

2.1

a.

x	y	$x - \bar{x}$	$(x - \bar{x})^2$	$y - \bar{y}$	$(x - \bar{x})(y - \bar{y})$
3	4	2	4	2	4
2	2	1	1	0	0
1	3	0	0	1	0
-1	1	-2	4	-1	2
0	0	-1	1	-2	2
$\sum x_i = 5$	$\sum y_i = 10$	$\sum (x_i - \bar{x}) = 0$	$\sum (x_i - \bar{x})^2 = 10$	$\sum (y_i - \bar{y}) = 0$	$\sum (x_i - \bar{x})(y_i - \bar{y}) = 8$

$$\bar{x} = 1, \bar{y} = 2 \quad \#$$

$$b_2 = \frac{\sum (x_i - \bar{x})(y_i - \bar{y})}{\sum (x_i - \bar{x})^2} = \frac{8}{10} = 0.8$$

$$b_1 = \bar{y} - b_2 \bar{x} = 2 - 0.8 \times 1 = 1.2 \quad \#$$

$$c_1 \quad \begin{aligned} \sum x_i^2 &= 3^2 + 2^2 + 1^2 + (-1)^2 + 0^2 = 15 \\ \sum x_i y_i &= 12 + 4 + 3 - 1 + 0 = 18 \end{aligned}$$

$$\begin{cases} \sum (x_i - \bar{x})^2 = 10 \\ \sum x_i^2 - N \bar{x}^2 = 15 - 5 \times 1^2 = 10 \end{cases}$$

$$\begin{cases} \sum (x_i - \bar{x})(y_i - \bar{y}) = 8 \\ \sum x_i y_i - N \bar{x} \bar{y} = 18 - 5 \times 1 \times 2 = 8 \end{cases}$$

d.

x_i	y_i	\hat{y}_i	\hat{e}_i	\hat{e}_i^2	$x_i \hat{e}_i$
3	4	3.6	0.4	0.16	1.2
2	2	2.8	-0.8	0.64	-4.6
1	3	2	1	1	1
-1	1	0.4	0.6	0.36	-0.6
0	0	1.2	-1.2	1.44	0
$\sum x_i = 5$	$\sum y_i = 10$	$\sum \hat{y}_i = 10$	$\sum \hat{e}_i = 0$	$\sum \hat{e}_i^2 = 3.6$	$\sum x_i \hat{e}_i = 0$

$$s_y^2 = \frac{\sum (y_i - \bar{y})^2}{N-1} = \frac{4+0+1+1+4}{4} = 2.5$$

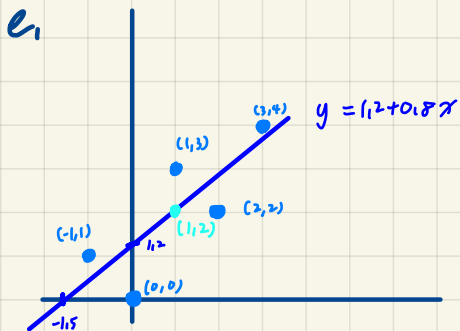
$$s_x^2 = \frac{\sum (x_i - \bar{x})^2}{N-1} = \frac{10}{4} = 2.5$$

$$s_{xy} = \frac{\sum (x_i - \bar{x})(y_i - \bar{y})}{n-1} = \frac{8}{4} = 2$$

$$r_{xy} = \frac{2}{\sqrt{2.5} \cdot \sqrt{2.5}} = 0.8$$

$$CV_x = 100 \times \frac{\sqrt{2.5}}{1} = 158.1139$$

$$x_{0.5} = 1$$



f. $(\bar{x}, \bar{y}) = (1, 2)$, the fitted line passes through the point.

g. $b_1 + b_2 \bar{x} = 1.2 + 0.8 \times 1 = 2 = \bar{y}$

h. $\bar{y} = \frac{\sum \hat{y}_i}{n} = \frac{10}{5} = 2 = \bar{y}$

i. $\hat{\sigma}^2 = \text{Var}(e_i) = \frac{3.6}{5-2} = 1.2$

j. $\hat{\text{Var}}(b_2 | X) = \frac{\hat{\sigma}^2}{\sum (x_i - \bar{x})^2} = \frac{1.2}{10} = 0.12$

se. $(b_2) = \sqrt{0.12} \approx 0.3464$

2.14

a. $\bar{y}_2 = 19.74$

$$\bar{x}_2 = (19.74 + 4.88) / 1.8$$

$$\varepsilon = b_2 \times \frac{\bar{x}_2}{\bar{y}_2} = \frac{19.74 + 4.88}{19.74} = 1.2472$$

b. $\bar{x}_1 = 13.68$

$$\bar{y}_1 = 13.68 \times 2.46 - 10.76$$

$$\text{se}(\varepsilon) = \sqrt{\text{Var}(\varepsilon)} = \sqrt{\text{Var}(b_2 \times \frac{\bar{x}_1}{\bar{y}_1})} = \frac{\bar{x}_1}{\bar{y}_1} \cdot \text{se}(b_2) = 0.0956$$

c. Urban: $12y: -10.76 + 2.46 \times 12 = 18.76$

$$16y: -10.76 + 2.46 \times 16 = 28.6$$

Rural: $12y: -4.88 + 1.8 \times 12 = 16.72$

$$16y: -4.88 + 1.8 \times 16 = 23.92$$

2.16

a. $r_j - r_f = \alpha_j + \beta_j(r_m - r_f) + e_j$

$r_j - r_f$ can be seen as dependent variable y

$r_m - r_f$ can be seen as independent variable x

$\Rightarrow y = \alpha + \beta x + e$

\therefore CAPM is a simple regression

b.

```
Company: ge
Intercept: -0.0009586682
Slope: 1.147952

Call:
lm(formula = y ~ x)

Residuals:
    Min       1Q   Median       3Q      Max
-0.173801 -0.033907 -0.003789  0.038858  0.202748

Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept) -0.0009587  0.0044244  -0.217    0.829
x            1.1479521  0.0895394   12.821 <2e-16 ***
```

$r_{ge} - r_f = -0.0010 + 1.1480(r_m - r_f)$

```
Company: ibm
Intercept: 0.00605255
Slope: 0.9768898

Call:
lm(formula = y ~ x)

Residuals:
    Min       1Q   Median       3Q      Max
-0.254880 -0.035266 -0.005322  0.031490  0.274520

Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept)  0.006053  0.004834   1.252    0.212
x            0.976890  0.097839   9.985 <2e-16 ***
```

$r_{ibm} - r_f = 0.0060 + 0.9769(r_m - r_f)$

```
Company: ford
Intercept: 0.003778911
Slope: 1.662031

Call:
lm(formula = y ~ x)

Residuals:
    Min       1Q   Median       3Q      Max
-0.27315 -0.07875 -0.01198  0.04720  1.08874

Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept)  0.003779  0.010225   0.370    0.712
x            1.662031  0.206937   8.032 1.27e-13 ***
```

$r_{ford} - r_f = 0.0038 + 1.6620(r_m - r_f)$

```
Company: msft
Intercept: 0.003249601
Slope: 1.20184

Call:
lm(formula = y ~ x)

Residuals:
    Min       1Q   Median       3Q      Max
-0.27424 -0.04744 -0.00820  0.03869  0.35801

Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept)  0.003250  0.006036   0.538    0.591
x            1.201840  0.122152   9.839 <2e-16 ***
```

$r_{msft} - r_f = 0.0032 + 1.2018(r_m - r_f)$

```
Company: dis
Intercept: 0.001046924
Slope: 1.011521

Call:
lm(formula = y ~ x)

Residuals:
    Min       1Q   Median       3Q      Max
-0.176233 -0.030021 -0.004232  0.029179  0.280528

Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept)  0.001047  0.004676   0.224    0.823
x            1.011521  0.094638   10.688 <2e-16 ***
```

$r_{dis} - r_f = 0.0010 + 1.0115(r_m - r_f)$

```
Company: xom
Intercept: 0.005283533
Slope: 0.4565208

Call:
lm(formula = y ~ x)

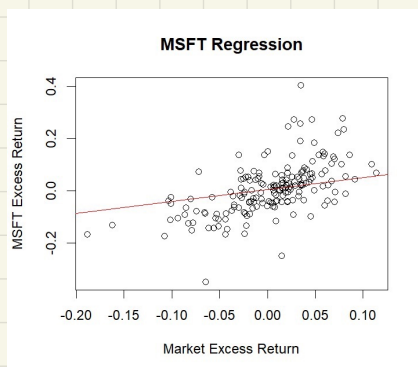
Residuals:
    Min       1Q   Median       3Q      Max
-0.114474 -0.030596 -0.001884  0.026849  0.215396

Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept)  0.005284  0.003535   1.494    0.137
x            0.456521  0.071550   6.380 1.48e-09 ***
```

$r_{xom} - r_f = 0.0053 + 0.4565(r_m - r_f)$

Ford is the most aggressive
Exxon-Mobile is the most defensive

c. Yes, all of these firm's intercept parameter show α_j is not significant (not to reject $\alpha_j = 0$)



Intercept no intercept

1.1480 →

```
Company: ge
Slope: 1.146763
```

0.9769 →

```
Company: ibm
Slope: 0.9843954
```

1.6620 →

```
Company: ford
Slope: 1.666717
```

1.2018 →

```
Company: msft
Slope: 1.205869
```

1.0115 →

```
Company: dis
Slope: 1.012819
```

0.4565 →

```
Company: xom
Slope: 0.4630727
```

d. Do not change much