”).
2. Efficient Portfolios are **instable** (small changes in expected returns can strongly affect the portfolio composition).
3. Estimations are supposed to be **perfect** (Asset managers are clairvoyant! Estimation error doesn't exist).
4. Efficient portfolios are **“estimation error maximizers”**

Extra-argument 1:  
Efficient Portfolios are **instable**

# Extra-argument 1: Efficient Portfolios are **instabile** (1/5)

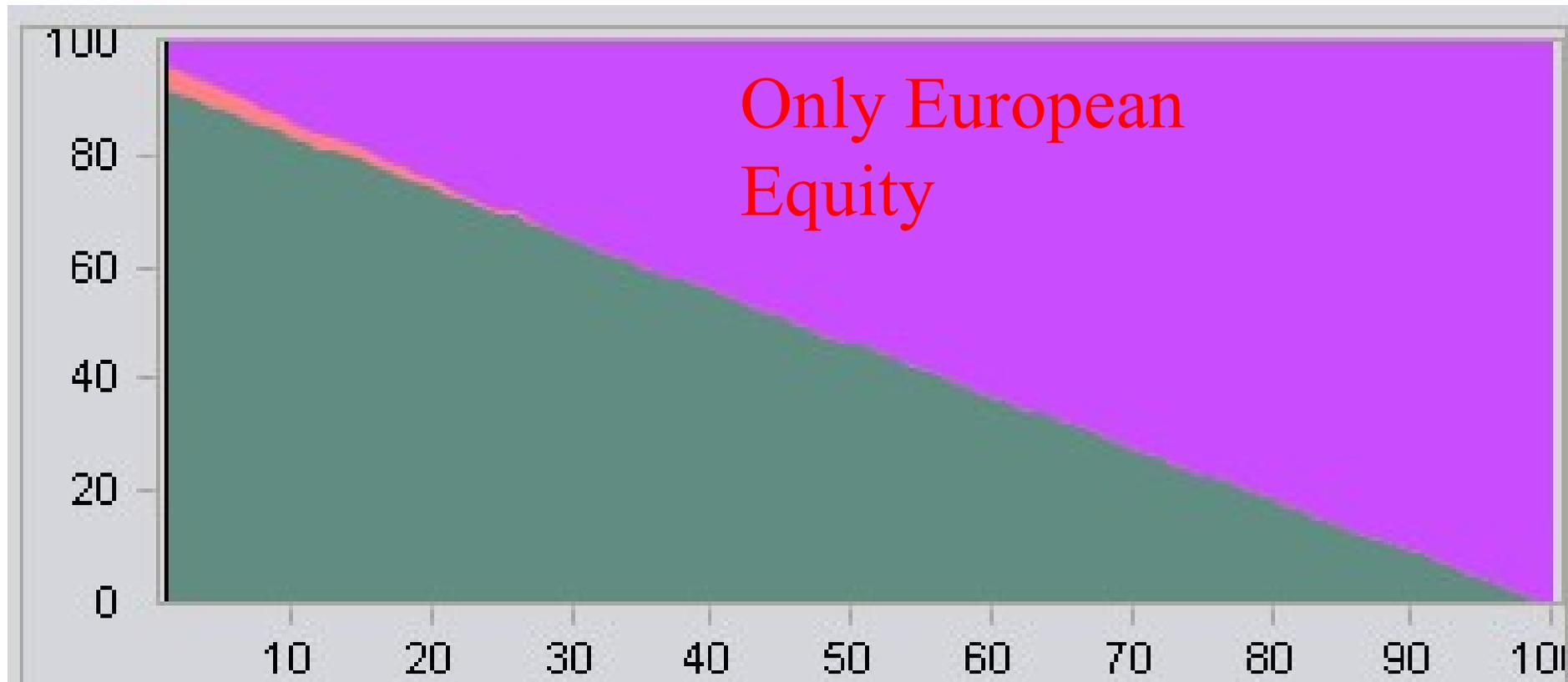
Asset Management Committee “Alfa” has the following estimation

	Rendimento atteso %	Rischio atteso %
JPM EMU Aggregate Tutte le	3	3,5
MSCI Nord America	7	20
MSCI Europa	7,4	20

	JPM EMU Aggregate Tutte le	MSCI Nord America	MSCI Europa
JPM EMU Aggregate	1	Storico	Storico
MSCI Nord America	-0,35	1	Manuale
MSCI Europa	-0,4	0,91	1

# Extra-argument 1: Efficient Portfolios are **instable** (2/5)

.....Optimal Portfolio are the following:



# Extra-argument 1: Efficient Portfolios are **instable** (3/5)

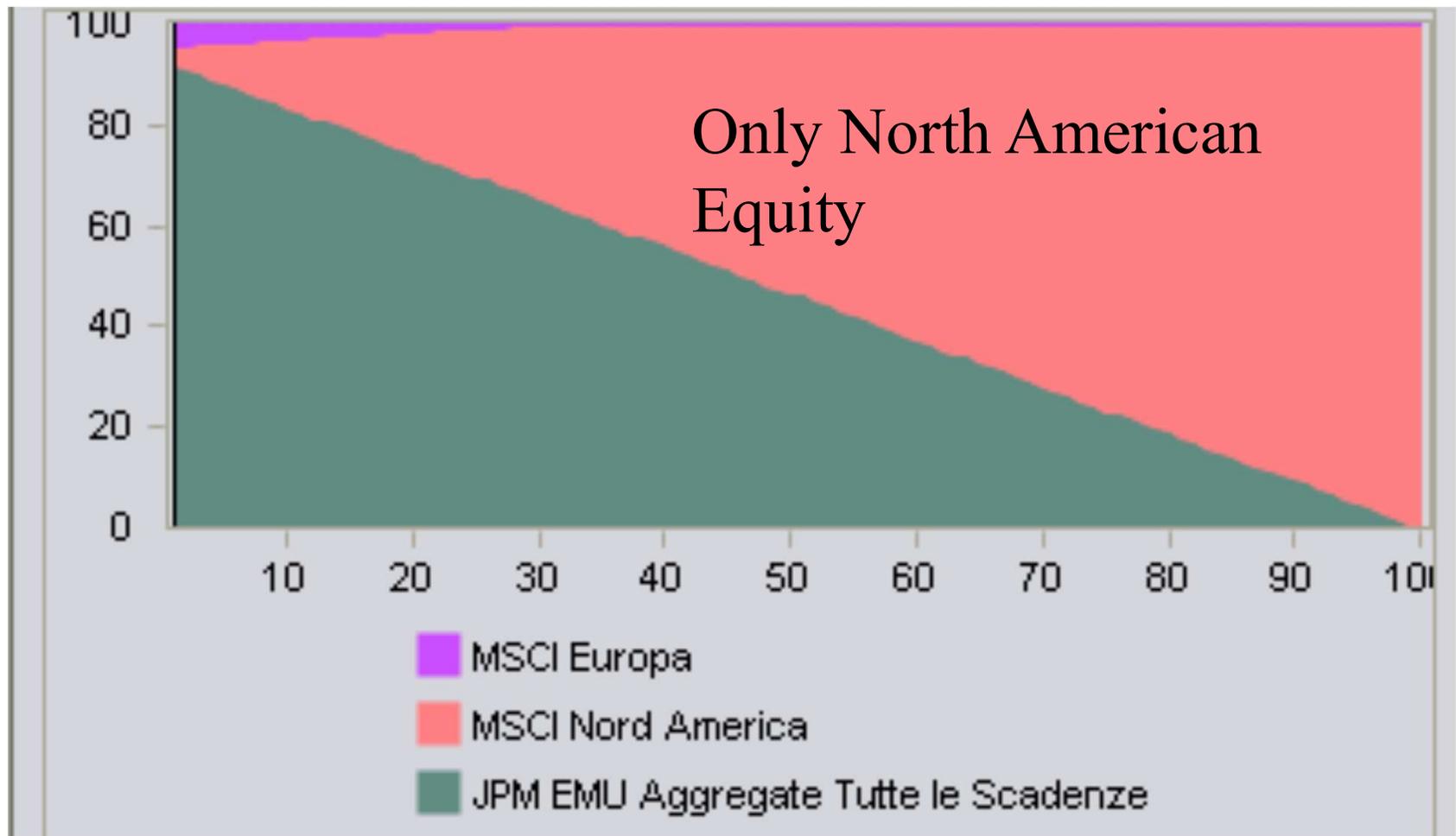
Asset Management Committee “Beta” has the same expectations. The only difference is the following:

	Rendimento atteso %
JPM EMU Aggregate Tutte le	3
MSCI Nord America	7.4%
MSCI Europa	7%

Very homogenous forecast.....similar views about the future trend of the markets, but.....<sub>111</sub>

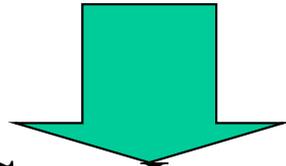
# Extra-argument 1: Efficient Portfolios are **instable** (4/5)

The portfolio composition is the opposite:



# Extra-argument 1: Efficient Portfolios are **instable** (5/5)

It is not encouraging/reassuring to realize that very small changes in expected returns can strongly affect the portfolio composition.



Question: Can I trust a model that give importance to basis points?

## Efficient portfolios are “estimation error maximizers”

Because of the estimation error, efficient portfolios are likely to have very bad performance.

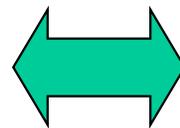
### Example:

- The Return of Emerg.Mkts Equity is expected to be 8.0% (the highest)  $\Rightarrow$
- Efficient portfolios with high risk are concentrated on Emerging Market Equity  $\Rightarrow$
- Ex-post we discover that the estimation is wrong, as this market collapses  $\Rightarrow$
- The concentration produces a “*bath blood*”.

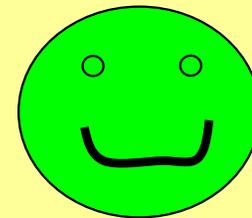
Efficient portfolios are  
“estimation error maximizers”

*Since estimation error is often large, portfolios selected according to the Markowitz criterion are likely not more efficient than a Naïve portfolio.*

MARKOWITZ



NAÏVE



# Agenda

## 6. “Putting Markowitz at work”

# Putting Markowitz at work

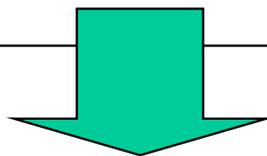
In order to “put Markowitz at work”, we need to remove the “*perfect estimation*” hypothesis

Asset Manager are not clairvoyant. They make mistakes.

Estimation error must be managed.

It is better to have portfolios with lower expected return, but with lower exposition to estimation error.

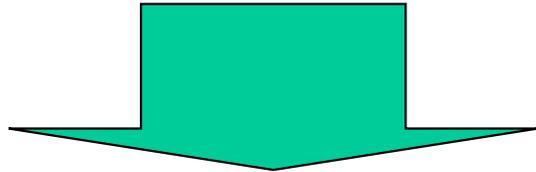
The problem is the portfolio concentration.



*We need to modify the model, in order to promote a greater diversification*

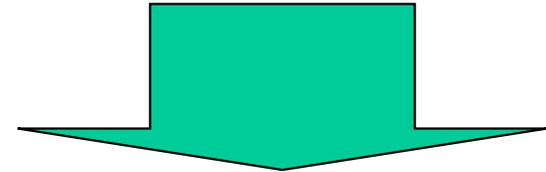
# Two techniques

**Heuristic Approaches**



**They adjust the  
Optimization Process**

**Bayesian Approaches**



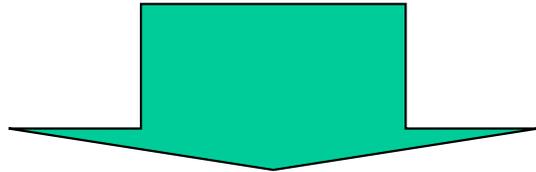
**They adjust the inputs  
estimated (*above all  
expected returns*)**

# Agenda

## 7. Heuristic techniques

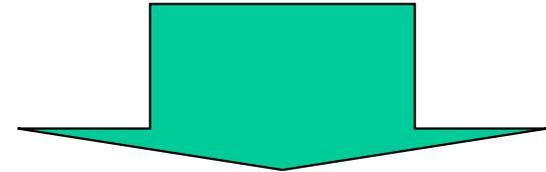
# Two Heuristic Approach

**Constrained Optimization**



**Easy**

**Resampling<sup>TM</sup>**



**Difficult**

# Constrained Optimization

It is necessary to add **supplementary constraints** to the Markowitz optimization

Find the weights ( $w_i$ ) able to:

*Objective function* →

**MIN**  $\sigma_{\text{Portfoglio}}$

*Constraints:*

*1<sup>st</sup> constraint:* →

Exp. Return =  $E(R)^*$

*2<sup>nd</sup> constraint:* →

$w_1 + \dots + w_i + \dots + w_n = 1$

*3<sup>rd</sup> constraint:* →

$w_i \geq 0$

*4<sup>th</sup> constraint:* →

$w_i \leq K_i$

**These Constraints drive a larger diversification**

# Constrained Optimization Example (1/3)

## Asset Classes selected:

Benchmark selezionati	
Nome	
JPM Euro 3 mesi	
JPM EMU Aggregate Tutte le Scadenze	
MSCI Europa	
MSCI Nord America	
MSCI Giappone	
MSCI Pacifico ex Giappone	
MSCI Emerging Market Free	

## Expected Returns estimated:

	Rendimento atteso %
JPM EMU Aggregate Tutte le	3,2
JPM Euro 3 mesi	2,8
MSCI Giappone	4,5
MSCI Pacifico ex Giappone	6
MSCI Nord America	6
MSCI Europa	7
MSCI Emerging Market Free	8

## Standard deviations estimated:

	Rischio atteso %
JPM EMU Aggregate Tutte le	1,8
JPM Euro 3 mesi	4,2
MSCI Giappone	22,8
MSCI Pacifico ex Giappone	23
MSCI Nord America	21
MSCI Europa	20
MSCI Emerging Market Free	29

## Correlations estimated:

Matrice delle correlazioni							
	JPM EMU Aggregate Tutte le	JPM Euro 3 mesi	MSCI Giappone	MSCI Pacifico ex Giappone	MSCI Nord America	MSCI Europa	MSCI Emerging Market Free
JPM EMU Aggregate	1	Storico	Storico	Storico	Storico	Storico	Storico
JPM Euro 3 mesi	0,27	1	Storico	Storico	Storico	Storico	Storico
MSCI Giappone	-0,26	-0,27	1	Storico	Storico	Storico	Storico
MSCI Pacifico ex	-0,27	-0,15	0,61	1	Storico	Storico	Storico
MSCI Nord America	-0,35	-0,13	0,61	0,74	1	Storico	Storico
MSCI Europa	-0,4	-0,2	0,53	0,72	0,85	1	Storico
MSCI Emerging Market	-0,34	-0,18	0,62	0,87	0,76	0,76	1

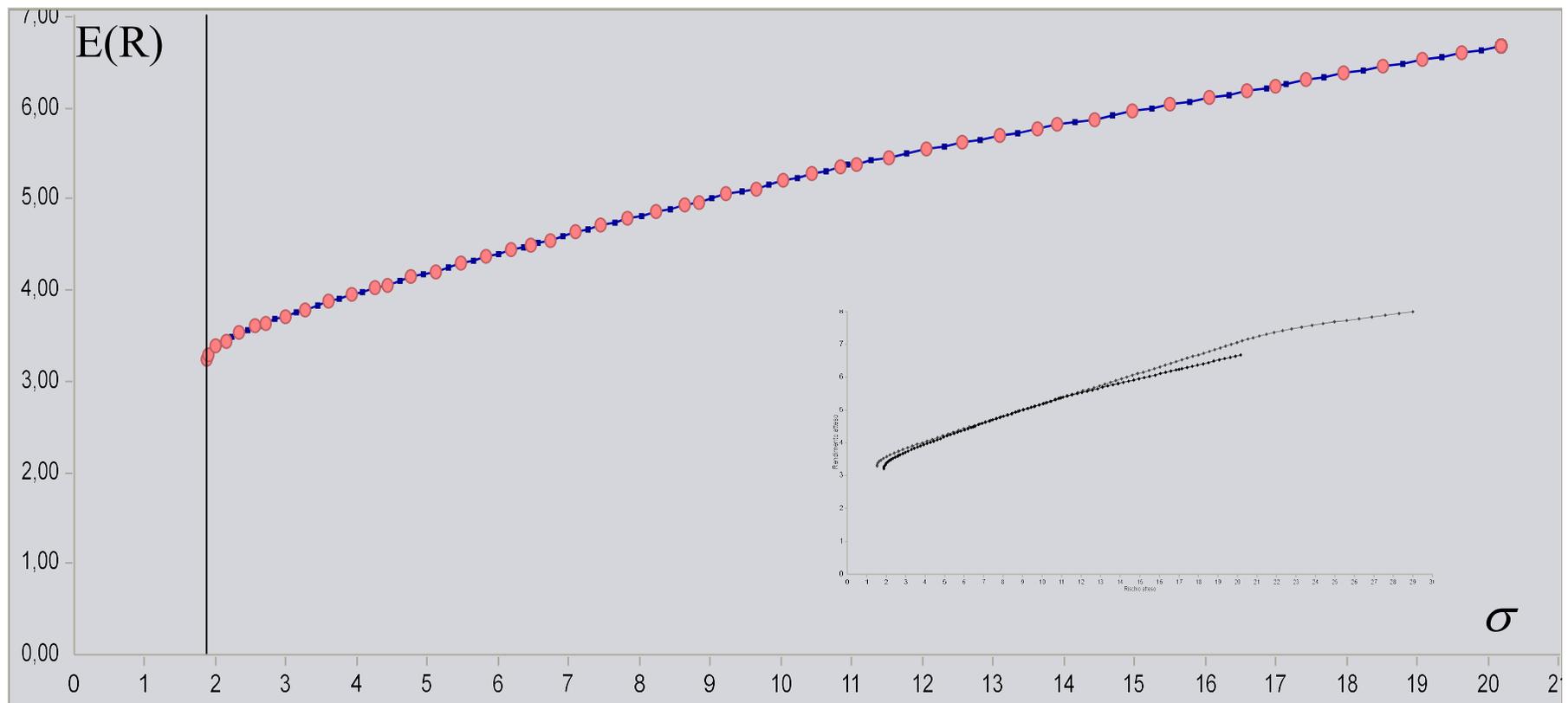
## Supplementary constraints

### Vincoli dei Benchmark

	min (%)	max (%)
JPM EMU Aggregate	0	60
JPM Euro 3 mesi	0	70
MSCI Giappone	0	20
MSCI Pacifico ex	0	8
MSCI Nord America	0	55
MSCI Europa	0	42
MSCI Emerging Market	0	13

# Constrained Optimization Example (2/3)

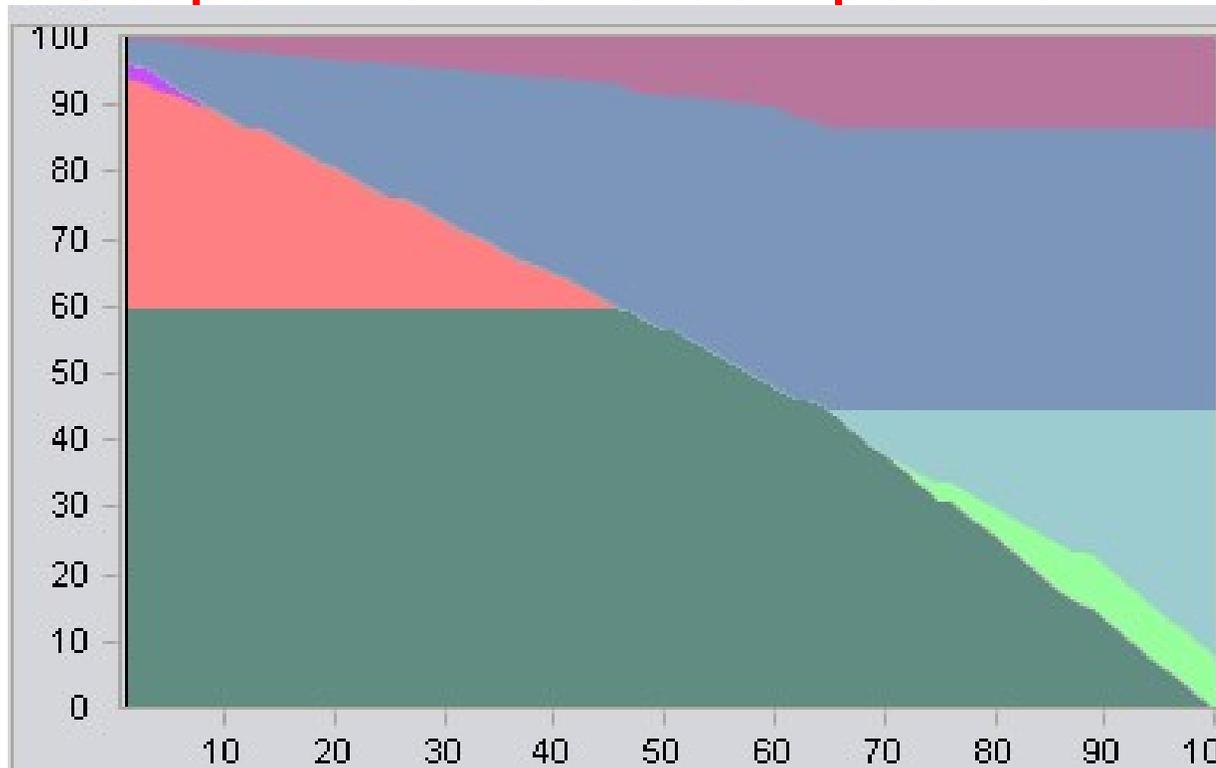
Output: Constrained Frontier



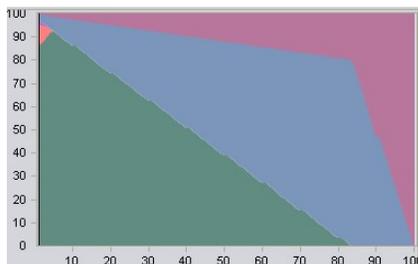
Constrained frontier is lower than the efficient frontier

# Constrained Optimization Example (3/3)

## Output: Portfolio composition



Diversification increases



# Extra-argument 2: Improving the Constrained Optimization

## Extra-argument 2:

### Improving the Constrained Optimization (1/4)

- We saw that using “traditional” constrained is hard to build all diversified portfolios;
- May-be, a portion of portfolios are well diversified.....but other portfolios are already concentrated;
- In order to well diversify all portfolios we can use different constraints named:

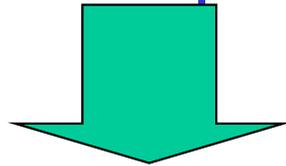
Infra-group constraints

## Extra-argument 2:

### Improving the Constrained Optimization (2/4)

#### Examples of infra-group constrains:

- At best, Emerging Market Equity is 18% of the equity portion (**upper bound**);
- North America Equity Market can not be less than 25% of the equity portion.



I use together upper & lower bound, so every risky asset is free to move inside a reasonable range.

# Extra-argument 2:

## Improving the Constrained Optimization Example: Input

(3/4)

### Asset Classes selected:

Benchmark selezionati
Nome
JPM Euro 3 mesi
JPM EMU Aggregate Tutte le Scadenze
MSCI Europa
MSCI Nord America
MSCI Giappone
MSCI Pacifico ex Giappone
MSCI Emerging Market Free

### Expected Returns estimated:

	Rendimento atteso %
JPM EMU Aggregate Tutte le	3,2
JPM Euro 3 mesi	2,8
MSCI Giappone	4,5
MSCI Pacifico ex Giappone	6
MSCI Nord America	6
MSCI Europa	7
MSCI Emerging Market Free	8

### Standard deviations estimated:

	Rischio atteso %
JPM EMU Aggregate Tutte le	1,8
JPM Euro 3 mesi	4,2
MSCI Giappone	22,8
MSCI Pacifico ex Giappone	23
MSCI Nord America	21
MSCI Europa	20
MSCI Emerging Market Free	29

### Correlations estimated:

Matrice delle correlazioni

	JPM EMU Aggregate Tutte le	JPM Euro 3 mesi	MSCI Giappone	MSCI Pacifico ex Giappone	MSCI Nord America	MSCI Europa	MSCI Emerging Market Free
JPM EMU Aggregate	1	Storico	Storico	Storico	Storico	Storico	Storico
JPM Euro 3 mesi	0,27	1	Storico	Storico	Storico	Storico	Storico
MSCI Giappone	-0,26	-0,27	1	Storico	Storico	Storico	Storico
MSCI Pacifico ex	-0,27	-0,15	0,61	1	Storico	Storico	Storico
MSCI Nord America	-0,35	-0,13	0,61	0,74	1	Storico	Storico
MSCI Europa	-0,4	-0,2	0,53	0,72	0,85	1	Storico
MSCI Emerging Market	-0,34	-0,18	0,62	0,87	0,76	0,76	1

## Infra-group constraints

### Azionario

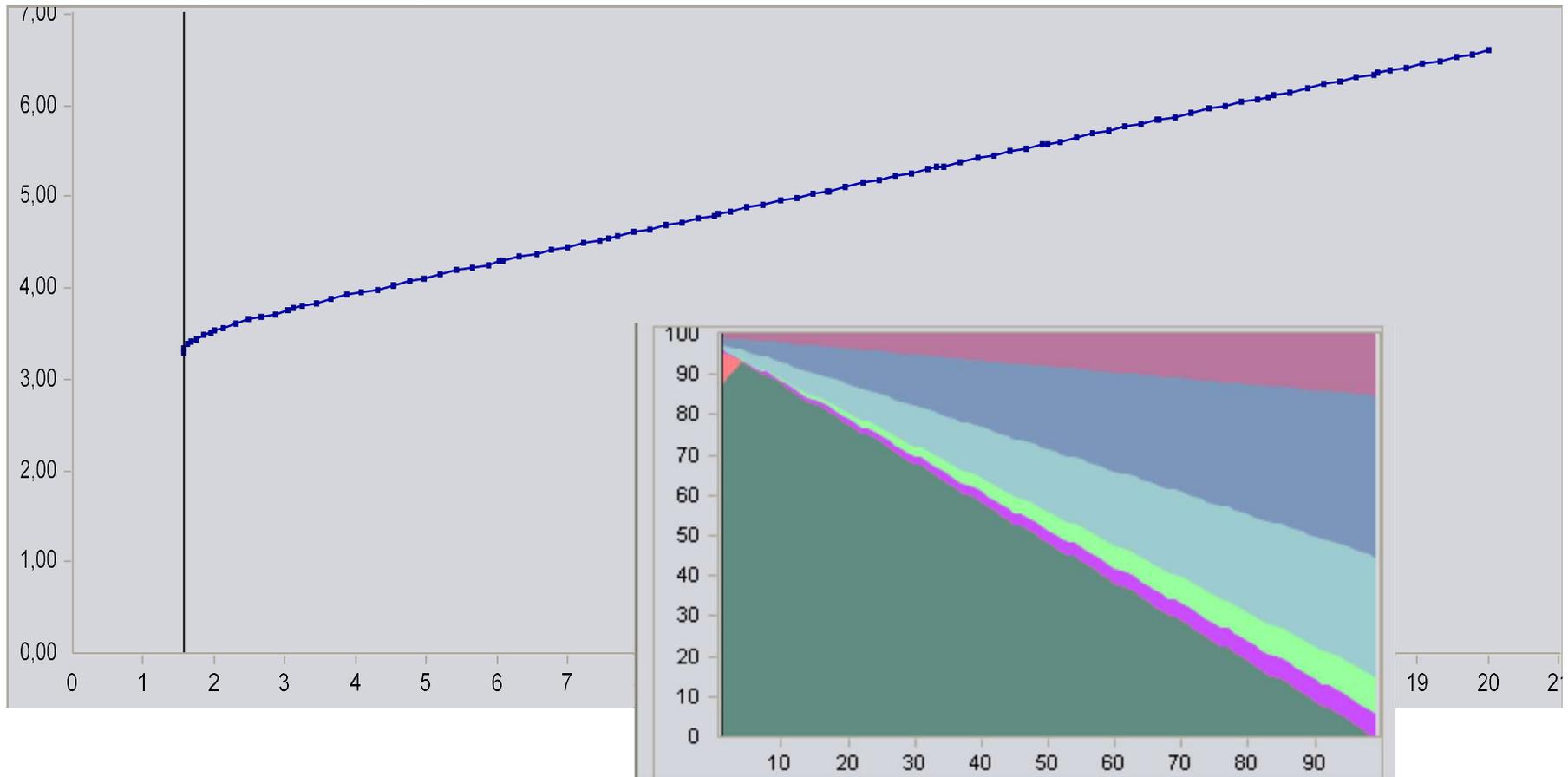
	min (%)	max (%)
MSCI Giappone	6	16
MSCI Pacifico ex	2	9
MSCI Nord America	25	60
MSCI Europa	15	40
MSCI Emerging Market	4	15

# Extra-argument 2:

## Improving the Constrained Optimization

(4/4)

Example: Output



All portfolios are well-diversified

# Resampling™

- Resampling is a methodology that force a certain level of portfolio diversification.
- Resampling is based on:
  1. The simulation of a large number of “statistically consistent” investment scenarios
  2. The simulated  $E(R)$ ,  $\sigma$  and  $\rho$  are used as input of a new Markowitz Optimization.
  3. After repeating steps 2. thousands of time the final portfolios (Resampled Portfolios) have the composition of the “average” efficient portfolio

# Resampling: Example

(1/3)

## Asset Classes selected:

Benchmark selezionati	
Nome	
JPM Euro 3 mesi	
JPM EMU Aggregate Tutte le Scadenze	
MSCI Europa	
MSCI Nord America	
MSCI Giappone	
MSCI Pacifico ex Giappone	
MSCI Emerging Market Free	

## Expected Returns estimated:

	Rendimento atteso %
JPM EMU Aggregate Tutte le	3,2
JPM Euro 3 mesi	2,8
MSCI Giappone	4,5
MSCI Pacifico ex Giappone	6
MSCI Nord America	6
MSCI Europa	7
MSCI Emerging Market Free	8

## Standard deviations estimated:

	Rischio atteso %
JPM EMU Aggregate Tutte le	1,8
JPM Euro 3 mesi	4,2
MSCI Giappone	22,8
MSCI Pacifico ex Giappone	23
MSCI Nord America	21
MSCI Europa	20
MSCI Emerging Market Free	29

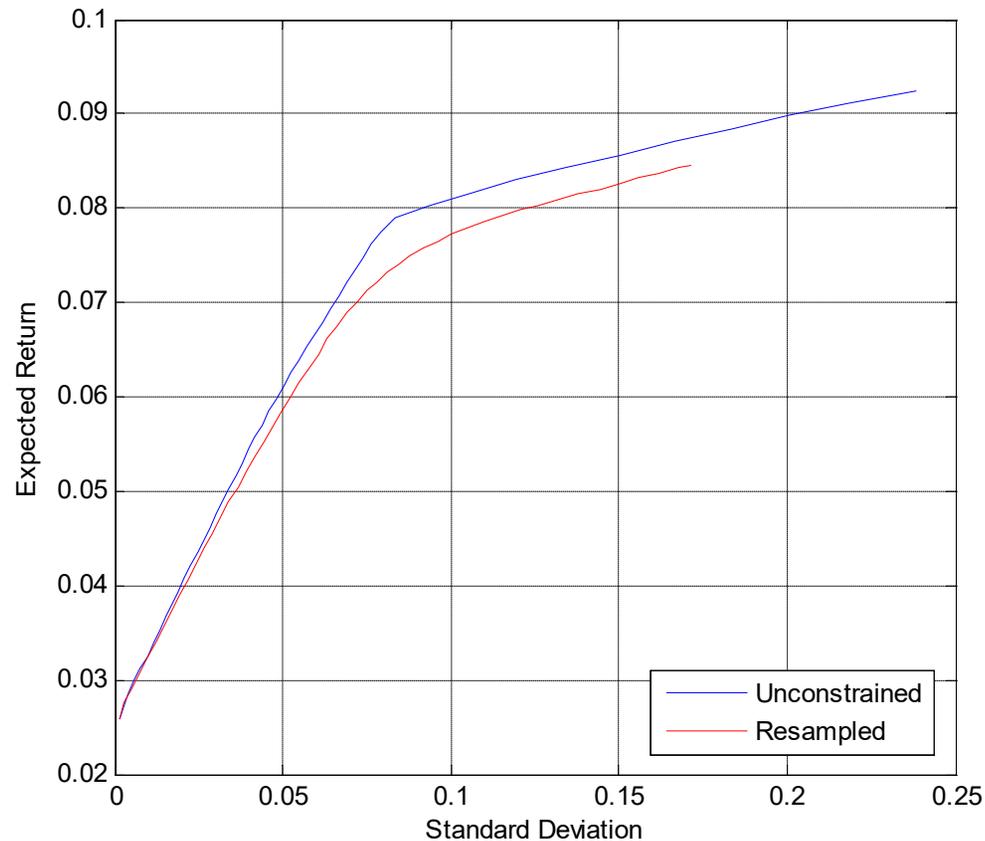
## Correlations estimated:

Matrice delle correlazioni							
	JPM EMU Aggregate Tutte le	JPM Euro 3 mesi	MSCI Giappone	MSCI Pacifico ex Giappone	MSCI Nord America	MSCI Europa	MSCI Emerging Market Free
JPM EMU Aggregate	1	Storico	Storico	Storico	Storico	Storico	Storico
JPM Euro 3 mesi	0,27	1	Storico	Storico	Storico	Storico	Storico
MSCI Giappone	-0,26	-0,27	1	Storico	Storico	Storico	Storico
MSCI Pacifico ex	-0,27	-0,15	0,61	1	Storico	Storico	Storico
MSCI Nord America	-0,35	-0,13	0,61	0,74	1	Storico	Storico
MSCI Europa	-0,4	-0,2	0,53	0,72	0,85	1	Storico
MSCI Emerging Market	-0,34	-0,18	0,62	0,87	0,76	0,76	1

# Resampling: Example

(2/3)

## Output: Resampled Frontier



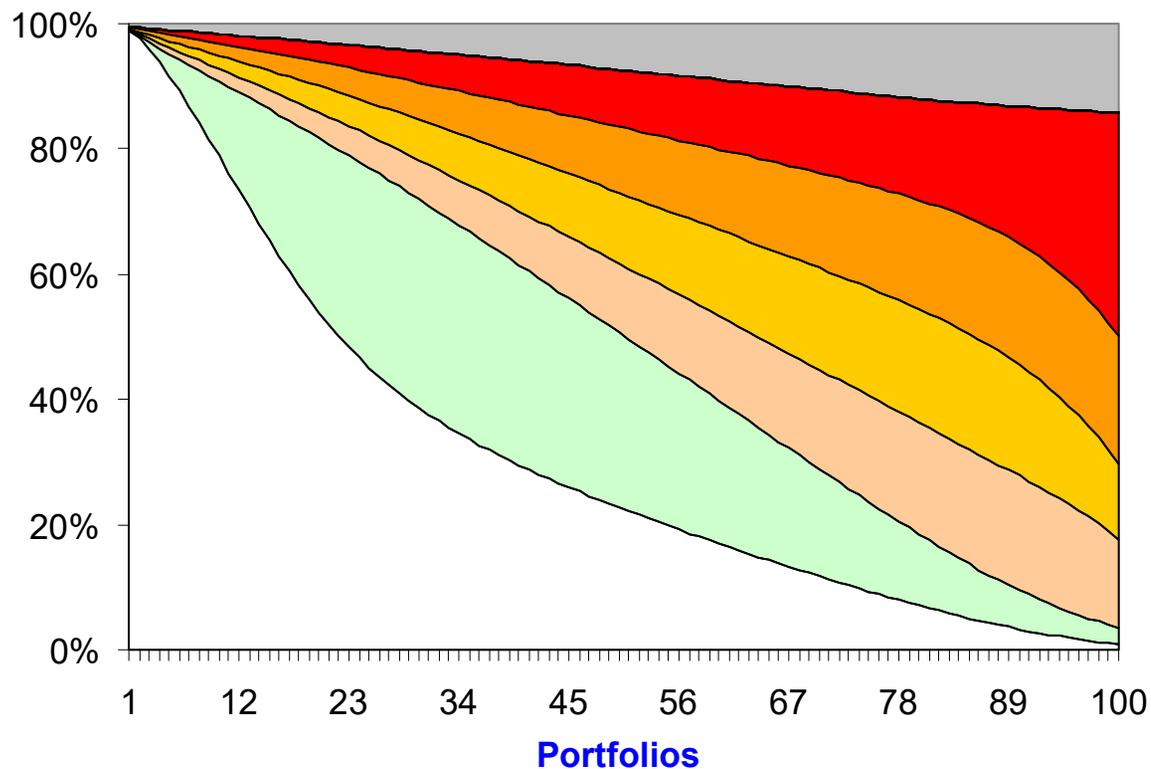
Resampled frontier is lower than the efficient frontier

# Resampling: Example

(3/3)

## Output: Portfolio composition

Weights

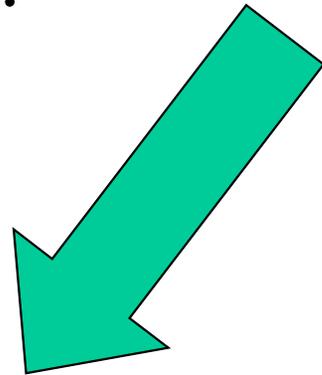


Diversification increases

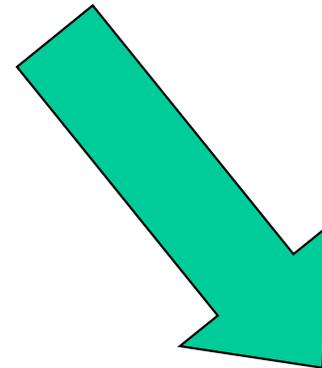
Extra-argument 3:  
A deeper analysis of Resampling<sup>TM</sup>

# Extra-argument 3: A deeper analysis of Resampling (1/7)

In order to process the resampling technique we need:



Markowitz  
Optimization



Simulation  
Process

# Extra-argument 3:

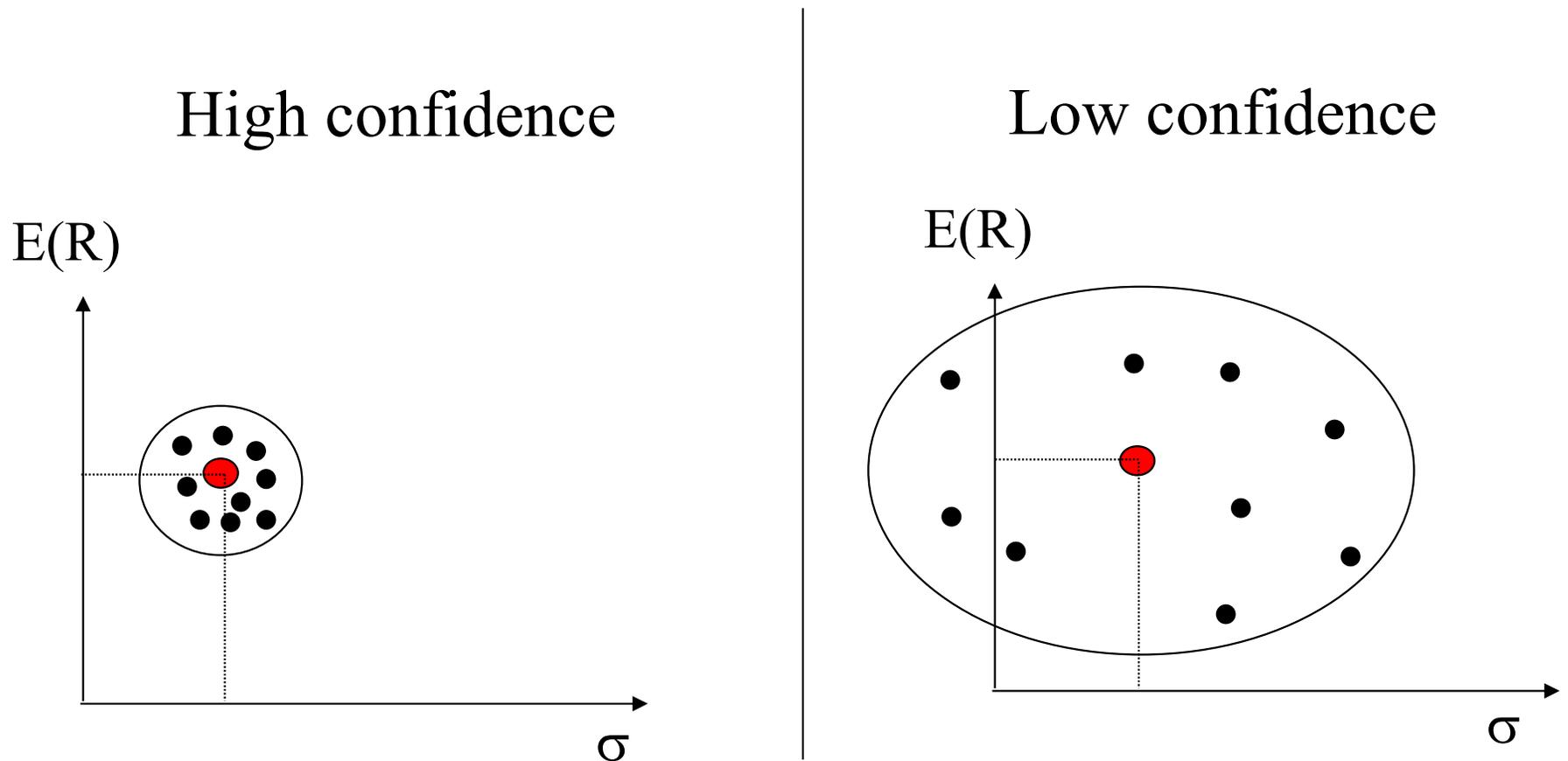
## A deeper analysis of Resampling (2/7)

### **The need to simulate returns:**

- We know that our expectations can be wrong;
- So in order to incorporate uncertainty, we can run a simulation process that return behaviours of market returns that are different from our expectation.

# Extra-argument 3: A deeper analysis of Resampling (3/7)

## Simulation: A graphical representation



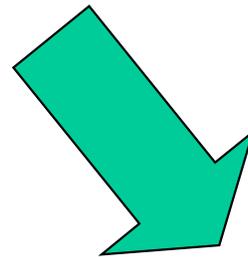
- Expectation
- Simulation

## Extra-argument 3:

# A deeper analysis of Resampling (4/7)

### What do we need in order to simulate?

- Forecasts ( $\Rightarrow E(R), \sigma, \rho$ )
- Confidence on estimations
- Random process that is able to make deviations from the expectation.



Simulation

# Extra-argument 3:

## A deeper analysis of Resampling (5/7)

### Example:

Microsoft Excel - FileExcel\_Pomante

Digitare una domanda.

From ESTIMATION..... To SIMULATION

	MSCI Europe	MSCI USA	MSCI Japan	MSCI EM
E(R)	7,0%	6,0%	4,5%	8,0%
$\sigma$	20,0%	21,0%	22,8%	29,0%

	MSCI Europe	MSCI USA	MSCI Japan	MSCI EM
E(R)	2,7%	2,0%	4,6%	4,3%
$\sigma$	21,5%	21,2%	20,3%	36,5%

	MSCI Europe	MSCI USA	MSCI Japan	MSCI EM
MSCI Europe	1			
MSCI USA	0,85	1		
MSCI Japan	0,60	0,65	1	
MSCI EM	0,76	0,76	0,65	1

	MSCI Europe	MSCI USA	MSCI Japan	MSCI EM
MSCI Europe	1			
MSCI USA	0,91	1		
MSCI Japan	0,54	0,64	1	
MSCI EM	0,90	0,85	0,62	1

Path Simulations

Cap

Time

LOW CONFIDENCE

Historical Series Graph1 Graph2 Graph3 Graph4 VaR Naive Ottim SimMC1 SimMC2 Tactical Weights

Pronto Fare clic per iniziare.

start Risorse del computer Finali Cartel1 FileExcel\_Pomante ExtraSlides MBA P... Slides MBA Pomant... Benchmark & Style... IT 11.07

# Extra-argument 3:

## A deeper analysis of Resampling (6/7)

**If we are able to simulate, we can estimate the  
Resampled portfolios**

# Extra-argument 3: A deeper analysis of Resampling (7/7)

Stage 1: Expectations

	MSCI Europe	MSCI USA	MSCI Japan	MSCI EM
E(R)	7,0%	6,0%	4,5%	8,0%
$\sigma$	20,0%	21,0%	22,8%	29,0%

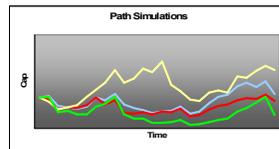
  

	MSCI Europe	MSCI USA	MSCI Japan	MSCI EM
MSCI Europe	1			
MSCI USA	0,85	1		
MSCI Japan	0,60	0,65	1	
MSCI EM	0,76	0,76	0,65	1

Stage 2: Confidence

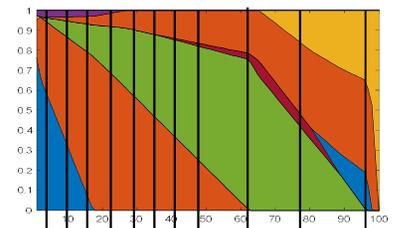
Very low-Low-Medium-High-Very high

Stage 2+1: 1<sup>st</sup> Simul.

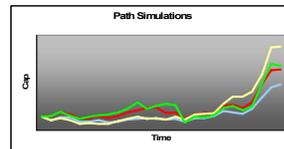


	MSCI Europe	MSCI USA	MSCI Japan	MSCI EM
E(R)	8,0%	6,9%	6,2%	2,3%
$\sigma$	16,0%	21,5%	29,2%	23,5%

Optimiz. →

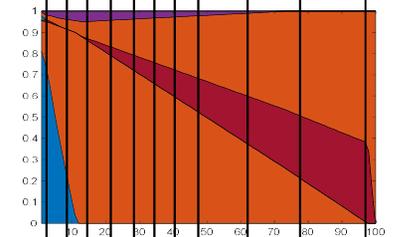


Stage 2+2: 2<sup>nd</sup> Simul.

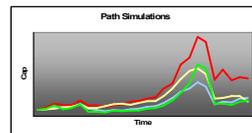


	MSCI Europe	MSCI USA	MSCI Japan	MSCI EM
E(R)	7,1%	8,6%	10,1%	11,6%
$\sigma$	20,9%	23,0%	23,3%	31,8%

Optimiz. →



Stage 2+3000: 3k<sup>st</sup> Simul.



	MSCI Europe	MSCI USA	MSCI Japan	MSCI EM
E(R)	7,6%	10,6%	6,0%	11,0%
$\sigma$	23,7%	25,4%	21,6%	33,3%

Optimiz. →

Average composition

