

2.11.1

1a)

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

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

1b)

$$b_2 = \frac{8}{10} = 0.8$$

$$b_1 = 2 - 0.8 \times 1 = 1.2$$

\Rightarrow fitted regression line

$$= \hat{y}_i = 1.2 + 0.8x_i$$

0.8 is the slope, 1.2 is the intercept.

1c)

$$\sum_{i=1}^5 x_i^2 = 9 + 4 + 1 + 1 + 0 = 15$$

$$\sum_{i=1}^5 x_i y_i = 12 + 4 + 3 - 1 + 0 = 18$$

$$\sum_{i=1}^5 (x_i - \bar{x})^2 = \sum_{i=1}^5 x_i^2 - N\bar{x}^2 = 15 - 5 \cdot 1^2 = 10$$

$$\sum_{i=1}^5 x_i y_i - N\bar{x}\bar{y} = 18 - 5 \times 1 \times 2 = 8 = \sum_{i=1}^5 (x_i - \bar{x})(y_i - \bar{y})$$

1d)

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	-1.6
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
$\sum x_i = 5$	$\sum y_i = 10$	$\sum \hat{y}_i = 10$	$\sum \hat{e}_i = 0$	$\sum \hat{e}_i^2 = 3.6$	$\sum x_i \hat{e}_i = 0$

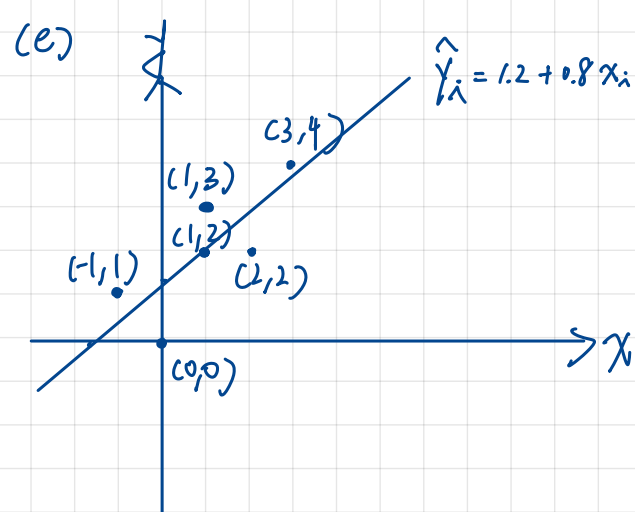
$$s_y^2 = \frac{2^2 + 0^2 + 1^2 + (-2)^2 + (-2)^2}{5-1} = 2.5$$

$$s_x^2 = \frac{2^2 + 1^2 + 0^2 + (-2)^2 + (-1)^2}{5-1} = 2.5$$

$$s_{xy} = \frac{2 \times 2 + 0 \times 1 + 1 \times 0 + (-1) \times (-2) + (-2) \times (-1)}{5-1} = 2$$

$$r_{xy} = \frac{2}{\sqrt{2.5} \cdot \sqrt{2.5}} = 0.8, CV_x = 100 \cdot \frac{\sqrt{2.5}}{1} = 100\sqrt{2.5}$$

Median $x=1$, 50th $x=1$



(f) Yes, it goes through $(1, 2) = (\bar{X}, \bar{Y})$

(g) $\bar{X} = 1$, $\bar{Y} = 2$, $\bar{Y} = 1.2 + 0.8 \cdot (1) = 2$

(h) $\hat{Y} = \frac{\sum \hat{Y}_i}{N} = 2 = \bar{Y}$

(i) $\hat{\sigma}^2 = \frac{\sum \hat{\epsilon}_i^2}{N} = \frac{3.6}{5} = 0.72$

(j) $\widehat{\text{var}}(b_1 | X) = \frac{\hat{\sigma}^2}{\sum (X_i - \bar{X})^2} = \frac{0.72}{10} = 0.072$, $\text{se}(b_2) = \sqrt{0.072} = 0.2683$

2.1.14

(a) $\widehat{\text{WAGE}} = -4.88 + 1.8 \text{EDUC}$
 $\widehat{\text{WAGE}} = 19.74 \Rightarrow \Sigma = 1.8 \times \frac{(19.74 + 4.88) + 1.8}{19.74} = 1.2472$

(b) $b_1 = 2.46$, $\text{EDUC} = 13.68$, $\widehat{\text{WAGE}} = 19.74$, $\text{se}(b_1) = 0.16$

$\Rightarrow \text{se}(\hat{\epsilon}) = 0.16 \times \frac{13.68}{19.74} = 0.11$

(c) ① $\text{EDUC} = 12$ $\begin{cases} \text{Urban} = -10.76 + 2.46 \times 12 = 18.76 \\ \text{Rural} = -4.88 + 1.8 \times 12 = 16.72 \end{cases}$

② $\text{EDUC} = 16$ $\begin{cases} \text{Urban} = -10.76 + 2.46 \times 16 = 28.6 \\ \text{Rural} = -4.88 + 1.8 \times 16 = 23.92 \end{cases}$

2.1.16

(a) The econometric model is a simple linear regression because it follows $y = \alpha + \beta x + e$, the dependent variable is $(r_j - r_f)$, independent variable is $(r_m - r_f)$

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

(b)

```
cam: ge
  beta= 1.147952
cam: ibm
  beta= 0.976890
cam: ford
  beta= 1.662031
cam: msft
  beta= 1.201840
cam: dis
  beta= 1.011521
cam: xom
  beta= 0.456521
```

most aggressive: Ford (最高Beta)

most defensive: xom (最低Beta)

(c)

```
> summary(capm_model1)

Call:
lm(formula = rj_c ~ rm_c)

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
rm_c         1.201840   0.122152   9.839 <2e-16 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.08083 on 178 degrees of freedom
Multiple R-squared:  0.3523,    Adjusted R-squared:  0.3486
F-statistic: 96.8 on 1 and 178 DF, p-value: < 2.2e-16
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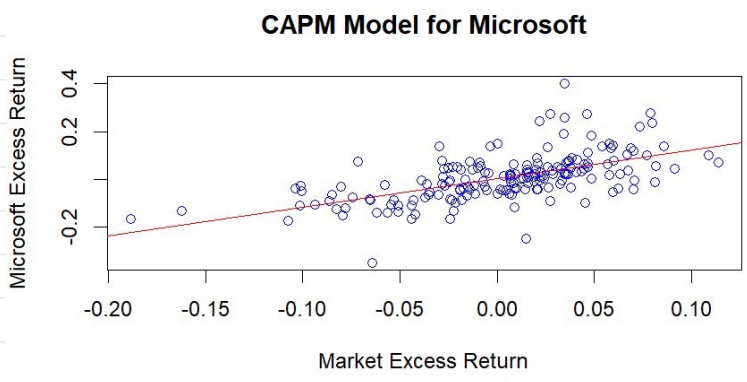
$$p = 0.591 > 0.05$$

\Rightarrow don't reject H_0

α 不显著, β 显著 > 1

\downarrow
CAPM 适用于 Msft

\downarrow
Msft 股票比市场更有 risk 和波动



(d)

```
com ge
Slpoe 1.146763
com ibm
Slpoe 0.9843954
com ford
Slpoe 1.666717
com msft
Slpoe 1.205869
com dis
Slpoe 1.012819
com xom
Slpoe 0.4630727
> |
```

$$\alpha=0 : \beta \approx 1.2059$$

$1.2018 \rightarrow 1.2059 \Rightarrow \beta$ 值变化小
 $\therefore \alpha$ 几乎没有影响
所以CAPM在msft合理