

實驗設計與分析 HW6 solution

Problem 1.

Do Problem 4 in Chapter 5 of textbook.

Prove that, for a resolution R design, its projection onto any $R - 1$ factors is a full factorial.
(Hint: If the projection is not a full factorial, it must be a fractional factorial with some defining relation among the $R - 1$ factors.)

假設一個 resolution R 實驗其投影至 $R - 1$ 個因子上並非 full factorial，則其為 fractional factorial，表示這 $R - 1$ 個因子間存在某個 defining relation L 。由於此 $R - 1$ 個因子所形成的設計矩陣必定為整個設計的子矩陣，因此 defining relation L 也必定為整個實驗的 defining relation，但因為 L 是由 $R - 1$ 個因子所構成，其 resolution 必小於 R ，和“此為 resolution R 的實驗相互矛盾”，故得證。

Problem 2.

Do Problem 13(a)(b) in Chapter 5 of textbook.

(a) Design an eight-run experiment to study the effect of the following five factors on yield: temperature (160 or 180°F), concentration (30 or 40%), catalyst (A or B), stirring rate (60 or 100 rpm), pH (low or high) with the prior knowledge that the combinations (180°F, 40%, 100 rpm) of (temperature, concentration, stirring rate) and (180°F, B, high) of (temperature, catalyst, pH) may lead to disastrous results and should be avoided. Write down the factor levels for each of the eight runs in a planning matrix.

根據題目此實驗為 2^{5-2} 設計，五個因子與其對應水準如下

Table 1: Experiment Table

		+	-
A	temperature	180	160
B	concentration	40	30
C	catalyst	A	B
D	stirring rate	100	60
E	pH	high	low

根據題目可知此實驗需避免 $(A, B, D) = (+, +, +)$ 及 $(A, C, E) = (+, -, +)$ ，因此取 defining relation

$$I = -ABD = ACE$$

其設計可先透過 A、B、C 建構全因子設計，其後 D、E 由 $D = -AB$ 及 $E = AC$ 生成，其 design matrix 及 planing matrix 如下

Table 2: Design Matrix

run	A	B	C	D	E
1	-1	-1	-1	-1	1
2	1	-1	-1	1	-1
3	-1	1	-1	1	1
4	1	1	-1	-1	-1
5	-1	-1	1	-1	-1
6	1	-1	1	1	1
7	-1	1	1	1	-1
8	1	1	1	-1	1

Table 3: Planning Matrix

run	temperature	concentration	catalyst	stirring rate	pH
1	160	30	B	60	high
2	180	30	B	100	low
3	160	40	B	100	high
4	180	40	B	60	low
5	160	30	A	60	low
6	180	30	A	100	high

run	temperature	concentration	catalyst	stirring rate	pH
7	160	40	A	100	low
8	180	40	A	60	high

(b) For the five factors in (a), find an eight-run design such that the catalyst-by-temperature interaction and the catalyst-by-concentration interaction are neither aliased with the main effects nor with each other.

根據題目所述，為使 AC 、 BC 不會與任何 main effect alias 或相互 alias，可取以下 defining relation

$$I = ABD = CDE = ABCE$$

其中 AC 、 BC 的 alias set 如下

$$\begin{aligned} AC &= BCD = ADE = BE \\ BC &= ACD = BDE = AE \end{aligned}$$

故符合題目需求。其設計可先透過 A 、 B 、 C 建構全因子設計，其後 D 、 E 由 $D = AB$ 及 $E = ABC$ 生成，其 design matrix 及 planing matrix 如下

Table 4: Design Matrix

run	A	B	C	D	E
1	-1	-1	-1	1	-1
2	1	-1	-1	-1	1
3	-1	1	-1	-1	1
4	1	1	-1	1	-1
5	-1	-1	1	1	1
6	1	-1	1	-1	-1
7	-1	1	1	-1	-1
8	1	1	1	1	1

Table 5: Planning Matrix

run	temperature	concentration	catalyst	stirring rate	pH
1	160	30	B	100	low
2	180	30	B	60	high
3	160	40	B	60	high
4	180	40	B	100	low
5	160	30	A	100	high
6	180	30	A	60	low
7	160	40	A	60	low
8	180	40	A	100	high

Problem 3.

Do Problem 15 in Chapter 5 of textbook.

(a) What is the resolution of each of the fractional factorial designs indicated below? Which design do you prefer? Justify your answers.

- (i) 2^{6-2} with $5 = 1234, 6 = 124,$
- (ii) 2^{6-2} with $5 = 123, 6 = 124,$

根據題目我們能得到

Defining contrast subgroup of (i) : $I = 12345 = 1246 = 356,$

因此 (i) 為 resolution III ,

Defining contrast subgroup of (ii) : $I = 1235 = 1246 = 3456,$

因此 (ii) 為 resolution IV , 因此 design (ii) 較 (i) 好。

(b) For the design in (ii), if we further know that any two-factor interaction involving factor 6 (i.e., 16, 26, 36, 46, 56) is negligible, which two-factor interactions are estimable under the usual assumptions that three-factor and higher interactions are negligible?

根據題目所述，首先忽略所有包含 6 的 2fi，我們能得到剩下的 2fi 有

2-factor interaction : 12, 13, 14, 15, 23, 24, 25, 34, 35,

然而因為

$$12 = 35, \quad 13 = 25, \quad 15 = 23$$

這幾項 2fi 相互 alias，因此剩下 14、24、34、45 可被估計。

(c) Under the same assumptions as in (b), find a scheme to arrange the design in (ii) in two blocks each of size 8. Explain why your choice is the best.

根據 (b) 小題的假設及 design (ii) 能發現有兩組 alias set 由 3fi 組成，分別為

$$\begin{aligned} 134 &= 245 = 236 = 156, \\ 136 &= 256 = 234 = 145, \end{aligned}$$

考慮以下兩種 blocking scheme

$$\begin{aligned} B_1 &= 134 (= 245 = 236 = 156), \\ B_2 &= 136 (= 256 = 234 = 145), \end{aligned}$$

此時所有 main effect 及 14、24、34、45 皆為 clear，且沒有 2fi 與 block confounded。

Problem 4.

Do Problem 16 in Chapter 5 of textbook.

Two choices of generators for a 2^{6-2} design are being considered:

$$\begin{aligned} A : 5 &= 1234, \quad 6 = 123, \\ B : 5 &= 123, \quad 6 = 234. \end{aligned}$$

(a) Which design, A or B, would you recommend? Why?

根據題目我們能得到

Defining contrast subgroup of A : $I = 12345 = 1236 = 456$,

因此 A 為 resolution III ,

Defining contrast subgroup of B : $I = 1235 = 2346 = 1456$,

因此 B 為 resolution IV , 因此 design B 較 A 好。

(b) Show that it is impossible to have a 2^{6-2}_V design. (Hint: Count the degrees of freedom.)

2^{6-2}_V design 共有 $2^4 = 16$ run , 因此自由度為 $16-1=15$, 若為 resolution V , 代表所有 main effect 及 2fi 皆為 clear , 然而在此情況下所需的自由度為 $6 + \binom{6}{2} = 6 + 15 = 21$, 故不可能有此設計。

Problem 5.

Do Problem 28 in Chapter 5 of textbook.

In a resistance spot welding experiment, five factors were chosen to study their effects on the tensile strength, which is the maximum load a weld can sustain in a tensile test. The five factors are: button diameter (A), welding time (B), holding time (C), electrode force (D), and machine type (E), each at two levels. The last factor is qualitative, while the others are quantitative. A 2^{5-1} design with $I = -ABCDE$ was used for the experiment. Each run has V three replicates. The data are given in Table 5.13.

Table 5.13 Design Matrix and Tensile Strength Data, Spot Welding Experiment

Run	Factor					Tensile Strength		
	A	B	C	D	E			
1	-	-	-	-	-	1330	1330	1165
2	+	+	-	-	-	1935	1935	1880
3	+	-	+	-	-	1770	1770	1770
4	-	+	+	-	-	1275	1275	1275
5	+	-	-	+	-	1880	1935	1880
6	-	+	-	+	-	1385	1440	1495
7	-	-	+	+	-	1220	1165	1440
8	+	+	+	+	-	2155	2100	2100
9	+	-	-	-	+	1715	1715	1660
10	-	+	-	-	+	1385	1550	1550
11	-	-	+	-	+	1000	1165	1495
12	+	+	+	-	+	1990	1990	1990
13	-	-	-	+	+	1275	1660	1550
14	+	+	-	+	+	1660	1605	1660
15	+	-	+	+	+	1880	1935	1935
16	-	+	+	+	+	1275	1220	1275

Figure 1: Table 5.13

(a) What effects are clear and strongly clear in this experiment?

因為 2^{5-1} design with $I = -ABCDE$ 可以得出 main effect 為 $A = -BCDE, B = -ACDE, \dots$, 都只和 4fi alias, 而 2fi 為 $AB = -CDE, AC = -BDE, \dots$ 都只和 3fi alias, 因此所有 2fi 為 clear, 所有 main effect 為 strongly clear。

(b) Analyze the location and dispersion effects separately, including the fitted models for each.

首先計算各個組合的 $\bar{y}、s^2、\ln s^2$, 其中將部分 $s^2 = 0$ 的以 0.001 代替

```
dat5 <- as.matrix(dat5)
dat5 <- cbind(unique(dat5[,-6]),matrix(dat5[,6],ncol=3,byrow=T))
dat5[which(dat5=="-",arr.ind=TRUE)]=-1
dat5[which(dat5=="+",arr.ind=TRUE)]=1
dat5 <- apply(dat5,2,as.numeric)
colnames(dat5)[1:5]=c("A","B","C","D","E")
dat5 <- as.data.frame(dat5)
ybar = apply(dat5[,6:8], 1, mean)
s2 = apply(dat5[,6:8], 1, var)
s2[s2==0]=0.001
lns2 = log(s2)
```

```

dat5_tab = cbind(dat5[,1:5],ybar,s2,lns2)
kable(dat5_tab, row.names = F, format = 'markdown',
      col.names = c("A","B","C","D","E","$\\bar{y}$","$s^2$","$\\ln s^2$"),
      caption = "Design Matrix",
      align = c(rep("r",8)),
      digits = 4)

```

Table 6: Design Matrix

A	B	C	D	E	\bar{y}	s^2	$\ln s^2$
-1	-1	-1	-1	-1	1275.000	9075.000	9.1133
1	1	-1	-1	-1	1916.667	1008.333	6.9161
1	-1	1	-1	-1	1770.000	0.001	-6.9078
-1	1	1	-1	-1	1275.000	0.001	-6.9078
1	-1	-1	1	-1	1898.333	1008.333	6.9161
-1	1	-1	1	-1	1440.000	3025.000	8.0147
-1	-1	1	1	-1	1275.000	21175.000	9.9606
1	1	1	1	-1	2118.333	1008.333	6.9161
1	-1	-1	-1	1	1696.667	1008.333	6.9161
-1	1	-1	-1	1	1495.000	9075.000	9.1133
-1	-1	1	-1	1	1220.000	63525.000	11.0592
1	1	1	-1	1	1990.000	0.001	-6.9078
-1	-1	-1	1	1	1495.000	39325.000	10.5796
1	1	-1	1	1	1641.667	1008.333	6.9161
1	-1	1	1	1	1916.667	1008.333	6.9161
-1	1	1	1	1	1256.667	1008.333	6.9161

接著對 location and disperation effect 估計

```

M = model.matrix(ybar~A*B*C*D+E,data = dat5) [,2:16]
y.effect = round(t(M)%*%ybar/8,4)
s.effect = round(t(M)%*%lns2/8,4)
effect = c("A=-BCDE", "B=-ACDE", "C=-ABDE", "D=-ABCE", "E=-ABCD",
          "AB=-CDE", "AC=-BDE", "BC=-ADE", "AD=-BCE", "BD=-ACE",
          "CD=-ABE", "AE=-BCD", "BE=-ACD", "CE=-ABD", "DE=-ABC")
tab = data.frame(effect,y.effect,s.effect)
kable(tab, row.names = F, format = 'markdown',
      col.names = c("Effect","$\\bar{y}$","$\\ln s^2$"),
      caption = "Effect Estimation Table",
      align = c("l",rep("r",2)),
      digits = 4)

```

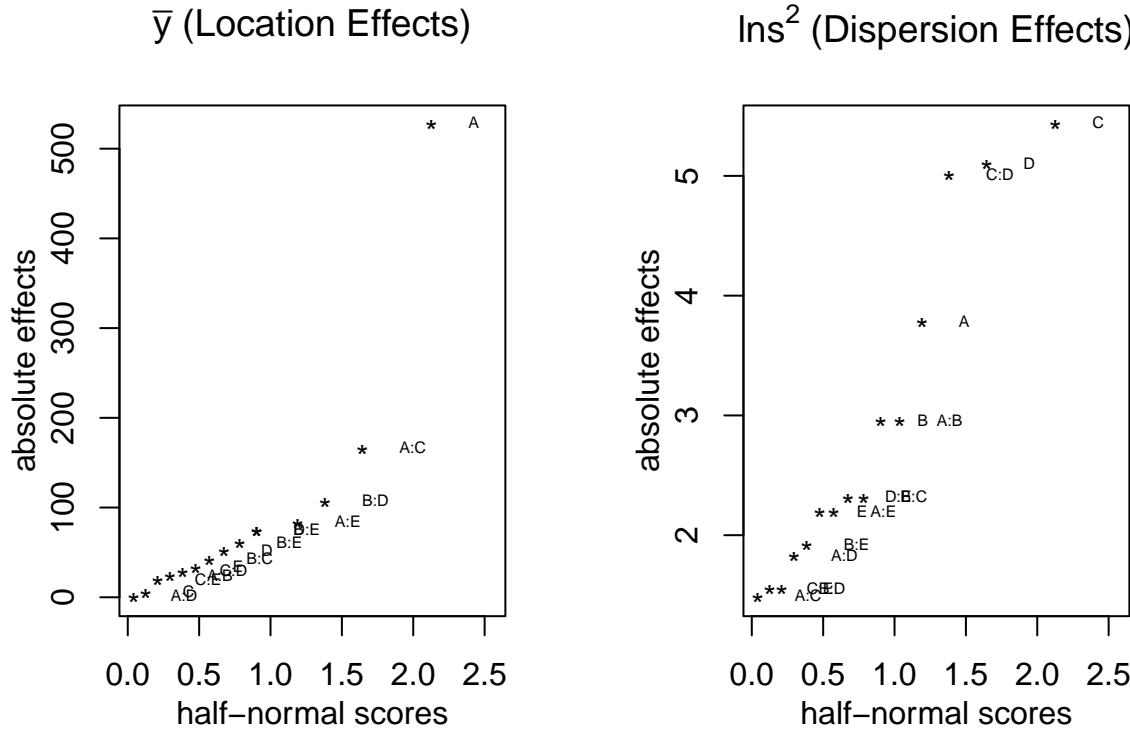
Table 7: Effect Estimation Table

Effect	\bar{y}	$\ln s^2$
A=-BCDE	527.0833	-3.7710
B=-ACDE	73.3333	-2.9471
C=-ABDE	-4.5833	-5.4300
D=-ABCE	50.4167	5.0926
E=-ABCD	-32.0833	2.1859
AB=-CDE	22.9167	2.9471
AC=-BDE	165.0000	-1.4819

Effect	\bar{y}	$\ln s^2$
BC=-ADE	41.2500	-2.3058
AD=-BCE	0.0000	1.8193
BD=-ACE	-105.4167	1.5447
CD=-ABE	27.5000	5.0006
AE=-BCD	-82.5000	-2.1859
BE=-ACD	-59.5833	-1.9113
CE=-ABD	18.3333	1.5447
DE=-ABC	-73.3333	-2.3058

由上表可知所有 2fi 為 clear，所有 main effect 為 strongly clear。接著做 half-normal plot

```
M1 = lm(ybar~A*B*C*D*E,data = dat5)
M2 = lm(lns2~A*B*C*D*E,data = dat5)
par(mfrow=c(1,2))
DanielPlot(M1,half = T,datax = F,autolab = F,
           main =TeX("\bar{y} (Location Effects)",cex.fac = .5)
DanielPlot(M2,half = T,datax = F,autolab = F,
           main =TeX("\ln s^2 (Dispersion Effects)",cex.fac = .5)
```



根據上圖能發現 location effect 中 A 為顯著效應，在 dispersion effect 中 C 、 D 、 CD 為顯著效應，因此我們能建立以下模型

$$\hat{y} = 1605 + 263.5A,$$

$$\ln \hat{s}^2 = 5.3456 - 2.715C + 2.5463D + 2.5003CD.$$

(c) For the dispersion effect model in (b), interpret the meaning of the significant effects. Use this model or an interaction effects plot to select optimal factor settings.

從 dispersion model 可以得知顯著效應有 C 、 D 、 CD ，為使實驗變異 $\ln\hat{s}^2$ 達到最小，optimal factor settings 為 $(C, D) = (+, -)$ 。

(d) Based on the location effect model in (b), select optimal factor settings in terms of maximizing the tensile strength.

從 location model 可以得知顯著效應有 A ，為使實驗的 mean \hat{y} 達到最大，optimal factor settings 為 $(A) = (+)$ 。

(e) Is there any adjustment factor? Based on the results in (c) and (d), can you devise a two-step procedure for design optimization? (Hint: Because the response is a larger-the-better characteristic, you should consider the two-step procedure in Section 6.2.)

在 two-step procedure for design optimization，(c) 對應 dispersion effect，(d) 對應 location effect，在 larger-the-better problems 中，因為 (c) 中的 factor C 及 factor D 沒有出現在 (d) 中，調整 factor C 及 factor D 時不會影響 step 1 的 location model，故 C 及 D 為 adjustment factor。

Two-step procedure for larger-the-better problems.

1. $(A) = (+)$ 實驗的 mean 達到最大， $\hat{y} = 1605 + 263.5 = 1868.5$ 。
2. $(C, D) = (+, -)$ 使實驗變異達到最小， $s^2 = \exp(5.3456 - 2.715 - 2.5463 - 2.5003) = 0.08927802$ 。