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# Correlations and Structured Products: Basket Derivatives and Certificates

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20541– Advanced Tools for Risk Management and  
Pricing

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# Multi-Underlyings Structured Products

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Until now we have only considered structured products composed by derivatives written on a single underlying; yet structured products on more than one underlying are common

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- Many types of multi-underlying structured products:
  - Products (notes/certificates) on a **basket** of underlyings (e.g., an equally weighted basket of shares)
  - Products whose payoff depends on the performance of the **worst** performing underlying
    - An option on the worst performing of a basket of underlying will always cost less than the option on the linear basket of the same underlyings
  - Products whose payoff depends on the performance of the **best** performing underlying
    - An option on the best performing of a basket of underlying will always cost more than the option on the linear basket of the very same underlyings

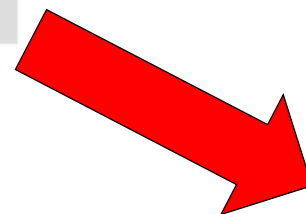
# Pricing of Multi-Underlyings Structured Products

Pricing Multi-Underlyings structured products is similar to pricing products on a single underlying, but we need an additional input: the correlation among the underlying assets – i.e. their **correlation matrix**

- Consider, for example, the correlation matrix of IBM, EADS and Rio Tinto:

|           | IBM  | EADS | Rio Tinto |
|-----------|------|------|-----------|
| IBM       | 1.00 | 0.36 | 0.22      |
| EADS      | 0.36 | 1.00 | 0.08      |
| Rio Tinto | 0.22 | 0.08 | 1.00      |

Correlation Coefficients



**Cholesky decomposition**

|        |        |        |
|--------|--------|--------|
| 1.0000 | 0.0000 | 0.0000 |
| 0.3600 | 0.9329 | 0.0000 |
| 0.2227 | 0.0020 | 0.9749 |

# Pricing of Multi-Underlyings Structured Products

The Choleski decomposition of the correlation matrix is then used to obtain/generate correlated random variables  $\epsilon$

- Pricing simulated paths:
  1. Generate a set of three correlated random numbers
  2. Multiply this set of numbers by the Cholesky decomposition of the correlation matrix

The diagram illustrates the process of generating correlated random variables. It shows a vector of three random numbers (labeled 1) multiplied by a Cholesky decomposition matrix (labeled 2) to produce a vector of three correlated random variables. The resulting vector is circled in red, and a red arrow points to it from the text below.

$$\begin{bmatrix} -0.063 \\ 0.836 \\ -0.558 \end{bmatrix} \begin{bmatrix} 1.000 & 0.000 & 0.000 \\ 0.360 & 0.933 & 0.000 \\ 0.223 & 0.002 & 0.975 \end{bmatrix} = \begin{bmatrix} -0.063 \\ 0.757 \\ -0.556 \end{bmatrix}$$

**Correlated random  
variables  $\epsilon$**

# Pricing of Multi-Underlyings Structured Products

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The Choleski decomposition of the correlation matrix is then used to obtain correlated random variables  $\phi$

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- Pricing paths:
  1. Generate a set of three correlated random numbers
  2. Multiply this set of numbers by the Cholesky decomposition of the correlation matrix
  3. Integrate the correlated random numbers into the pricing path of the shares:

$$S_i(t + \delta t) = S_i(t) \exp \left( \left( r - \frac{1}{2} \sigma_i^2 \right) \delta t + \sigma_i \sqrt{\delta t} \phi_i \right)$$

Where  $\phi_i$  is the random shock originated as discussed in the previous slide

# Pricing of Multi-Underlyings Structured Products

|    | A   | B      | C                | D    | E         | F         | G        | H        | I |
|----|---|--------|------------------|------|-----------|-----------|----------|----------|---|
| 1  | Asset1  | Asset2 |                  | Time | Random1   | Random2   | Asset1   | Asset2   |   |
| 2  | 100   | 80     |                  | 0    | 0.046223  | -1.59903  | 100      | 80       |   |
| 3  |   |        |                  | 0.01 | -0.158143 | -0.960557 | 99.78371 | 77.97375 |   |
| 4  | Drift1  | Drift2 |                  | 0.02 | -0.540749 | 0.340648  | 98.80434 | 78.18732 |   |
| 5  | 0.1   | 0.2    | = D3 + \$B\$12   | 0.03 | 0.859933  | -1.754755 | 100      |          |   |
| 6  |   |        |                  | 0.04 | -0.268174 | 0.896078  | 100      |          |   |
| 7  | Vol1  | Vol2   |                  | 0.05 | -0.810562 | -2.361049 | 98.0     |          |   |
| 8  | 0.2   | 0.3    | = RAND() + RAN   | 0.06 | -0.974247 | 0.569597  | 96.8     |          |   |
| 9  |   |        | D() + RAND() + R | 0.07 | 0.576045  | 1.016849  | 98.0     |          |   |
| 10 | Correl.   | 0.5    | AND() + RAND()   | 0.08 | -0.969892 | -0.409346 | 96.7     |          |   |
| 11 |   |        | + RAND() + RAN   | 0.09 | -0.839252 | -1.013799 | 94.0     |          |   |
| 12 | Timestep  | 0.01   | D() + RAND() + R | 0.10 | 0.372974  | -0.409777 | 95.4     |          |   |
| 13 |   |        | AND() + RAND()   | 0.11 | -0.542291 | -0.597359 | 94.8     |          |   |
| 14 | Sqrt(1-correl^2)  |        | + RAND() + RAN   | 0.12 | 0.248432  | -0.643216 | 95.0     |          |   |
| 15 | 0.866025  |        | D() -6           | 0.13 | 0.963828  | 1.237832  | 97.0     |          |   |
| 16 |   |        |                  | 0.14 | 0.591412  | 2.049829  | 98.2     |          |   |
| 17 |   |        |                  | 0.15 | -0.243018 | 0.321826  | 97.0     |          |   |
| 18 | = SQRT(1-B10*B10)   |        |                  | 0.16 | 0.558761  | 1.187003  | 99.078   | 79.25634 |   |
| 19 |   |        |                  | 0.17 | -0.951554 | 1.985109  | 97.29151 | 82.37122 |   |
| 20 |   |        |                  | 0.18 | 0.502183  | -1.61082  | 98.36597 | 79.70918 |   |
| 21 | = G19*(1 + A5*\$B\$12 + A8*SQRT(\$B\$12)*E20)                         |        |                  | 0.19 | 0.72      | -1.502543 | 95.92222 | 75.212   |   |
| 22 |   |        |                  | 0.20 | 0.348241  | -1.424413 | 96.68622 | 72.97191 |   |
| 23 |   |        |                  | 0.21 | 2.09392   | -1.355961 | 100.832  | 72.8391  |   |
| 24 |   |        |                  | 0.22 | 1.044282  | -1.60787  | 103.0387 | 71.08298 |   |
| 25 |   |        |                  | 0.23 | 0.902542  | 1.108671  | 105.0017 | 74.23496 |   |
| 26 |   |        |                  | 0.24 | 1.517426  | -1.260644 | 109.2934 | 73.64175 |   |
| 27 | = H24*(1 + B5*\$B\$12 + B8*SQRT(\$B\$12)*(\$B\$10*E25 + \$A\$15*F25)) |        |                  | 0.25 | 0.8629    | 0.8629    | 108.629  | 75.89836 |   |
| 28 |   |        |                  | 0.26 | -0.640458 | 0.532359  | 107.3462 | 76.37077 |   |
| 29 |   |        |                  | 0.27 | 0.344599  | -0.17648  | 108.1934 | 76.5681  |   |
| 30 |   |        |                  | 0.28 | 1.082126  | 0.483641  | 110.6431 | 78.92619 |   |
| 31 |   |        |                  | 0.29 | 0.041153  | -0.865378 | 110.8448 | 77.35825 |   |
| 32 |   |        |                  | 0.30 | -0.029683 | 0.034647  | 110.8899 | 77.54816 |   |

An example of how the pricing path of two correlated shares can be simulated into Excel

# Pricing of Multi-Underlyings Structured Products

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Virtually any of the payoffs studied in the previous lectures can have multi-underlyings; let's consider for example a Bonus Cap

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- At maturity a Bonus Cap pays an amount higher than Eur 100 if the underlying has never touched the barrier during the life of the product, otherwise it pays  $\text{Min} [\text{Cap}, \text{Underlying}(T)/\text{Strike}]$ 
  - The underlying may be a single shares or a single index  
=> Already discussed
  - The underlying can be an equally weighted basket of shares/indices  
=> Underlying value at any  $t$  is equal to the average of the values of the shares / indices at time  $t$
  - The underlying can have a “worst of” feature on a basket of shares/indices  
=> Only the worst performing share/index is observed

# Pricing of Multi-Underlyings Structured Products

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All else being equal, which certificate will have the higher Bonus amount? The one on linear basket or the one with Worst Of feature?

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- It is evident that a certificate with the Worst Of feature will deliver a higher bonus amount than a certificate on a linear basket, as it implies higher risk, as it is showed in the next slides

## **Payoff 1**

Bonus Cap on linear Basket of Fiat and Eni

Strike = 15 Eur (average 13 and 17)

Barrier = 80% (12 Eur ), only observed at maturity

Maturity = 3 months

## **Payoff 2**

Bonus Cap Worst of on Fiat and Eni

Strike (Fiat 13 Eur; Eni =17 Eur )

Barrier = 80% ( Fiat =10.4; Eni=13.6) only observed at maturity

Maturity = 3 months



# Pricing of Multi-Underlyings Structured Products

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All else being equal, which certificate will have the higher Bonus amount? The one on linear basket or the one with Worst Of feature?

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- Scenario 1 = Both Eni and Fiat at maturity are above 80% of their initial value

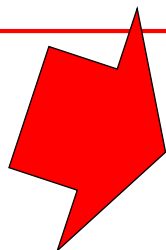
**Payoff 1 → Bonus Amount**

**Payoff 2 → Bonus Amount**

- Scenario 2 = One of the shares (suppose Eni) is below 80% of its initial value (suppose that Eni = 12 Eur , Fiat = 15 Eur, average =13.5 Eur)

**Payoff 1 → Bonus Amount (13.5 is higher than 12)**

**Payoff 2 → Eur 100 x WorstOf (T)/WorstOf (0) = Eur 100 x 12/17 = 70.5 Euro**



**Higher risk, higher return  
(higher Bonus amount)**

# Bonus Cap Example

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Choosing shares with higher correlation increases the Bonus amount of a Bonus Cap on a linear portfolio; shares with higher correlation decreases the Bonus of a Bonus Cap under Worst Of

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- You recall that the Bonus Cap is replicated with ZCB plus a short Down & In put with barrier equal to the barrier of the product
  - The Down & In put is more expensive when the correlation is higher
  - You have more money to spend for the Bonus which will increase as correlations increase
  - When the Bonus Cap includes the “Worst Of” features, as correlations increase, the value of the put declines because it becomes unlikely that one underlying may decouple from others
  - You have less money to spend for the Bonus which will decrease as correlations increase