

12.0 FRACTIONAL FACTORIAL DESIGNS

(Updated Spring,2001)

We now need to look at a class of designs that are good for studying many variables in a relatively limited number of tests, fractional factorial designs.

Previously we saw that if we carefully selected levels for variables we could obtain a Latin square, or Graeco-Latin square design, we didn't need to look at all combinations of the variable levels. Fractional Factorial Designs, 2^{k-p} designs, are analogous to these designs.

Let's say we're thinking about a 2^3 full factorial design.

Test	I	1	2	3	12	13	23	123
1	+	-	-	-	+	+	+	-
2	+	+	-	-	-	-	+	+
3	+	-	+	-	-	+	-	+
4	+	+	+	-	+	-	-	-
5	+	-	-	+	+	-	-	+
6	+	+	-	+	-	+	-	-
7	+	-	+	+	-	-	+	-
8	+	+	+	+	+	+	+	+

We want to examine a 4th variable, but only have enough resources for 8 tests. We can introduce variable 4 thru interaction 123

Test	1	2	3	4
1	-	-	-	-
2	+	-	-	+

Test	1	2	3	4
3	-	+	-	+
4	+	+	-	-
5	-	-	+	+
6	+	-	+	-
7	-	+	+	-
8	+	+	+	+

Recipe Matrix: tells us how to run experiment

Test	I	1	2	3	4	12	13	14	23	24	34	123	124	134	234	1234
1	+	-	-	-	-	+	+	+	+	+	+	-	-	-	-	+
2	+	+	-	-	+	-	-	+	+	-	-	+	-	-	+	+
3	+	-	+	-	+	-	+	-	-	+	-	+	-	+	-	+
4	+	+	+	-	-	+	-	-	-	-	+	-	-	+	+	+
5	+	-	-	+	+	+	-	-	-	-	+	+	+	-	-	+
6	+	+	-	+	-	-	+	-	-	+	-	-	+	-	+	+
7	+	-	+	+	-	-	-	+	+	-	-	-	+	+	-	+
8	+	+	+	+	+	+	+	+	+	+	+	+	+	+	+	+

We note that the following columns are the same:

I & 1234 3 & 124 23 & 14
 1 & 234 12 & 34 123 & 4
 2 & 134 13 & 24

We have deliberately confounded Variable 4 and Interaction 123, in doing so we have also confounded I & 1234, 1 & 234, 2 & 134, etc. We have conducted a 2^{4-1} fractional factorial design 4 variables are examined in $2^{4-1} = 8$ tests.

Some Basic Ideas/Definitions

For 8 tests = $2^3 \rightarrow$ Base design Calculation Matrix is:

I	1	2	3	12	13	23	123
+	-	-	-	+	+	+	-
+	+	-	-	-	+	+	+
+	-	+	-	-	+	-	+
+	+	+	+	+	-	-	-
+	-	-	+	+	-	-	+
+	+	-	+	-	+	-	-
+	-	+	+	-	-	+	-
+	+	+	+	+	+	+	+

We introduced variable 4 thru int. 123. Symbolically, $4 = 123$

- Any column multiplied by itself produces a column of all “+” signs, I.
- $4 \cdot 4 = I$
- Any column multiplied by I gives the original column:

$$I \cdot 4 = 4$$

$$4 \cdot 4 \cdot 4 = 4$$

Consider again how we introduced variable 4 into the design.

$$4 = 123$$

This is termed the generator for the design.

$$4 \cdot 4 = 1234$$

and

$$I = 1234$$

is the alternative form for the generator.

In this case, this is the defining relationship for our 2^{4-1} design. It tells us the confounding/aliasing pattern for our design.

Question: Consider columns 123&4. How do we calculate E_{123} & E_4 , in other words, estimates of μ_{123} & μ_4 ?

Since we only have 8 tests, we only can calculate 8 unique quantities. We calculate these quantities using the 8 columns in the base design calculation matrix (I,1,2,3,12,13,23,123)

$$\begin{aligned} & \text{Tests} \\ & \sum_{i=1}^4 \text{Col}_{12} \bullet y \\ \# &= \frac{i=1}{4} = l_{12} \end{aligned}$$

These calculated values will be referred to as: $l_1, l_2, l_3, l_{12}, l_{13}, l_{23}, l_{123}$

We know what to calculate and how to do it, just like full factorials. How to interpret the results though?

We have observed that col. 123 = col. 4. When we calculate l_{123} , is this an estimate of μ_{123} or μ_4 or what? In fact, l_{123} estimates $\mu_{123} + \mu_4$. It is the effect of a linear combination of 123 & 4.

In our problem,

l_1 estimates $\mu_1 + \mu_{234}$ (shorthand notation for $\mu_1 + \mu_{234}$)

l_2 estimates $\mu_2 + \mu_{134}$

l_3 estimates $\mu_3 + \mu_{124}$

l_{12} estimates $\mu_{12} + \mu_{34}$

l_{13} estimates $\mu_{13} + \mu_{24}$

l_{23} estimates $\mu_{23} + \mu_{14}$

l_{123} estimates $\mu_{123} + \mu_4$

and

l_1 estimates $I + \mu_{1234}/2$

Another example: Study 6 variable in 8 tests 2^{6-3} design. $2^{k-p} =$

2^m tests

$k = 6$ variables to be studied

$p = 3$ variables to be introduced using interactions from base design

$m = 3$ number of variables in the base design

Base Design

Test	I	1	2	3	12	13	23	123
1	+	-	-	-	+	+	+	-
2	+	+	-	-	-	-	+	+
3	+	-	+	-	-	+	-	+
4	+	+	+	-	+	-	-	-
5	+	-	-	+	+	-	-	+
6	+	+	-	+	-	+	-	-
7	+	-	+	+	-	-	+	-
8	+	+	+	+	+	+	+	+

Introduce 4,5,6 using 3 of these interactions

Introduce variables 4,5, & 6 using interactions 12,13, & 23

Generators: $4 = 12$, $5 = 13$, $6 = 23$ or $I = 124$, $I = 135$, $I = 236$

Recipe Matrix

Test	1	2	3	4	5	6
1	-	-	-	+	+	+
2	+	-	-	-	-	+
3	-	+	-	-	+	-
4	+	+	-	+	-	-
5	-	-	+	+	-	-
6	+	-	+	-	+	-
7	-	+	+	-	-	+
8	+	+	+	+	+	+

Defining Relationship

We have p generators

$$I = 124 = 135 = 236$$

Consider their products 2 at a time

$$124 \bullet 135 = 2345$$

$$124 \bullet 236 = 1346$$

$$135 \bullet 236 = 1256$$

Products 3 at a time

$$124 \bullet 135 \bullet 236 = 456$$

In general, all the way up to “products p at a time”. So,
Defining Relationship:

$$I = 124 = 135 = 236 = 2345 = 1346 = 1256 = 456$$

There are 2^p words = 8 in the defining relationship. “Word” = a string such as “236”

Design Resolution

“The length of the shortest word in the defining relation (excluding I) is termed the design resolution” For this example, Design Resolution = III.

What does Design Resolution mean?

- Des. Res. = III: Main effects are confounded with two factor interactions (1 with 2) and 3=1+2.
- Des. Res. = IV: Main effects are confounded with three factor interactions (1 with 3) and 4=1+3. It can also be two factor interactions are confounded with one another (2 with 2) and 4=2+2.
- Des. Res. = V: Main effects confounded with 4 factor c (1 + 4 = 5), or, two-factor interactions are confounded with three factor interactions (2 + 3 =5)
- Des. Res. = II: Main effects confounded with one another. This would be very, very bad.

Linear Combinations of Effects

Consider the columns in the base design calc. matrix:

I,1,2,3,12,13,23,123. Multiply each of the columns above by each word in the defining relation.

l_I estimates

$$I + \frac{\mu_{124}}{2} + \frac{\mu_{135}}{2} + \frac{\mu_{236}}{2} + \frac{\mu_{2345}}{2} + \frac{\mu_{1346}}{2} + \frac{\mu_{1256}}{2} + \frac{\mu_{456}}{2}$$

$$l_1 \text{ est } 1+24+35+1236+12345+346+256+1456$$

$$l_2 \text{ est } 2+14+1235+36+345+12346+156+2456$$

$$l_3 \text{ est } 3+1234+15+26+245+146+12356+3456$$

$$l_{12} \text{ est } 12+4+235+136+1345+2346+56+12456$$

$$l_{13} \text{ est } 13+234+5+126+1245+46+2356+13456$$

$$l_{23} \text{ est } 23+134+125+6+45+1246+1356+23456$$

$$l_{123} \text{ est } 123+34+25+16+145+246+356+123456$$

Note that each linear combination has 8 terms, the same as the number of words in the defining relation.

For 6 variables,

1 average

6 main effects

15 2-factor interactions

20 3-factor interactions

15 4-factor interactions

6 5-factor interactions

1 6-factor interaction

Total 64 effects

Conduct the 8 tests,

In std. order (24.5, 16.0, 16.0, 23.0, 25.0, 13.5, 17.0, 24.0).

Under the assumption that 3 - factor and higher order interactions are negligible,

$l_I = 20$ estimates I

$l_1 = -1.25$ estimates $\mu_1 + \mu_{24} + \mu_{35}$

$l_2 = 0.50$ estimates $\mu_2 + \mu_{14} + \mu_{36}$

$l_3 = 0.25$ estimates $\mu_3 + \mu_{15} + \mu_{26}$

$l_{12} = 8.75$ estimates $\mu_{12} + \mu_4 + \mu_{56}$

$l_{13} = -0.5$ estimates $\mu_{13} + \mu_5 + \mu_{46}$

$l_{23} = 1.25$ estimates $\mu_{23} + \mu_6 + \mu_{45}$

$l_{123} = 1.0$ estimates $\mu_{34} + \mu_{25} + \mu_{16}$

Note that main effects are confounded with 2-factor interactions, and the design resolution is III.

Normal plot shows l_{12} important. Other μ 's, may also be important but cancel one another out. If the sum " $\mu_{12} + \mu_4 + \mu_{56}$ " is important, it could be due to μ_{12} and/or μ_4 and/or

μ₅₆

Given no other experiment results, we must apply process knowledge to figure out what is important.

Fractional Fact Designs: Good for screening experiments. Find out important variables from a large list

- Want to develop/create a 2^{k-p} FFD. $k-p=m$
- Write out calculation matrix for 2^m full factorial design - base design
- Introduce the p new variables thru the interaction columns in the base design calculation matrix - p generators
- Write out the recipe matrix for the 2^{k-p} design
- Use the generators to develop defining relationship.
 - # of words = 2^p
 - design resolution = length of shortest word in the defining relation.
- Write out the linear combination of effects that may be estimated - one linear combination for each col in the base design calculation matrix.
- Run expt. using recipe matrix
- Calculate l_f, l_1, l_2, \dots by multiplying each column in the base design calculation matrix by the column of responses and divide by the appropriate # of “+” signs.

More nomenclature:

2^7 full factorial = 128 tests

2^{7-1} frac. fact = 64 tests = 1/2 fraction

2^{7-2} frac. fact = 32 tests = 1/4 fraction

2^{7-3} frac. fact = 16 tests = 1/8 fraction

2^{7-4} frac. fact = 8 tests = 1/16 fraction

2⁴ Full Factorial in Std. Order

Test	1	2	3	4
1	-	-	-	-
2	+	-	-	-
3	-	+	-	-
4	+	+	-	-
5	-	-	+	-
6	+	-	+	-
7	-	+	+	-
8	+	+	+	-
9	-	-	-	+
10	+	-	-	+
11	-	+	-	+
12	+	+	-	+
13	-	-	+	+
14	+	-	+	+
15	-	+	+	+
16	+	+	+	+

2⁴⁻¹ design with 4=123

Test	1	2	3	4= 123
1	-	-	-	-
10	+	-	-	+
11	-	+	-	+
4	+	+	-	-
13	-	-	+	+
6	+	-	+	-
7	-	+	+	-
16	+	+	+	+

What about the remaining 8 tests?

Test	1	2	3	4	123
2	+	-	-	-	+
3	-	+	-	-	+
5	-	-	+	-	+
8	+	+	+	-	+
9	-	-	-	+	-
12	+	+	-	+	-
14	+	-	+	+	-
15	-	+	+	+	-

The generator for this 2^{4-1} design is $4 = -123$. Thus, variable 4 is introduced by negating/switching/folding the signs on the 123 interaction column.

For 2^{4-1} design we can introduce variable 4 using 123 or -123. The generator is a member of $4 = \pm 123$ family of generators.

We might consider using the family of generators $4 = \pm 12$, but this is only of resolution III, while $4 = \pm 123$ is of resolution IV. We “Always” want designs with highest resolution.

Let’s examine the linear combinations that are obtained from the defining relationships of our two 2^{4-1} designs with $4 = \pm 123$

$$I = 1234$$

$$l_1 \text{ est } I + \frac{\mu_{1234}}{2}$$

$$l_1 \text{ est } 1 + 234$$

$$l_2 \text{ est } 2 + 134$$

$$I = -1234$$

$$l'_1 \text{ est } I - \frac{\mu_{1234}}{2}$$

$$l'_1 \text{ est } 1 - 234$$

$$l'_2 \text{ est } 2 - 134$$

$l_3 \text{ est } 3 + 124$	$l'_3 \text{ est } 3 - 124$
$l_{12} \text{ est } 12 + 34$	$l'_{12} \text{ est } 12 - 34$
$l_{13} \text{ est } 13 + 24$	$l'_{13} \text{ est } 13 - 24$
$l_{23} \text{ est } 23 + 14$	$l'_{23} \text{ est } 23 - 14$
$l_{123} \text{ est } 123 + 4$	$l'_{123} \text{ est } 123 - 4$

The results from the 2 FFD's may be combined to resolve effect confounding

$(l_1 + l'_1)/2 \text{ est } I$	$(l_1 - l'_1) \text{ est } 1234$
$(l_1 + l'_{1'})/2 \text{ est } 1$	$(l_1 - l'_{1'})/2 \text{ est } 234$
$(l_2 + l'_2)/2 \text{ est } 2$	$(l_2 - l'_2)/2 \text{ est } 134$
$(l_3 + l'_3)/2 \text{ est } 3$	$(l_3 - l'_3)/2 \text{ est } 124$
$(l_{12} + l'_{12})/2 \text{ est } 12$	$(l_{12} - l'_{12})/2 \text{ est } 34$
$(l_{13} + l'_{13})/2 \text{ est } 13$	$(l_{13} - l'_{13})/2 \text{ est } 24$
$(l_{23} + l'_{23})/2 \text{ est } 23$	$(l_{23} - l'_{23})/2 \text{ est } 14$
$(l_{123} + l'_{123})/2 \text{ est } 123$	$(l_{123} - l'_{123})/2 \text{ est } 4$

Thus, if we were to run a 2^{4-1} FFD ($4=123$) followed up by a 2^{4-1} FFD ($4=-123$), we may calculate/recover all the information associated with a 2^4 full factorial