

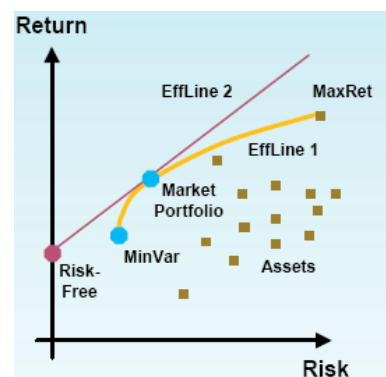
L'approche de Black-Litterman

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Motivations

- L'approche de Markowitz suggère d'utiliser les rendements anticipés et la matrice de covariance pour construire la frontière efficiente
- le CAPM nous donne les relation d'équilibre: tout investisseur détient un portefeuille combinaison linéaire du portefeuille de marché et de l'actif sans risque



En absence d'information supplémentaire, il suffit d'acheter l'indice pour être efficient

Mais qu'en est il si on possède des anticipations propres (vues) sur les rendements anticipés ? Comment concilier "au mieux" ces anticipations avec celles du marché ?

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Exemple: les secteurs européens

| Asset | hist. Return | hist. Volatility | Market Cap | beta |
|-------|-----------------|---------------------|---------------|------|
| AUTO | 8.32% | 25.09% | 1.65% | 1.42 |
| BANK | 15.14% | 21.21% | 15.04% | 0.96 |
| BRES | 7.31% | 23.56% | 1.22% | 1.09 |
| CHEM | 12.25% | 20.81% | 1.80% | 0.99 |
| CONS | 6.56% | 18.92% | 1.26% | 0.97 |
| CYCL | 5.24% | 19.94% | 2.85% | 1.30 |
| CNYL | 11.80% | 16.66% | 2.90% | 0.71 |
| ENGY | 14.92% | 20.72% | 10.30% | 0.80 |
| FISV | 13.01% | 20.91% | 4.12% | 0.60 |
| FBEV | 10.47% | 15.72% | 4.59% | 0.78 |
| INDS | 13.45% | 19.35% | 5.19% | 1.11 |
| INSU | 17.43% | 19.68% | 6.89% | 1.55 |
| MEDA | 14.63% | 25.17% | 3.27% | 1.20 |
| PHRM | 22.83% | 16.20% | 10.24% | 0.67 |
| RETL | 9.49% | 16.16% | 2.27% | 0.74 |
| TECH | 25.95% | 28.60% | 11.03% | 1.92 |
| TELE | 18.99% | 25.69% | 10.56% | 0.87 |
| UTLY | 11.77% | 17.25% | 4.82% | 0.54 |

gamma 3

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Le portefeuille optimal est obtenu en recherchant la solution d'un problème d'optimisation quadratique (voir cours) :

$$\max_w w^T \Pi - \frac{1}{2} \gamma w^T \Sigma w \quad (1)$$

avec w les pondérations, Σ la matrice de covariance, γ l'aversion au risque et Π le vecteur des rendements espérés.

Sans contrainte sur les pondérations (ventes à découvert et emprunts autorisés), cette optimisation conduit à la solution suivante (voir cours) :

$$w^* = \frac{1}{\gamma} \Sigma^{-1} \Pi \quad (2)$$

Notes :

- $w_0 = \sum w_i$ représente la proportion investie dans l'actif sans risque. Il y a emprunt lorsque $w_0 < 0$.
- Π représente le vecteur des rendements en excès du rendement sans risque. Les rendements sont $\Pi' = \Pi + r_{free}$. Le programme d'optimisation peut s'écrire de manière équivalente :

$$\max_w w^T \Pi' + w_0 r_{free} - \frac{1}{2} \gamma w^T \Sigma w$$

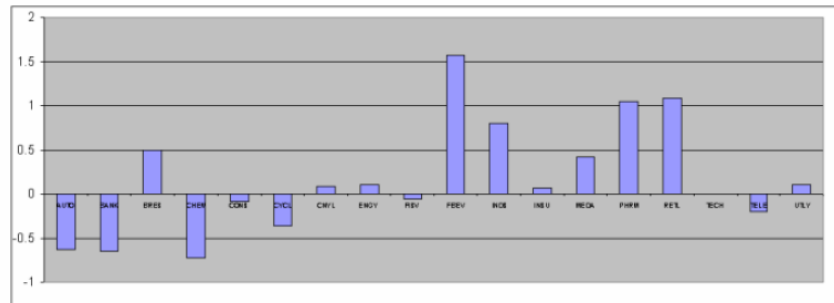
avec $w_0 = 1 - \sum_{i=1,n} w_i$. Dans la suite les rendements sont des rendements en excès $\Pi' - r_{free}$.

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les inconvénients de l'approche de Markowitz

Dans un premier temps, on considère des rendements espérés tous égaux, $\Pi_i = 0.13$ pour tout $i = 1, 18$. On prendra $\gamma = 3$, comme étant l'aversion au risque standard.

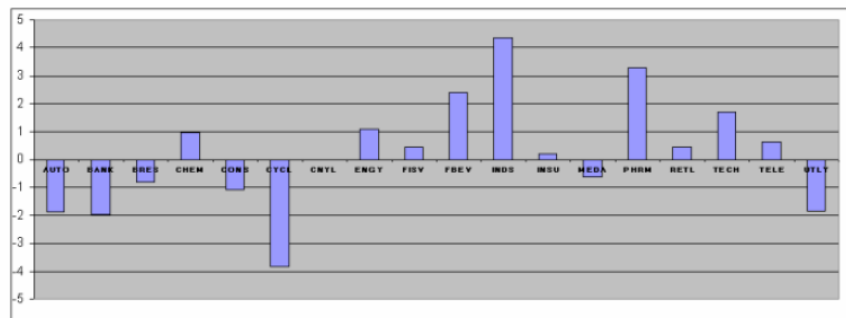


○ L'utilisation de rendements espérés arbitraire conduit à un portefeuille peu réaliste ...

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Utilisation des rendements historiques



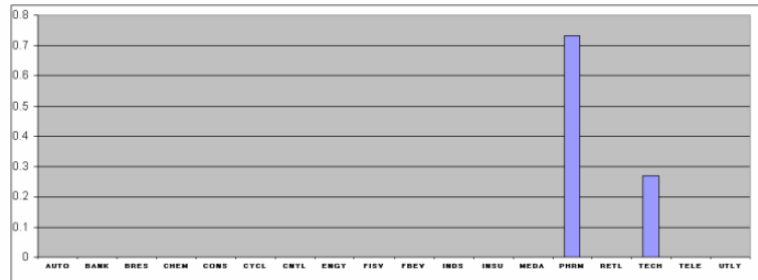
○ L'utilisation des rendements historiques conduit à des pondérations extrêmes

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Ajouts de contraintes d'optimisation

○ pas de vente à découvert, ni levier



○ Il ne reste que quelques actifs ...

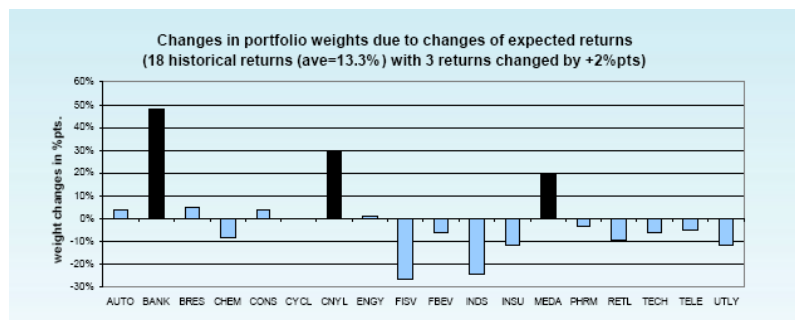
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Limitations de l'approche Espérance - Variance

○ Forte sensibilité aux rendements espérés

☞ de petites modifications des rendements conduisent à des grandes variations des pondérations



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Rendements implicites

- Etant donné un indice que l'on suppose être sur la frontière efficiente (portefeuille de marché), on peut en déduire les rendements attendus implicites
- Le portefeuille de marché est proportionnel à

$$\Sigma w$$

avec

w le vecteur des pondérations des capitalisations

Σ la matrice de covariance

$$\Pi = \gamma \Sigma w$$

Π représente les rendements implicites du marché

γ est l'aversion au risque (de l'ordre de 3 à 4)

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Estimation de l'aversion au risque

Satchell & Scowcroft and Best & Grauer:

$$\text{Let } \gamma = (r_M - r_f) / \sigma_M^2$$

where (a) σ_M = StDev(Market Portfolio) or (b) $\sigma_M^2 = w^T \Omega w$, w = market cap.

Zimmermann et al.: (a) σ_M = 16.95% p.a. from STOXX-Data (own calculation)

Let $\gamma = 3$, which corresponds to a risk premium of 8.6%.

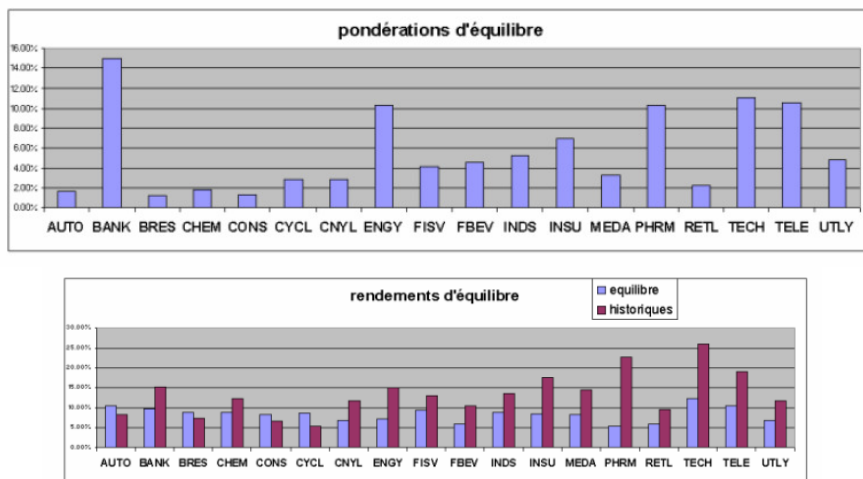
(b) Using $w^T \Omega w$ yields σ_M = 16.85% p.a. \Rightarrow risk premium = 8.5%.

En pratique, le coefficient d'aversion pour le risque est de l'ordre de 4

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$$\Pi = \gamma \Sigma w_{cap}$$



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$$\lambda = \frac{E(r) - r_f}{\sigma^2} = \frac{\text{Risk Premium}}{\text{Variance}}$$

$\lambda = \frac{\text{Risk Premium}}{\text{Variance}}$

- Assume a Risk Premium of 4%
- The Historical (Long-term) Standard Deviation for the Market Portfolio is 10.84%. The Variance is the Standard Deviation Squared, 1.117%

$$\lambda = \frac{\text{Risk Premium}}{\text{Variance}}$$

$$\lambda = \frac{4.00\%}{1.117\%}$$

$$\lambda = 3.404$$

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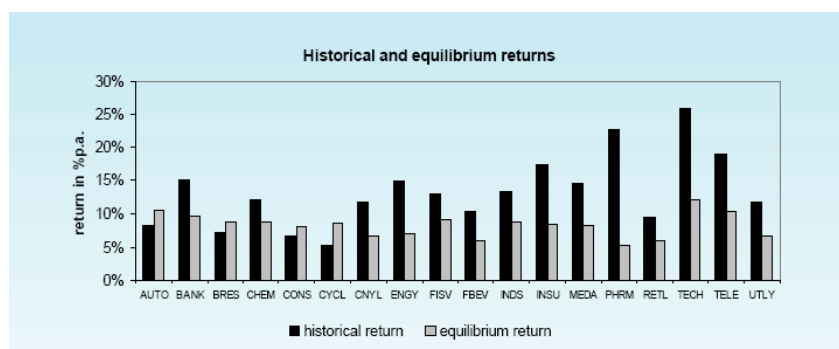
Rendements Implicites

| Symbol | Historical Return Vector | CAPM Return Vector | Implied Equilibrium Return Vector (Π) |
|-----------|--------------------------------|--------------------------|--|
| aa | 17.30 | 15.43 | 13.81 |
| ge | 16.91 | 12.15 | 13.57 |
| jnj | 16.98 | 9.39 | 9.75 |
| msft | 23.95 | 14.89 | 20.41 |
| axp | 15.00 | 15.65 | 14.94 |
| gm | 4.59 | 13.50 | 12.83 |
| jpm | 5.31 | 15.89 | 16.46 |
| pg | 7.81 | 8.04 | 7.56 |
| ba | -4.18 | 14.16 | 11.81 |
| hd | 29.38 | 11.59 | 12.52 |
| ko | -0.57 | 10.95 | 10.92 |
| sb | 10.10 | 7.76 | 8.79 |
| c | 24.55 | 16.59 | 16.97 |
| hon | -0.05 | 16.89 | 14.50 |
| mcd | 2.74 | 10.70 | 10.44 |
| t | -1.24 | 8.88 | 10.74 |
| cat | 7.97 | 13.08 | 10.92 |
| hwp | -4.97 | 14.92 | 14.45 |
| mmm | 9.03 | 10.43 | 8.66 |
| utx | 13.39 | 16.51 | 15.47 |
| dd | 0.44 | 12.21 | 10.98 |
| ibm | 21.99 | 13.47 | 14.66 |
| rno | 10.47 | 7.57 | 6.96 |
| wmt | 30.23 | 10.94 | 12.77 |
| dis | -2.59 | 12.89 | 12.41 |
| intc | 13.59 | 15.83 | 18.70 |
| mrk | 8.65 | 8.95 | 9.22 |
| xom | 11.10 | 8.39 | 7.88 |
| ek | -17.00 | 11.08 | 10.61 |
| ip | 1.24 | 14.80 | 12.92 |
| Average | 9.07 | 12.45 | 12.42 |
| Std. Dev. | 10.72 | 2.88 | 3.22 |
| High | 30.23 | 16.89 | 20.41 |
| Low | -17.00 | 7.57 | 6.86 |

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Rendements implicites vs historiques - Secteurs du STOXX



- These implicit returns serve as reference returns for the following investigations („market neutral starting point “)
- Note that equilibrium returns are just calculated; they do not require any estimation procedure.

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Rendements implicites vs Historiques - Actions du DJIA

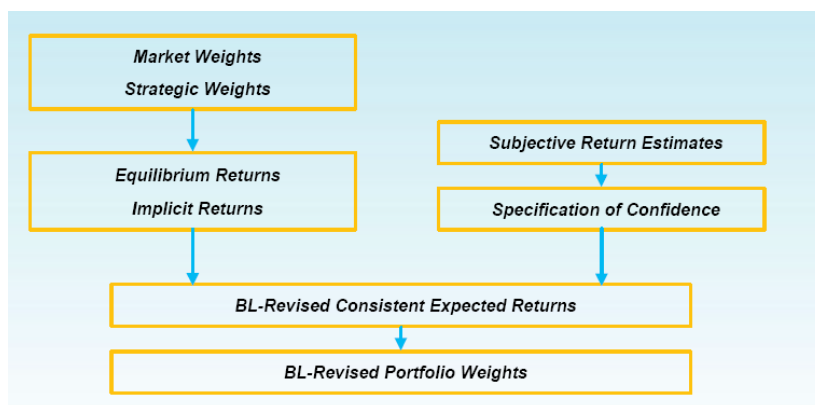
| Symbol | Historical Weight | CAPM Weight | Implied Equilibrium Weight | Market Capitalization Weight |
|--------|-------------------|-------------|----------------------------|------------------------------|
| aa | 223.86% | 2.67% | 0.88% | 0.88% |
| ge | -65.44% | 9.80% | 11.62% | 11.62% |
| jinj | -70.08% | 6.11% | 5.29% | 5.29% |
| msft | 3.54% | 3.22% | 10.41% | 10.41% |
| axp | -15.38% | 5.54% | 1.39% | 1.39% |
| gm | 5.76% | 3.44% | 0.79% | 0.79% |
| jpm | -213.39% | 1.94% | 2.09% | 2.09% |
| pg | 92.00% | -1.33% | 2.99% | 2.99% |
| ba | -111.35% | 4.71% | 0.90% | 0.90% |
| hd | 280.01% | 0.11% | 3.49% | 3.49% |
| ko | -151.58% | 5.70% | 3.42% | 3.42% |
| sbc | 17.11% | -4.28% | 3.84% | 3.84% |
| c | 293.90% | 5.11% | 7.58% | 7.58% |
| hon | 15.65% | 2.71% | 0.80% | 0.80% |
| mcd | -61.68% | 1.32% | 0.99% | 0.99% |
| t | -86.44% | 4.04% | 1.87% | 1.87% |
| cat | -70.67% | 5.10% | 0.52% | 0.52% |
| hwp | -163.02% | 6.60% | 1.16% | 1.16% |
| mmm | 56.84% | 4.73% | 1.35% | 1.35% |
| utx | -23.80% | 4.38% | 0.88% | 0.88% |
| dd | -131.99% | 1.03% | 1.29% | 1.29% |
| ibm | 36.92% | 5.57% | 6.08% | 6.08% |
| mo | 136.78% | 1.31% | 2.90% | 2.90% |
| wmt | 21.03% | 0.89% | 7.49% | 7.49% |
| dis | 5.75% | -2.35% | 1.23% | 1.23% |
| intc | 97.81% | -1.96% | 6.16% | 6.16% |
| mrk | 144.34% | 4.61% | 3.90% | 3.90% |
| xom | 218.75% | 4.10% | 7.85% | 7.85% |
| ek | -148.36% | 2.04% | 0.25% | 0.25% |
| ip | -113.07% | 4.76% | 0.57% | 0.57% |
| High | 293.90% | 9.80% | 11.62% | 11.62% |
| Low | -213.39% | -4.28% | 0.25% | 0.25% |

Not surprisingly, the Historical Return Vector produces an extreme portfolio. However, despite the similarity between the CAPM Return Vector and the Implied Equilibrium Return Vector (\mathbf{II}), the vectors produce two rather distinct portfolios (the correlation coefficient (ρ) is 18%). The CAPM-based portfolio contains four short positions and almost all of the weights are significantly different than the benchmark market capitalization weighted portfolio. As one would expect (since the process of extracting the Implied Equilibrium returns given the market capitalization weights was reversed), the Implied Equilibrium Return Vector (\mathbf{II}) leads back to the market capitalization weighted portfolio. In the absence of views that differ from the Implied Equilibrium return, investors should hold the market portfolio. The Implied Equilibrium Return Vector (\mathbf{II}) is the market-neutral starting point for the Black-Litterman Model.

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L'approche de Black Litterman



Permet de combiner les relations d'équilibre à long terme
avec des anticipations estimées à court terme

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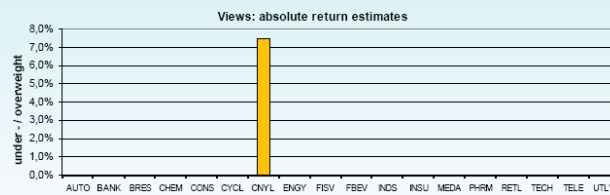
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Black-Litterman Approach

Absolute Views

- An **absolute View** can be formulated as follows: „The sektor of non-cyclical goods will perform better than stated by the equilibrium return of 6.66%. Our target return is 7.5% with 90% of confidence for an interval of $\pm 1.5\%$ “.

$$1 \cdot E(R_{CNIL}) = 7.5\% + (0.91\%)^2$$



abus de notation $(0.91\%)^2$ désignant une
erreur normale d'écart type 0.91%

Example of an Absolute View

- Opinion: "I think that International Equity is going to do well."
- View: International Developed Equity will have a return of 11%
- Confidence of View: 55%

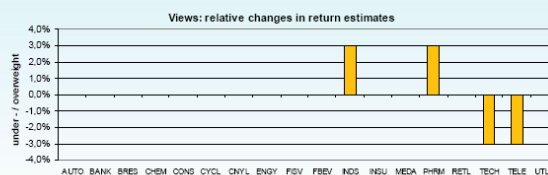
Black-Litterman Approach

Relative Views

- A **relative View** can be formulated as follows: „The sectors Pharmacy and Industry will outperform Telecom and Technology by 3% with a confidence of 90% for $\pm 1\%$ “:

$$\begin{aligned} & [w_{PHRM} \cdot E(R_{PHRM}) + w_{INDU} \cdot E(R_{INDU})] \\ & - [w_{TELE} \cdot E(R_{TELE}) + w_{TECH} \cdot E(R_{TECH})] = 3\% + (0.61\%)^2 \end{aligned}$$

- Thus: A *long*-portfolio with outperformers, a *short*-portfolio with underperformers.



If all views are relative views then only the stocks which are part of those views will have their weights affected. If there are absolute views then all the stocks will have change in returns and weights, since each individual return is linked to the other returns via the covariance matrix of returns.

Example of a Relative View

- Opinion: "I believe that Growth is going to outperform Value."
- View: US Large Growth and US Small Growth will outperform US Large Value and US Small Value by 3%.
- Confidence of View: 80%

Exprimer les vues - formalisme

- Relative and absolute views form a system of linear equations as a constraint to the optimization problem:

$$P \cdot E(R) = V + e$$

where ($k = \#Views$ and $n = \#Assets$, with $k \leq n$):

$E(R)$ = $n \times 1$ vector of expected asset returns, unknown
 P = $k \times n$ matrix, representing the Views
 V = $k \times 1$ vector, absolute / relative return expectations (i.e., levels or over-/underperforming)
 e = $k \times 1$ vector of squared StDev's

Ω = Diagonal covariance matrix of error terms in expressed views

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Combined Views

- Combining the aforementioned Views using $P \cdot E = V + e$, we get:

$$P \cdot \begin{pmatrix} E(R_{AUTO}) \\ \vdots \\ E(R_{UTLY}) \end{pmatrix} = \begin{pmatrix} 3\% \\ 7.5\% \end{pmatrix} + \begin{pmatrix} (0.61\%)^2 \\ (0.91\%)^2 \end{pmatrix}$$

- with

$$P = \begin{pmatrix} View1, rel. \\ View2, abs. \end{pmatrix} = \begin{pmatrix} 0 \cdots 0 & 0.34 & 0 \cdots 0 & 0.66 & 0 & -0.51 & -0.49 & 0 \\ 0 & \cdots & 0 & 1 & 0 & \cdots & \cdots & 0 \end{pmatrix}$$

long positions
short positions

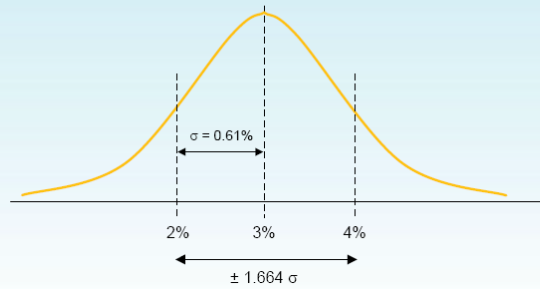
asset selection for absolute estimate

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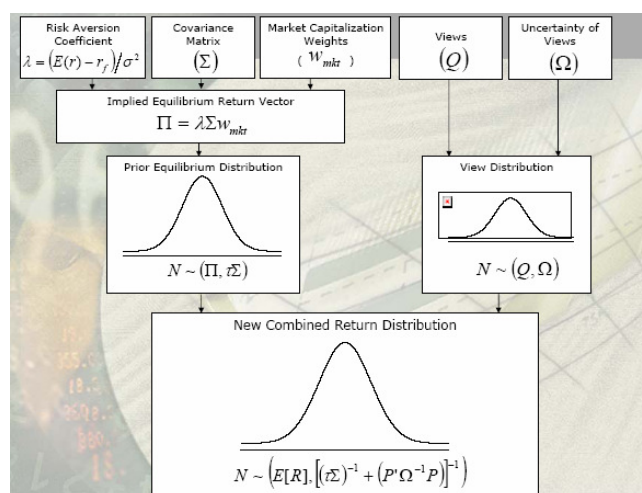
Remark: Technical note on „confidence“

- Comment on e (outlined for the relative return estimate): The fact that the amount of relative outperformance of $3\% \pm 1\%$ is assigned a 90% probability is interpreted within a normal distribution, with **mean** = $\mu = 3\%$ and **variance** = $VAR = \sigma^2 = (0.61\%)^2 \equiv e$.



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Approche bayésienne - rappel

A=rendements anticipés (les vues)

B=rendements d'équilibre (Π)

Prob(A,B)

$$= \text{Prob}(A|B) * \text{Prob}(B)$$

$$= \text{Prob}(B|A) * \text{Prob}(A)$$

$$\text{Prob}(A|B) = \text{Prob}(B|A) * \text{Prob}(A) / \text{Prob}(B)$$

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Black-Litterman - une approche bayésienne

$$\Pr(E(r)|\Pi) = \frac{\Pr(\Pi|E(r)) \Pr(E(r))}{\Pr(\Pi)}$$

$$PE(r) = V + e$$

$$PE(r) \sim N(V, \Omega)$$

$$\Pi|E(r) \sim N(E(r), \tau\Sigma)$$

Ω = matrice diagonale k*k

P = matrice des vues

avec Σ la matrice de variance covariance

et τ une pondération que l'on se donne sur les rendements historiques qui dépend de la confiance

de l'investisseur dans les rendements historiques implicites versus ses propres anticipations

τ petit \rightarrow erreur sur Π petite

\rightarrow confiance plus grande dans les rendements implicites

\rightarrow moins de confiance dans les vues

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La meilleure estimation des rendements est obtenue en minimisant la variance autour des rendements d'équilibre

$$\min_{E[R]} [E[R] - \Pi]^T (\tau \Sigma)^{-1} [E[R] - \Pi]$$

sous les contraintes

$$P \cdot E[R] = \begin{cases} V & \text{avec des vues certaines} \\ V + e & \text{avec des vues incertaines} \end{cases}$$

La solution est

a faire en exercice ...

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La formule de Black Litterman

$$E[R] = \left[(\tau \Sigma)^{-1} + P' \Omega^{-1} P \right]^{-1} \left[(\tau \Sigma)^{-1} \Pi + P' \Omega^{-1} V \right]$$

k = number of views

n = number of assets

$E[R]$ = New (posterior) Combined Return Vector ($n \times 1$ column vector)

τ = Scalar $\tau = 0.3$ "plausible" (used for numerical evaluations throughout).

Σ = Covariance Matrix of Returns ($n \times n$ matrix)

P = Identifies the assets involved in the views ($k \times n$ matrix or $1 \times n$ row vector in the special case of 1 view)

Ω = Diagonal covariance matrix of error terms in expressed views representing the level of confidence in each view ($k \times k$ matrix)

Π = Implied Equilibrium Return Vector ($n \times 1$ column vector)

V = View Vector ($k \times 1$ column vector)

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Exemple simple

$$E[R] = \left[(\tau \Sigma)^{-1} + P^T \Omega^{-1} P \right]^{-1} \left[(\tau \Sigma)^{-1} \Pi + P^T \Omega^{-1} V \right]$$

Prenons le cas d'un seul actif

$$\Sigma = (\sigma^2)$$

$$\Pi = (\pi)$$

$$P = (1)$$

$$V = (v)$$

$$\Omega = (\omega^2)$$

$$\sigma = 20\%$$

$$\pi = 30\%$$

$$v = 35\%$$

$$\omega = 3\%$$

$$\tau = 0.3$$

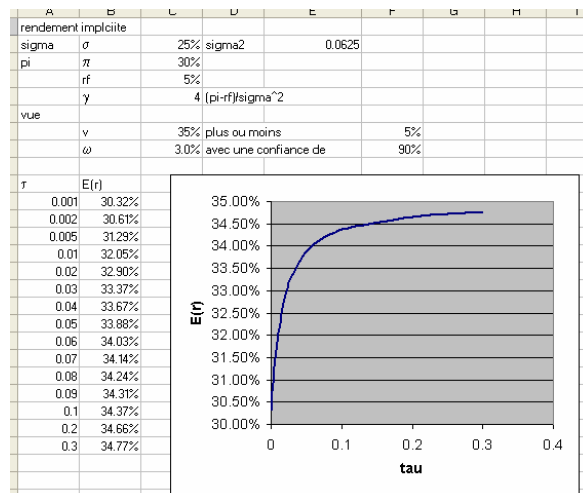
$$E[R] = 34.77\%$$

$$E[R] = \frac{\frac{\pi}{\tau \sigma^2} + \frac{q}{\omega^2}}{\frac{1}{\tau \sigma^2} + \frac{1}{\omega^2}}$$

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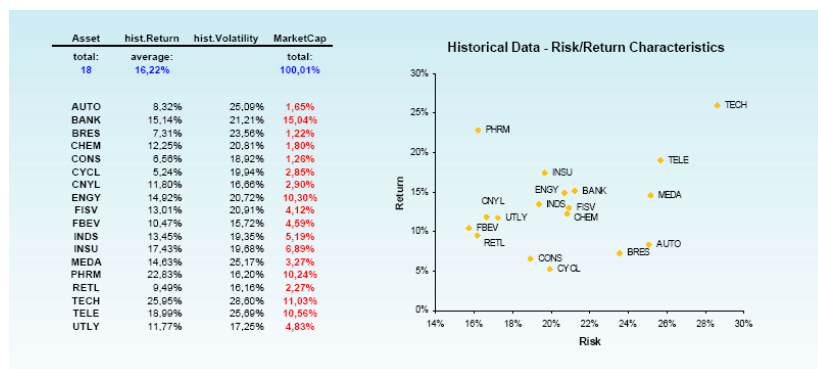
Exemple - simple



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Rendement mensuels - secteurs STOXX



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Correlation matrix of Dow Jones STOXX sectors

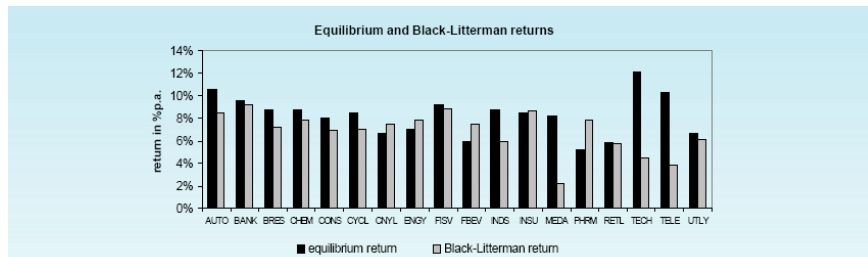
■ Calculation based on monthly returns

| | AUTO | BANK | BRES | CHEM | CONS | CYCL | CNYL | ENGY | FISV | FBEV | INDS | INSU | MEDA | PHRM | RETL | TECH | TELE | UTLY |
|------|------|------|------|------|------|------|------|------|------|------|------|------|------|------|------|------|------|------|
| AUTO | 100% | 74% | 73% | 83% | 78% | 75% | 73% | 55% | 73% | 71% | 79% | 72% | 46% | 43% | 68% | 69% | 65% | 64% |
| BANK | 74% | 100% | 63% | 73% | 74% | 75% | 71% | 59% | 82% | 75% | 75% | 87% | 39% | 63% | 62% | 67% | 59% | 64% |
| BRES | 73% | 63% | 100% | 63% | 81% | 78% | 60% | 66% | 69% | 56% | 78% | 56% | 44% | 31% | 56% | 62% | 52% | 41% |
| CHEM | 83% | 73% | 83% | 100% | 85% | 82% | 72% | 67% | 72% | 74% | 82% | 70% | 51% | 45% | 69% | 66% | 57% | 57% |
| CONS | 78% | 74% | 81% | 85% | 100% | 96% | 72% | 66% | 75% | 76% | 89% | 64% | 54% | 36% | 67% | 66% | 63% | 64% |
| CYCL | 75% | 75% | 78% | 82% | 96% | 100% | 67% | 64% | 79% | 70% | 87% | 63% | 56% | 43% | 67% | 76% | 63% | 59% |
| CNYL | 73% | 71% | 60% | 72% | 72% | 67% | 100% | 65% | 69% | 74% | 69% | 74% | 41% | 68% | 73% | 63% | 59% | 71% |
| ENGY | 55% | 59% | 66% | 67% | 66% | 64% | 55% | 100% | 59% | 59% | 59% | 54% | 28% | 43% | 55% | 43% | 32% | 46% |
| FISV | 73% | 92% | 69% | 72% | 75% | 79% | 69% | 59% | 100% | 75% | 73% | 85% | 39% | 61% | 58% | 64% | 57% | 58% |
| FBEV | 71% | 75% | 56% | 74% | 76% | 70% | 75% | 59% | 75% | 100% | 62% | 74% | 27% | 63% | 61% | 40% | 41% | 66% |
| INDS | 79% | 75% | 78% | 82% | 88% | 87% | 60% | 59% | 73% | 62% | 100% | 65% | 72% | 38% | 68% | 82% | 77% | 67% |
| INSU | 72% | 87% | 56% | 70% | 64% | 63% | 74% | 54% | 85% | 74% | 65% | 100% | 36% | 67% | 61% | 60% | 56% | 68% |
| MEDA | 46% | 39% | 44% | 51% | 54% | 56% | 41% | 28% | 39% | 27% | 72% | 36% | 100% | 21% | 42% | 73% | 77% | 57% |
| PHRM | 45% | 63% | 31% | 46% | 39% | 43% | 56% | 43% | 61% | 63% | 38% | 67% | 21% | 100% | 45% | 35% | 37% | 66% |
| RETL | 66% | 62% | 59% | 69% | 67% | 67% | 73% | 55% | 58% | 61% | 66% | 61% | 42% | 43% | 100% | 62% | 53% | 57% |
| TECH | 69% | 67% | 62% | 65% | 66% | 70% | 53% | 43% | 64% | 40% | 82% | 60% | 73% | 35% | 52% | 100% | 81% | 55% |
| TELE | 65% | 69% | 62% | 67% | 63% | 63% | 59% | 32% | 57% | 41% | 77% | 56% | 77% | 37% | 53% | 81% | 100% | 70% |
| UTLY | 64% | 64% | 41% | 57% | 64% | 56% | 71% | 46% | 58% | 66% | 67% | 68% | 57% | 58% | 57% | 55% | 70% | 100% |

■ Covariance matrix via $\sigma_{ij} = \sigma_i \sigma_j \rho_{ij}$

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Returns s.t. Views:

- *Short portfolio* Return expectations significantly lowered for TECH und TELE.
- *Long portfolio* Return expectation higher in PHRM but lower in INDS (ok, because the *relative View* „INDS better than TELE und TECH“ remains intact !)
- For CNYL, expected return shifts from 6.66% to 7.48% (90% confidence in estimate of 7,5%).
- Via correlations (e.g.) MEDA (75% to TECH, 77% to TELE) has significantly lower return.

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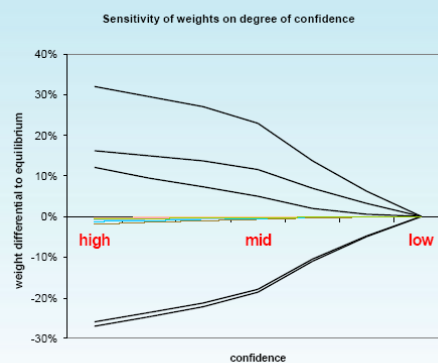
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Influence of degree of confidence on BL-returns and BL-weights, II

Observations:

- For low degrees of confidence the BL-weights converge to the weights in equilibrium (=market cap's).
- Weights approach equilibrium values from either underweighting (*short*) or overweighting (*long*) path.
- Significant weight changes for the assets with View.



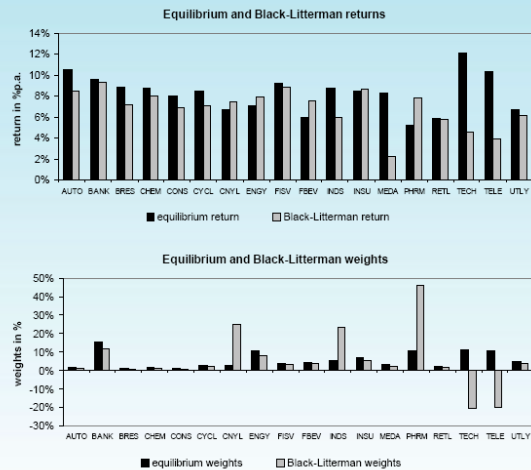
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Black-Litterman Approach

Strong confidence

- Large changes in weights due to the "strong views" (change up to 35%pt)



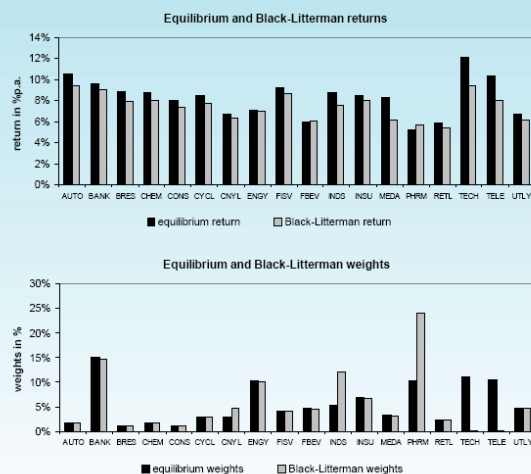
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Black-Litterman Approach

Mid confidence

- Moderate changes in weights (change up to 14%pt)



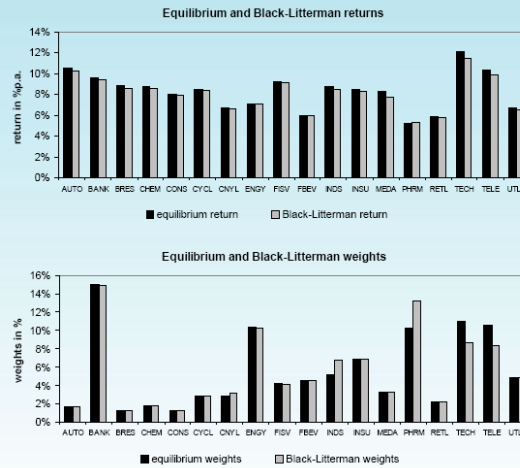
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Black-Litterman Approach

Poor confidence

- Weights close to equilibrium weights (change up to 3 %pt)



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Black-Litterman Approach - Conclusion I

Traditional vs BL approach

| | simple approach | Black-Litterman |
|-------------------|---|--|
| Return estimates: | for <u>each</u> asset assumed as <u>certain</u> <u>absolute</u> returns subjective | for <u>selected</u> assets degree of <u>confidence</u> <u>relative or absolute</u> estimates consistent |
| Reference return | none | equilibrium returns |

- Black F. und Litterman R.: *Global Portfolio Optimization*, Fin.Analysts Journal, Sep.1992, p.28-43
- Black F. und Litterman R.: *Asset Allocation: Combining investor views with market equilibrium*, Goldman-Sachs, Fixed Income Research, Sep.1990
- Zimmermann H., Drobetz W. und Oertmann P.: *Global Asset Allocation: New methods and applications*, publ. by Wiley & Sons, Nov.2002

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Annexe - calculs

One minimizes the objective:

$$\min_{\mu} (\mu - \Pi)'(\tau\Sigma)^{-1}(\mu - \Pi)$$

under the constraint $P\mu = Q$, and obtain:

$$\mu' = \Pi + \Sigma P'(P\Sigma P')^{-1}(Q - P\Pi')$$

- We introduce a vector of k Lagrange multipliers λ , and the problem becomes:

$$\min_{\mu} (\mu - \Pi)' \tau \Sigma^{-1} (\mu - \Pi) - 2\lambda'(P\mu - Q)$$

- Differentiating with respect to μ , we get:

$$\mu' \Sigma^{-1} - \Pi' \Sigma^{-1} + \lambda' P = 0$$

and so:

$$\mu = \Pi - \Sigma P' \lambda$$

- Plugging this result into the constraint, we solve for λ :

$$\lambda = (P\Sigma P')^{-1}(P\Pi - Q)$$

and hence:

$$\mu = \Pi + \Sigma P'(P\Sigma P')^{-1}(Q - P\Pi)$$

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- When views are associated with some error matrix ε , then the optimization objective is:

$$\min_{\mu} (\mu - \Pi)'(\tau\Sigma)^{-1}(\mu - \Pi) + (P\mu - Q)'\Omega^{-1}(P\mu - Q)$$

- The parameter τ measures the relative confidence in views vs. market equilibrium. A smaller τ indicates more confidence in equilibrium.
- The solution to the above problem is:

$$\mu = ((\tau\Sigma)^{-1} + P'\Omega^{-1}P)^{-1}((\tau\Sigma)^{-1}\Pi + P'\Omega^{-1}Q)$$

which is known as the **Master Formula**.

- Differentiating the objective, we obtain the equation:

$$((\tau\Sigma)^{-1} + P'\Omega^{-1}P)\mu - ((\tau\Sigma)^{-1}\Pi + P'\Omega^{-1}Q) = 0$$

- and therefore:

$$\mu = ((\tau\Sigma)^{-1} + P'\Omega^{-1}P)^{-1}((\tau\Sigma)^{-1}\Pi + P'\Omega^{-1}Q)$$

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[illegible]

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