## SIMPLIFY INPUTS --

## SINGLE INDEX CASE

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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		N = 10	N = 100	N = 200
$\overline{R}_{i}$	Ν	10	100	200
$s_{i}$	Ν	10	100	200
$r_{ik}$ or $s_{ik}$	<u>N (N-1)</u>	<u>90</u>	<u>9,900</u>	<u>39,800</u>
	2	2	2	2

Analysts may be able to estimate  $\overline{R}_i$  and  $s_i$ , 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**

$$R_{it} = a_i + b_i R_{mt} + e_{it}$$

**Systematic Part** 

Where

- (1)  $E(e_i) = 0$ (2)  $E(e_i e_j) = 0$  all i and j (3)  $E(R_m e_i) = 0$
- (4)  $\boldsymbol{a}_{i}$  and  $\boldsymbol{b}_{i}$  are constants

### 1. Expected Value

$$\overline{R}_{i} = E\left(R_{i}\right) = E\left(a_{i} + b_{i}R_{mt} + e_{it}\right)$$
$$= E\left(a_{i}\right) + E\left(b_{i}R_{mt}\right) + E\left(e_{it}\right)$$

$$= a_{i} + b_{i} \overline{R}_{m}$$

#### 2. Variance

$$s_{i}^{2} = E\left(R_{it} - \overline{R}_{i}\right) = E\left[\left(a_{i} + b_{i}R_{mt} + e_{it}\right) - \left(a_{i} + b_{i}\overline{R}_{m}\right)\right]^{2}$$
$$= E\left[b_{i}\left(R_{mt} - \overline{R}_{m}\right) + e_{it}\right]^{2}$$
$$s_{i}^{2} = E\left[b_{i}^{2}\left(R_{mt} - \overline{R}_{m}\right)^{2} + 2b_{i}e_{it}\left(R_{mt} - \overline{R}_{m}\right) + e_{it}^{2}\right]$$
$$= b^{2}E\left[R_{mt} - \overline{R}_{m}\right]^{2} + E\left[e_{i}\right]^{2}$$

$$=\boldsymbol{b}_{i}^{2} \mathbf{E} \left(\mathbf{R}_{mt} - \overline{\mathbf{R}}_{m}\right)^{2} + \mathbf{E} \left(\boldsymbol{e}_{i}\right)^{2}$$

#### 3. Covariance

$$\mathbf{s}_{ij} = E\left[\left(R_{it} - \overline{R}_{i}\right)\left(R_{jt} - \overline{R}_{j}\right)\right]$$

$$= \mathbf{E}\left[\left(\mathbf{a}_{i} + \mathbf{b}_{i}\mathbf{R}_{mt} + \mathbf{e}_{it}\right) - \left(\mathbf{a}_{i} + \mathbf{b}_{i}\overline{\mathbf{R}}_{mt}\right)\right]$$

$$\left[\left(\boldsymbol{a}_{j}+\boldsymbol{b}_{j}R_{mt}+\boldsymbol{e}_{jt}\right)-\left(\boldsymbol{a}_{j}+\boldsymbol{b}_{j}\overline{R}_{mt}\right)\right]$$

$$= E\left[\left(\boldsymbol{b}_{i}\left(\boldsymbol{R}_{mt} - \overline{\boldsymbol{R}}_{m}\right) + \boldsymbol{e}_{it}\left(\boldsymbol{b}_{j}\left(\boldsymbol{R}_{mt} - \overline{\boldsymbol{R}}_{m}\right) + \boldsymbol{e}_{jt}\right)\right]$$

$$= \boldsymbol{b}_{i} \boldsymbol{b}_{j} \boldsymbol{E} \left( \boldsymbol{R}_{mt} - \overline{\boldsymbol{R}}_{m} \right)^{2} + \boldsymbol{b}_{j} \boldsymbol{E} \left[ \left( \boldsymbol{R}_{mt} - \overline{\boldsymbol{R}}_{m} \right) \boldsymbol{e}_{it} \right]$$

$$+ \boldsymbol{b}_{i} \mathbf{E} \left[ \left( \mathbf{R}_{mt} - \overline{\mathbf{R}}_{m} \right) \boldsymbol{e}_{jt} \right] + \mathbf{E} \left( \boldsymbol{e}_{i} \boldsymbol{e}_{j} \right)$$

$$= \boldsymbol{b}_{i} \boldsymbol{b}_{j} \boldsymbol{s}_{R}^{2} + E \left( \boldsymbol{e}_{i} \boldsymbol{e}_{j} \right)$$

	Observe			
Date	Return GM	Return Market		
Dec	11	5		
Nov	3	4		
Oct	7	7		
Sept	0	-2		
Aug	6	4		
July	<u>9</u> 36	<u>6</u> 24		

Good Fairy says Beta = 1

Return Gm	(constant) +	Beta (return market)	0
0111	<u>a</u>	$\underline{b_i^{R}}$ mt <sup>+</sup>	<u>- e<sub>it</sub></u>
11	2	5	4
3	2	4	-3
7	2	7	-2
0	2	-2	0
6	2	4	0
<u>9</u> 36	<u>2</u> 12	<u>6</u> 24	<u>+1</u> 0

Did the Good Fairy <u>lie</u>???

## Assume b = 1.5

Return

GM	a	<b>b</b> _R_m	+ <i>e</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
<u>9</u> 36	<u>0</u> 0	<u>9</u> 36	<u>0</u> 0

**Covariance with market** 

$\underline{\boldsymbol{b}=1}$	<b>b</b> =1.5
1 * 4 = 4	1 * 3.5 = 3.5
0 * -3 = 0	0 * -3 = 0
3 * -2 = -6	3 * -3.5 = -10.5
-6 * 0 = <b>0</b>	-6 * 3 = -18
0 * 0 = 0	0 * 0 = 0
$\frac{2 * 1 = 2}{0}$	<u>2 * 0 = 0</u> -25

$$\boldsymbol{b}_{i} = \frac{\operatorname{cov}\left(\operatorname{R}_{i}\operatorname{R}_{m}\right)}{\operatorname{Var}\left(\operatorname{R}_{m}\right)}$$

<u>Note</u>:

(1). Mean return unchanged

$$\overline{R}_{i} = a_{i} + b_{i}\overline{R}_{m}$$
  
6=2+1.4

(2). Standard deviation unchanged

$$s_i^2 = b_i^2 s_m^2 + s_{e_i}^2$$
  
 $13\frac{1}{3} = 1^2 \cdot 8\frac{1}{3} + 5$ 

(3). Only change is covariance

$$E\left(e_{i}e_{j}\right)=0$$

True <u>only</u> by assumption.

Economic content is only reason securities move together is common response to market movements.

	Consider Ford				
<u>Month</u>	Return	<u>a</u>	<b>b</b> _R	<b>e</b> 1	
Dec	9.5	2	7.5	0	
Nov	7.5	2	6	5	
Oct	9.5	2	10.5	-3	
Sept	-2	2	-3	-1	
Aug	11	2	6	+3	
July	<u>12.5</u> 48	<u>2</u> 12	<u>9</u> 36	<u>1.5</u> 0	

$$\frac{\text{cov}\left(R,R,i\right)}{5*1.5=7.5} \qquad E\left(e,e,i\right) \\ 4*0=0 \\ -3*-.5=+1.5 \\ 1*1.5=1.5 \\ -6*-10=60 \\ 0*3=0 \\ 3*4.5=\frac{13.5}{84} \\ 1*1.5=\frac{1.5}{9}$$

$$\operatorname{cov}\left(\operatorname{R}_{i}\operatorname{R}_{j}\right) = \boldsymbol{b}_{i}\boldsymbol{b}_{j}\boldsymbol{s}_{m}^{2} + \operatorname{E}\left(\boldsymbol{e}_{i}\boldsymbol{e}_{j}\right)$$
$$= 1 \cdot (1.5)(8\frac{1}{3}) + 1.5$$
$$14 = 12\frac{1}{2} + 1.5$$

#### Adjusting Beta

<u>True</u>	.6	.8	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
 
$$\boldsymbol{b}_{11}$$
 $\boldsymbol{b}_{21}$ 

 2
  $\boldsymbol{b}_{12}$ 
 $\boldsymbol{b}_{22}$ 

 3
  $\boldsymbol{b}_{13}$ 
 $\boldsymbol{b}_{23}$ 

Get adjustment from first to second period

$$\boldsymbol{b}_{i2} = a + b \boldsymbol{b}_{i1} + e_i$$

This shows normal adjustment to forecast for 1995-2000.

$$\boldsymbol{b}_{adj} = \hat{a} + \hat{b} \boldsymbol{b}_{i2}$$

#### II. Vasichek



Puts more weight on mean if imperfect estimates of Beta.

**III.** Fundamentals

$$\boldsymbol{b}_{adj} = \boldsymbol{b}_{hist}$$
 + fundamentals

- a. Dividend Payment
- b. Growth

+ firm variables