## SIMPLIFY INPUTS --

## SINGLE INDEX CASE

A little later we will look at the solution after we have put some more structure on the problem. This will give us greater insight.
inputs
$\mathrm{N}=10$
$\mathrm{N}=100$
$\mathrm{N}=200$
$\overline{\mathrm{R}}{ }_{i}$
10
100
200
$\sigma_{i}$
N
10
100
200
$\rho_{\mathrm{ik}}$ or $\sigma_{\mathrm{ik}} \frac{\mathrm{N}(\mathrm{N}-1)}{2} \quad \frac{90}{2} \quad \frac{9,900}{2} \quad \frac{39,800}{2}$

Analysts may be able to estimate $\bar{R}_{i}$ and $\sigma_{i}$, but covariance comes from models. Thus, index models were developed to estimate covariances. Many more uses have been developed for them and they have become very important.

## Index Models

A. Single index $=$ splits return into unique and systematic.

## Unique Part

$$
\mathrm{R}_{\mathrm{it}}=\alpha_{\mathrm{i}}+\beta_{\mathrm{i}} \mathrm{R}_{\mathrm{mt}}+\varepsilon_{\mathrm{it}}
$$

## Systematic Part

## Where

(1) $E\left(\varepsilon_{i}\right)=0$
(2) $E\left(\varepsilon_{i} \varepsilon{ }_{j}\right)=0$
all $\mathbf{i}$ and $\mathbf{j}$
(3) $\mathrm{E}\left(\mathrm{R}_{\mathrm{m}}^{\mathrm{i}} \mathrm{\varepsilon}_{\mathrm{i}}\right)=0$
(4) $\alpha_{i}$ and $\beta_{i}$ are constants

## 1. Expected Value

$$
\begin{aligned}
\bar{R}_{i}=E\left(R_{i}\right) & =E\left(\alpha_{i}+\beta_{i} R_{m t}+\varepsilon_{i t}\right) \\
& =E\left(\alpha_{i}\right)+E\left(\beta_{i} R_{m t}\right)+E\left(\varepsilon_{i t}\right) \\
& =\alpha_{i}+\beta_{i} \bar{R}_{m}
\end{aligned}
$$

## 2. Variance

$$
\begin{aligned}
& \sigma_{\mathrm{i}}^{2}=\mathrm{E}\left(\mathrm{R}_{\mathrm{it}}-\overline{\mathrm{R}}_{\mathrm{i}}\right)=\mathrm{E}\left[\left(\alpha_{\mathrm{i}}+\beta_{\mathrm{i}} \mathrm{R}_{\mathrm{mt}}+\varepsilon_{\mathrm{it}}\right)-\left(\alpha_{\mathrm{i}}+\beta_{\mathrm{i}} \overline{\mathrm{R}}_{\mathrm{m}}\right)\right]^{2} \\
& =\mathrm{E}\left[\beta_{\mathrm{i}}\left(\mathrm{R}_{\mathrm{mt}}-\overline{\mathrm{R}}_{\mathrm{m}}\right)+\varepsilon_{\mathrm{it}}\right]^{2} \\
& \sigma_{\mathrm{i}}^{2}=\mathrm{E}\left[\beta_{\mathrm{i}}^{2}\left(\mathrm{R}_{\mathrm{mt}}-\overline{\mathrm{R}}_{\mathrm{m}}\right)^{2}+2 \beta_{\mathrm{i}} \varepsilon_{\mathrm{it}}\left(\mathrm{R}_{\mathrm{mt}}-\overline{\mathrm{R}}_{\mathrm{m}}\right)+\varepsilon_{\mathrm{it}}^{2}\right]
\end{aligned}
$$

$$
=\beta_{\mathrm{i}}^{2} \mathrm{E}\left(\mathrm{R}_{\mathrm{mt}}-\overline{\mathrm{R}}_{\mathrm{m}}\right)^{2}+\mathrm{E}\left(\varepsilon_{\mathrm{i}}\right)^{2}
$$

3. Covariance

$$
\sigma_{i j}=E\left[\left(R_{i t}-\bar{R}_{i}\right)\left(R_{j t}-\bar{R}_{j}\right)\right]
$$

$$
=E\left[\left(\alpha_{i}+\beta_{i} R_{m t}+\varepsilon_{i t}\right)-\left(\alpha_{i}+\beta_{i} \overline{\mathrm{R}}_{\mathrm{mt}}\right)\right]
$$

$$
\left[\left(\alpha_{j}+\beta_{j} R_{m t}+\varepsilon_{j t}\right)-\left(\alpha_{j}+\beta_{j} \bar{R}_{m t}\right)\right]
$$

$$
=E\left[\left(\beta_{i}\left(R_{m t}-\bar{R}_{m}\right)+\varepsilon_{i t}\right)\left(\beta_{j}\left(R_{m t}-\bar{R}_{m}\right)+\varepsilon_{j t}\right)\right]
$$

$$
\begin{aligned}
& =\beta_{i} \beta_{j} E\left(R_{m t}-\bar{R}_{m}\right)^{2}+\beta_{j} E\left[\left(R_{m t}-\bar{R}_{m}\right) \varepsilon_{i t}\right] \\
& +\beta_{i} E\left[\left(R_{m t}-\bar{R}_{m} \varepsilon_{j t}\right]+E\left(\varepsilon_{i} \varepsilon_{j}\right)\right. \\
& =\beta_{i} \beta_{j} \sigma_{R_{m}}^{2}+E\left(\varepsilon_{i} \varepsilon_{j}\right)
\end{aligned}
$$

| Date | Observe |  |
| :--- | :---: | :---: | \(\left.\begin{array}{r}Return <br>

Market\end{array}\right\}\)

Good Fairy says Beta $=1$

| Return Gm | (constant) + $\qquad$ $\alpha_{i}$ $\qquad$ | Beta (return market) $\qquad$ $\beta_{i} \mathrm{R}_{\mathrm{mt}}{ }^{+}$ $\qquad$ | $\underline{\varepsilon_{i t}}$ |
| :---: | :---: | :---: | :---: |
| 11 | 2 | 5 | 4 |
| 3 | 2 | 4 | -3 |
| 7 | 2 | 7 | -2 |
| 0 | 2 | -2 | 0 |
| 6 | 2 | 4 | 0 |
| $\frac{9}{36}$ | $\frac{2}{12}$ | $\frac{6}{24}$ | $\frac{+1}{0}$ |

## Did the Good Fairy lie???

## Assume $\beta=1.5$

| Return <br> GM | $\alpha_{i}$ | $\beta_{i} R_{m}$ | $+\varepsilon_{i}$ |
| :--- | :--- | :--- | :--- |
| 11 | 0 | 7.5 | -3.5 |
| 3 | 0 | 6 | -3 |
| 7 | 0 | 10.5 | -3.5 |
| 0 | 0 | -3 | 3 |
| 6 | 0 | 6 | 0 |
| $\frac{9}{36}$ | $\frac{0}{0}$ | $\frac{9}{36}$ | $\frac{0}{0}$ |

Covariance with market

$$
\begin{array}{ll}
\frac{\beta=1}{1 * 4=4} & \frac{\beta=1.5}{1 * 3.5=3.5} \\
0 *-3=0 & 0 *-3=0 \\
3 *-2=-6 & 3 *-3.5=-10.5 \\
-6 * 0=0 & -6 * 3=-18 \\
0 * 0=0 & 0 * 0=0 \\
\frac{2 * 1=2}{0} & \frac{2 * 0=0}{-25}
\end{array}
$$

Note:
(1). Mean return unchanged

$$
\begin{aligned}
& \overline{\mathrm{R}}_{\mathrm{i}}=\alpha_{i}+\beta_{i} \overline{\mathrm{R}}_{\mathrm{m}} \\
& 6=2+1 \cdot 4
\end{aligned}
$$

(2). Standard deviation unchanged

$$
\begin{aligned}
& \sigma_{\mathrm{i}}^{2}=\beta{ }_{\mathrm{i}}^{2} \sigma_{\mathrm{m}}^{2}+\sigma_{\mathrm{i}}^{2} \\
& 131 / 3=1^{2} \cdot 81 / 3+5
\end{aligned}
$$

(3). Only change is covariance
$E\left(\varepsilon_{i} \varepsilon_{j}\right)=0$
True only by assumption.
Economic content is only reason securities move together is common response to market movements.

## Consider Ford

| Month | Return | $\alpha_{i}$ | $\beta_{i} \mathrm{R}_{\mathrm{m}}$ | $\varepsilon_{\mathrm{i}}$ |
| :--- | :--- | :--- | :--- | :--- |
| Dec | 9.5 | 2 | 7.5 | -1 |
| Nov | 7.5 | 2 | 6 | -.5 |
| Oct | 9.5 | 2 | 10.5 | -3 |
| Sept | -2 | 2 | -3 | -1 |
| Aug | 11 | 2 | 6 | +3 |
| July | $\frac{12.5}{48}$ | $\frac{2}{12}$ | $\frac{9}{36}$ | $\frac{1.5}{0}$ |

$$
\begin{aligned}
& 5 \text { * } 1.5=7.5 \\
& -3^{*}-.5=+1.5 \\
& 1 \text { * } 1.5=1.5 \\
& -6 \text { * }-10=60 \\
& 0 \text { * } 3=0 \\
& 3 * 4.5=\frac{13.5}{84} \\
& 0 * 3=0 \\
& 1 * 1.5=\frac{1.5}{9} \\
& \operatorname{cov}\left(R_{i} R_{j}\right)=\beta_{i} \beta_{j} \sigma_{m}^{2}+E\left(\varepsilon_{i} \varepsilon_{j}\right) \\
& =1 \cdot(1.5)(81 / 3)+1.5 \\
& 14=121 / 2+1.5
\end{aligned}
$$

## Adjusting Beta

| True | .6 | .8 | observe <br> 1.0 | 1.2 | 1.4 |
| :--- | ---: | ---: | ---: | ---: | ---: |
| 1.2 |  |  | .2 | .6 | .2 |
| 1.0 |  | .2 | .6 | .2 |  |
| .8 | .2 | .6 | .2 |  |  |

Extremely high betas likely upper
Tail of a true lower beta thus
(1) Improve if adjust to mean
(2) Improve if use information about company

## Some Adjustment Techniques

I. Blumes

Companies
1985-1990
1990-1995

1

$\beta_{21}$

2

$\beta_{22}$

3

$\beta_{23}$

Get adjustment from first to second period

$$
\beta_{i 2}=a+b \beta_{i 1}+e_{i}
$$

This shows normal adjustment to forecast for 19952000.

$$
\beta_{\mathrm{adj}}=\hat{\mathrm{a}}+\overline{\mathrm{b}} \beta_{\mathrm{i} 2}
$$

## II. Vasichek

$$
\beta_{\mathrm{adj}}=\frac{\sigma_{\beta_{\mathrm{i} 1}^{2}}^{\sigma_{\beta_{1}}^{2}+\sigma_{\beta}^{2}} \bar{\beta}_{\mathrm{i} 1}}{\sigma_{\bar{\beta}_{1}}^{2}+\sigma_{\beta_{\mathrm{i} 1}^{2}}^{2}} \beta_{\mathrm{il}}
$$

Puts more weight on mean if imperfect estimates of Beta.
III. Fundamentals

$$
\beta_{\mathrm{adj}}=\beta_{\mathrm{hist}}+\text { fundamentals }
$$

a. Dividend Payment
b. Growth

+ firm variables

