

Displays for Statistics 5303

Lecture 36

December 2, 2002

Christopher Bingham, Instructor

612-625-7023 (St. Paul)  
612-625-1024 (Minneapolis)

Class Web Page

<http://www.stat.umn.edu/~kb/classes/5303>

© 2002 by Christopher Bingham

Efficiency of a Incomplete Box Design

The efficiency is usually stated relative to a RCB design with the same number  $r$  of replicates *and the same*  $\sigma$ .

$$E_{\text{BIBD:RCB}} = g(k-1)/((g-1)k) = 1-(g-k)/((g-1)k)$$

Since number of treatments =  $g > k =$  block size,  $E_{\text{BIBD:RCB}} < 1$ .

Of course, one reason for using an incomplete block design is that  $\sigma^2$  tends to be smaller for small blocks than for large. Since  $E_{\text{BIBD:RCB}}$  assumes the same  $\sigma^2$  for a size  $k$  block as for a size  $g$  block, it may not provide a meaningful comparison of designs if the variances are very different.

Still  $E_{\text{BIBD:RCB}}$  is a useful number since it appears in a number of formulas.

$E_{\text{BIBD:RCB}}$  mainly enters through the **effective number of replicates**

$$r' = E_{\text{BIBD:RCB}} \times r < r.$$

Here are several ways  $r'$  is used.

**Estimates of  $\alpha_i$**

Let  $v_{ij} = y_{ij} - \bar{y}_{.j}$  = response with block mean removed. Then

$$\hat{\alpha}_i = v_{i.}/r' \quad ((\text{treatment sum})/r')$$

**Treatment SS**

$$SS_{\text{trt}} = r' \sum_i \hat{\alpha}_i^2$$

**Variance and standard error of a contrast**

$$V(\sum_i c_i \hat{\alpha}_i) = (\sum_i c_i^2) \sigma^2 / r'$$

$$SE(\sum_i c_i \hat{\alpha}_i) = \sqrt{\{(\sum_i c_i^2) \sigma^2 / r'\}}$$

These are like the complete block formulas with the effective replication  $r'$  in place of the actual replication  $r$ .

Analysis of BIBD Example 14.2 using MacAnova.

The analysis is virtually identical with a RCB design, although the "by hand" formulas are more complicated.

```

Cmd> tab14_1 <- read("", "exmpl14.2")
exmpl14.2      36      3
) A data set from Oehlert (2000) \emph{A First Course in Design
) and Analysis of Experiments}, New York: W. H. Freeman.
)
) Data originally from John, P. W.-M. (1961). ``An application x
) and balanced incomplete block design
) '' {\em Technometrics\}/~{\em 3}, 51--54.
)
) Table 14.1, p. 359
) Test of 9 different detergents. There are three basins that
) are used simultaneously at the same rate with a different
) detergent in each basin. Response is number of plates until
) foam disappears in a basin.
) Column 1 is session. Column 2 is treatment (kind of detergent)
) Column 3 is response (number of dishes)
) Treatments 1-4 are detergent base 1
) with (3, 2, 1, or 0) parts additive
) Treatments 5-8 are detergent base
) 2 with (3, 2, 1, or 0) parts additive
) Treatment 9 is a control.
Read from file "TP1:Stat5303:Data:OeCh14.dat"
Cmd> makecols(tab14_1, session, treatment, count)
Cmd> session <- factor(session)
Cmd> treatment <- factor(treatment)
    
```

The blocking factor is `session`. It must appear in the model before `treatment`.

```
Cmd> anova("count=session + treatment",fstat:T)
Model used is count=session+treatment
WARNING: summaries are sequential
      DF      SS      MS      F      P-value
CONSTANT 1 13572 13572 16469.69663 1.5466e-25
session  11 412.75  37.523  45.53320 6.0284e-10
treatment 8 1086.8  135.85  164.85393 6.8089e-14
ERROR1   16  13.185  0.82407
```

You can do pairwise multiple comparisons as for a CRD and CRB.

```
Cmd> pairwise("treatment",.05,hsd:T)
      4 -12.9
      3 -6.22
      2 -2.22
      8 -0.222
      1  0.333
      7  1.67
      6  3.56
      5  5.89
      9 10.1
```

You can check that standard errors for different pairwise comparisons are the same:

```
Cmd> contrast(treatment,vector(1,-1,rep(0,7)))
component: estimate
(1) 2.5556
component: ss
(1) 9.7963
component: se
(1) 0.7412

Cmd> contrast(treatment,vector(1,0,-1,rep(0,6)))
component: estimate
(1) 6.5556
component: ss
(1) 64.463
component: se
(1) 0.7412
```

### Recovery of interblock information

The incomplete randomized block model is

$$y_{ij} = \mu + \alpha_i + \beta_j + \epsilon_{ij}$$

where not all possible (i,j) occur.

Consider the block sums  $y_{.j} \equiv \sum_i y_{ij}$  where the sum is over all treatments in block j. Then

$$y_{.j} = \sum_i y_{ij} = k\mu + \sum_i n_{ij} \alpha_i + k\beta_j + \sum_i n_{ij} \epsilon_{ij} \\ = k\mu + \sum_i n_{ij} \alpha_i + \eta_j$$

where

- $n_{ij} = 1$  if treatment i is in block j
- $n_{ij} = 0$  if treatment i is not in block j
- $\eta_j \equiv k\beta_j + \sum_i n_{ij} \epsilon_{ij}$

```
Cmd> trtmeans <- tabs(count,treatment,mean:T)
Cmd> blkmeans <- tabs(count,session,mean:T)
Cmd> tabs(count,treatment,count:T) # reps of each treatment
(1) 4 4 4 4 4 4
(6) 4 4 4 4 4 4
Cmd> tabs(count,session,count:T) # sizes of each block
(1) 3 3 3 3 3 3
(6) 3 3 3 3 3 3
(11) 3 3 3 3 3 3
Cmd> r <- 4; k <- 3; b <- 12; g <- 9;
Cmd> eff <- g*(k-1)/((g-1)*k); eff
(1) 0.75
Cmd> effective_r <- r*eff; effective_r
(1) 3
```

Thus  $E_{\text{BIBD:RCB}} = .75$  and  $r' = 3$ .

Remove block means from the data.

```
Cmd> v <- count
Cmd> for(i,1,b){ # loop over blocks
v[session == i] <- v[session == i] - blkmeans[i];;}
Cmd> tabs(v,treatment,sum:T)/effective_r # alphahats
(1) 0.33333 -2.2222 -6.2222 -12.889 5.8889
(6) 3.5556 1.6667 -0.22222 10.111
Cmd> coefs(treatment) # same by black box
(1) 0.33333 -2.2222 -6.2222 -12.889 5.8889
(6) 3.5556 1.6667 -0.22222 10.111
Cmd> c <- vector(1, -1, rep(0,g-2)); c # pairwise contrast
(1) 1 -1 0 0 0 0
(6) 0 0 0 0 0 0
Cmd> mse <- SS[4]/DF[4] # estimated sigma^2
Cmd> sqrt(mse*sum(c^2)/effective_r) # standard error
(1) 0.7412
Cmd> contrast(treatment,c)$se # Standard error from black box
(1) 0.7412
```

Note that  $\sum_i n_{ij} = k$  so that  $k\mu = \sum_i n_{ij} \mu$  and

$$y_{.j} = \sum_i n_{ij} (\mu + \alpha_i) + \eta_j$$

This has the form of a multiple regression *with no constant term*

$$y_{.j} = \sum_i \tilde{\beta}_i x_{ij} + \eta_j$$

where the regression coefficients and predictor variables are

$$\tilde{\beta}_i = \mu + \alpha_i \text{ and } x_{ij} = n_{ij}$$

Provided the block effects  $\beta_j$ 's are random and independent,  $\eta_j$  are independent with constant variance

$$\tilde{\sigma}^2 = k^2 \sigma_\beta^2 + k\sigma^2.$$

When this is the case we can get estimates of  $\mu + \alpha_i$  by least squares regression with  $y_{.j}$  as response variable and  $x_{ij} = n_{ij}$  as predictor variables.

```

Cmd> blksums <- tabs(count,session,sum:T) # y_dot_j
Cmd> nij <- tabs(count,session, treatment,count:T)
Cmd> print(nij,format:"2.0f")
nij:
(1,1) 1 1 1 0 0 0 0 0 0 0 Trt 1, 2, 3 in block 1
(2,1) 0 0 0 1 1 1 0 0 0 0 Trt 4, 5, 6 in block 2
(3,1) 0 0 0 0 0 0 0 1 1 1 Trt 7, 8, 9 in block 3
(4,1) 1 0 0 1 0 0 1 0 0 1 Trt 1, 4, 7 in block 4
(5,1) 0 1 0 0 1 0 0 1 0 0 etc
(6,1) 0 0 1 0 0 1 0 0 1 0
(7,1) 1 0 0 0 1 0 0 0 1 0
(8,1) 0 1 0 0 0 1 1 0 0 0
(9,1) 0 0 1 1 0 0 0 1 0 0
(10,1) 1 0 0 0 0 1 0 1 0 0
(11,1) 0 1 0 1 0 0 0 0 1 0
(12,1) 0 0 1 0 1 0 1 0 0 0

Cmd> makecols(nij,x1,x2,x3,x4,x5,x6,x7,x8,x9)

```

Do regression without constant term by adding -1 to the model:

```

Cmd> regress("blksums=x1+x2+x3+x4+x5+x6+x7+x8+x9-1")
Model used is blksums=x1+x2+x3+x4+x5+x6+x7+x8+x9-1

```

	Coef	StdErr	t
x1	19.75	1.147	17.219
x2	15.417	1.147	13.441
x3	13.417	1.147	11.697
x4	6.4167	1.147	5.5944
x5	26.083	1.147	22.741
x6	24.083	1.147	20.997
x7	19.75	1.147	17.219
x8	19.417	1.147	16.928
x9	30.417	1.147	26.519

N: 12, MSE: 4.3056, DF: 3, R^2: 0.99969  
Regression F(9,3): 1082.4, Durbin-Watson: 0.63656  
To see the ANOVA table type 'anova()'

```

Cmd> anova("count=session + treatment",silent:T)
Cmd> coefs(treatment)
(1) 0.33333 -2.2222 -6.2222 -12.889 5.8889
(6) 3.5556 1.6667 -0.22222 10.111

```

These match column 1.

The combined est column is a linear combination which weights the estimates inversely proportional to their variance.

**Caution:** This interblock recovery works *only* when blocks are random.

- The standard errors of the inter-block estimates of treatment effects are always bigger than the intra-block estimates.
- Unless the number of blocks is large, there is unlikely to be much benefit from this procedure.

To get estimates of  $\alpha_i$  which sum to 0, you need to subtract the mean of the g coefficients from each coefficient.

```

Cmd> COEF - sum(COEF)/g # interblock estimates of alpha's

```

	x1	x2	x3	x4	x5
	x6	x7	x8	x9	
	0.33333	-4	-6	-13	6.6667
	4.6667	0.33333	-3.5527e-15	11	

Macro interblock() provides a "black box" way to get these estimates

```

Cmd> interblock(count,session,treatment)# response,block,treat

```

	intra est	intra se	inter est	inter se	inter secombined	est	combined se
	0.33333	0.49414	0.33333	0.91174	0.33333	0.43443	
	-2.2222	0.49414	-4	0.91174	-2.6259	0.43443	
	-6.2222	0.49414	-6	0.91174	-6.1718	0.43443	
	-12.889	0.49414	-13	0.91174	-12.914	0.43443	
	5.8889	0.49414	6.6667	0.91174	6.0655	0.43443	
	3.5556	0.49414	4.6667	0.91174	3.8078	0.43443	
	1.6667	0.49414	0.33333	0.91174	1.3639	0.43443	
	-0.22222	0.49414	-1.7764e-15	0.91174	-0.17177	0.43443	
	10.111	0.49414	11	0.91174	10.313	0.43443	

The intra est column has estimates as computed by coefs() following anova(). They are *intra-block* estimates because they are implicitly computed from within block differences.

The inter est column has the same values as just found, the *interblock* estimates.

Balanced incomplete blocks designs have the property

- All treatment effects  $\alpha_i$  can be estimated with equal accuracy
- All contrasts with the same  $\sum c_i^2$  have the same standard error, *always larger* than the standard error of the contrast from a RCB with r blocks and the same  $\sigma^2$ .

Essentially, the loss of efficiency is spread among all contrasts equally.

This isn't always desirable.

- Some contrasts are more important than others.
- Some contrasts may be assumed to be 0

Consider a  $2^3$  factorial with means

$$\mu_{ij} = \mu + \alpha_i + \beta_j + \gamma_k + \alpha\beta_{ij} + \alpha\gamma_{ik} + \beta\gamma_{jk} + \alpha\beta\gamma_{ijk}, \quad 1 \leq i, j, k \leq 2$$

You can express  $\mu$  and the main and interaction effects in terms of the  $\mu_{ij}$ 's.

$$\mu = (\mu_{111} + \mu_{211} + \mu_{121} + \mu_{221} + \mu_{112} + \mu_{212} + \mu_{122} + \mu_{222})/8$$

$$\alpha_2 = -\alpha_1 = (-\mu_{111} + \mu_{211} - \mu_{121} + \mu_{221} - \mu_{112} + \mu_{212} - \mu_{122} + \mu_{222})/8$$

$$\beta_2 = -\beta_1 = (-\mu_{111} - \mu_{211} + \mu_{121} + \mu_{221} - \mu_{112} - \mu_{212} + \mu_{122} + \mu_{222})/8$$

$$\alpha\beta_{22} = \alpha\beta_{11} = -\alpha\beta_{12} = -\alpha\beta_{21} = (-\mu_{111} + \mu_{211} + \mu_{121} - \mu_{221} - \mu_{112} + \mu_{212} + \mu_{122} - \mu_{222})/8$$

$$\dots$$

$$\alpha\beta\gamma_{222} = -\alpha\beta\gamma_{111} = \alpha\beta\gamma_{211} = \dots = -\alpha\beta\gamma_{221} = (-\mu_{111} + \mu_{211} + \mu_{121} - \mu_{221} + \mu_{112} - \mu_{212} - \mu_{122} + \mu_{222})/8$$

These are all (except for  $\mu$ ) contrasts in the  $g = 2^3 = 8$  treatment means  $\mu_{ijk}$ .

On p. 612-613 are several BIBD plans for  $g = 8$  treatments, with block sizes  $k = 2, 3, 4, 5, 6,$  and  $7$ . Using any of you can estimate all effects equally accurately.

But often ABC is not important, or doesn't need to be estimated at all because you assume  $\alpha\beta\gamma_{ijk} = 0$ .

Here is an incomplete block design that is not a BIB for the  $g = 8$  factorial treatments in blocks of size 4

I	II
(1)	a
ab	b
ac	c
bc	abc

Blocks I has the treatments for which the ABC contrast coefficients are -1 and block II has the treatments for which the ABC contrast coefficients are +1.

Except for the divisor 8, each effect can be found using a column of this table

	I	A	B	C	AB	AC	BC	ABC
(1)	1	-1	-1	-1	1	1	1	-1
a	1	1	-1	-1	-1	-1	1	1
b	1	-1	1	-1	-1	1	-1	1
ab	1	1	1	-1	1	-1	-1	-1
c	1	-1	-1	1	1	-1	-1	1
ac	1	1	-1	1	-1	1	-1	-1
bc	1	-1	1	1	-1	-1	1	-1
abc	1	1	1	1	1	1	1	1

or, equivalently, of this one

	I	A	B	C	AB	AC	BC	ABC
(1)	+	-	-	-	+	+	+	-
a	+	+	-	-	-	-	+	+
b	+	-	+	-	-	+	-	+
ab	+	+	+	-	+	-	-	-
c	+	-	-	+	+	-	-	+
ac	+	+	-	+	-	+	-	-
bc	+	-	+	+	-	-	+	-
abc	+	+	+	+	+	+	+	+

The estimate of  $\alpha\beta\gamma_{222}$  is

$$\hat{\alpha}\hat{\beta}\hat{\gamma}_{222} = (-y_{111} + y_{211} + y_{121} - y_{221} + y_{112} - y_{212} - y_{122} + y_{222})/8$$

$$= \alpha\beta\gamma_{222} + (\beta_1 - \beta_2)/2 + \sum c_{ijk}^{ABC} \epsilon_{ijk}$$

where  $c_{ijk}^{ABC} = \pm 1/8$

Note that this has block effects "contaminating" the interaction effect. This "contamination" is known as *confounding*.

Here the interaction is confounded with the difference between blocks.

How about the other contrasts?

Because the contrasts are orthogonal

$$\hat{\alpha}_2 = \alpha_2 + \sum c_{ijk}^A \epsilon_{ijk}$$

$$\hat{\beta}_2 = \beta_2 + \sum c_{ijk}^B \epsilon_{ijk}$$

$$\hat{\beta}\hat{\gamma}_{jk} = \beta\gamma_{jk} + \sum c_{ijk}^{BC} \epsilon_{ijk}, \text{ etc.}$$

These are unconfounded with block effects.

Because of the way the blocks were chosen, ABC is called the *defining contrast* for the design.

Any other column of the table of contrasts (except the column of all +1's) could be the defining contrast for a design with blocks of size  $2^{k-1} = 4$ . The corresponding main effect or interaction would be confounded with blocks.

Block I consists of all the treatments with -1 and block II consists of all the treatments with +1 on the defining contrast.