

Properties of Detection Matrix and Parallel Flats Fraction for 3^n Search Design⁺

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ABSTRACT

A parallel flats fraction for the 3^n design is defined as union of flats $\{t | At = \underline{c}_i \pmod{3}\}$, $i=1, 2, \dots, f$ and is symbolically written as $At = C$ where A is rank r . The A matrix partitions the effects into $u+1$ alias sets where $u=(3^{n-r}-1)/2$. For each alias set the f flats produce an $ACPM$ from which a detection matrix is constructed. The set of all possible parallel flats fraction C can be partitioned into equivalence classes. In this paper, we develop some properties of a detection matrix and C .

1. Introduction

A parallel flats fraction for the 3^n factorial experiment is defined as the union of flats $\{t | At = \underline{C}_i \pmod{3}, i=1, 2, \dots, f\}$ and is symbolically written as $At = C$ where A is a rxn matrix with rank r and $C = (\underline{C}_1, \underline{C}_2, \dots, \underline{C}_f)$ is a $rx f$ matrix. Note that f denotes the number of flats.

The A matrix partitions the effects into $u+1$ alias sets where $u=(3^{n-r}-1)/2$. For each alias set the f flats produce an alias component permutation matrix ($ACPM$) with elements from the permutation group S_3 .

Um(1981) showed that the set of all possible parallel flats fraction C for a given A and given size can be partitioned into equivalence classes. Table 1 shows the equivalence classes of C matrix for the 3^4 factorial.

A detection vector of the $ACPM$ was constructed for each combination of k or fewer two-factor interactions by Um(1983). Also the relationship between the detection vectors

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*Research supported by University Research Grant.

has been shown. Table 2 shows the detection matrix for the 3^4 factorial.

2. Basic Lemmas

Suppose that the design T is obtained from solution to $A\underline{t}=C$, where A is $r \times n$ with rank r and C is $r \times f$ matrix, Then the following lemmas about the C matrix can be summarized from Um(1981).

Lemma 1. Let the design T^* be obtained from solutions to $A\underline{t}=C^*$ where C^* is obtained from C by permuting columns of C except the first column. Then the designs T and T^* are equivalent and C, C^* belong to the same equivalence class.

Lemma 2. Let the design T^{**} be obtained from solutions to $A\underline{t}=C^{**}$ where C^{**} is obtained by adding the vector \underline{v} with elements in $GF(3)$ to each of the columns of C . Choose \underline{v} such that there exists one column of 0's after adding \underline{v} to each column of C , then the designs T and T^{**} are equivalent and C, C^{**} belong to the same equivalence class.

Lemma 3. Let the design T^{***} be obtained from solutions to $A\underline{t}=C^{***}$ where $C^{***}=2C$. Then the designs T and T^{***} are equivalent and C, C^{***} belong to the same equivalence class.

Lemma 2 and Lemma 3 can be combined to establish designs which are equivalent. If T^* is obtained from solutions to $A\underline{t}=C^*$ where $C^*=2C_+(\underline{v}, \underline{v}, \dots, \underline{v})$, then the designs T and T^* are equivalent and C, C^* belong to the same equivalence class.

3. Main Results

Note that elements of $ACPM$ depend on a C matrix and the detection vectors are obtained from $ACPM$. It is important to relate the detection vectors to the C matrix. We now develop some relationships between the equivalence class of C matrix and the detection vectors.

Lemma 4. Let the matrix C^* be obtained from C by permuting columns of C except the first column. Then the detection vector obtained from C^* are just a permutation of elements of the detection vectors obtained from C .

Proof. Each column of C matrix corresponds to one row of $ACPM P_i$, $i=1, 2, \dots, u$, where $u+1$ is the number of alias sets.

Suppose that two columns of C , say column 2 and column 3, are permuted. Then for every i the corresponding rows of Pi , that is row 2 and row 3, are interchanged. Therefore, the column1-2 (the difference between row 1 and row 2) of the detection vector obtained from C becomes the column 1-3 (the difference between row 1 and row 3) of the detection vector obtained from C^* . This is true for any permuting columns of C except the first column. This completes the proof.

Lemma 5. Two detection vectors obtained from C and C^{**} , where $C^{**}=2C$, are the same.

Proof. In order to get $ACPM$, suppose that we have

$$P^*=(0, x_2 \underline{t}_0' \underline{e}_2, x_3 \underline{t}_0' \underline{e}_3, \dots, x_q \underline{t}_0' \underline{e}_q)(\text{See Um(1980)}).$$

Then the $ACPM$ whose elements are composed of 0,1 and 2 can be obtained. Therefore, multiplying the matrix C by 2 implies simply that the elements of $ACPM$ are multiplied by 2. After this the transformations are performed. Then (012) obtained from C becomes (021) obtained from C^{**} , and (021) becomes (012). Hence the detection vectors are not affected. This completes the proof.

The implication of Lemma 5 is that if one column of C^{**} can be obtained by multiplying the corresponding column of C by 2 then the detection elements for the difference between the first row and the corresponding row of $ACPM$ are the same for C and C^{**} .

Lemma 6. Let the matrix C^{***} be obtained from C by adding nonzero vector \underline{v} . Then the detection vector obtained from C^{***} are the same with the detection vector obtained from C or a permutation of columns of the detection vector obtained from C .

Proof. Suppose that we choose \underline{v} such that the second column will have $\underline{0}$ after adding \underline{v} to the C . Then the first column of C^{***} is $\underline{0}$ and the second column is \underline{v} . This implies that the second column of C^{***} can be obtained by multiplying the second column of C by 2. Therefore, the detection elements for the difference between the first row and the second row of $ACPM$ are not changed with C and C^{***} . This means that for any choice of column of C the detection element for the difference the first row and the corresponding row of $ACPM$ are the same for C and C^{***} .

Without lose of generality let $P^*=(0, c_1, c_2)$ where the corresponding effects are E_1, E_2, E_3 . Let $C=\begin{bmatrix} 0 & c_{12} & c_{13} \\ 0 & c_{22} & c_{23} \end{bmatrix}$ where $c_{ij} \in GF(3)$ and the columns are different from each other, and let $V=(-c_{12} -c_{22})$. Consider the detection vectors for various values of c_{ij} .

(Case 1). One of c_{12} and c_{22} is zero. Suppose that $c_{22}=0$. Then clearly c_{12} is 1 or 2,

$C^{***} = \begin{bmatrix} 0 & -c_{12} & -c_{12} + c_{13} \\ 0 & 0 & c_{23} \end{bmatrix}$, and the following *ACPM* are obtained from P^* :

| <i>ACPM</i> for C | | | <i>ACPM</i> for C^{***} | | |
|--|-------|-------|---|-------|-------|
| E_1 | E_2 | E_3 | E_1 | E_2 | E_3 |
| $\begin{bmatrix} 0 & 0 & 0 \\ 0 & c_{12} & 0 \\ 0 & c_{13} & c_{23} \end{bmatrix}$ | | | $\begin{bmatrix} 0 & 0 & 0 \\ 0 & -c_{12} & 0 \\ 0 & -c_{12} + c_{13} & c_{23} \end{bmatrix}$ | | |

For any choice of c_{23} the detection vector for the effect E_3 is the same with C and C^{***} .

Consider the detection vectors for the effect E_2 . Suppose that $c_{12} = c_{13}$. Then $-c_{12} \neq 0$ and $-c_{12} + c_{13} = 0$.

Therefore, the detection vector for C^{***} can be obtained by interchanging the second column with the third column of the detection vector for C . Suppose that $c_{12} \neq c_{13}$. Then there are four possible cases:

$(c_{12}, c_{13}) = (1, 0), (2, 0), (1, 2), (2, 1)$.

a) Let $c_{12} = 1$ and $c_{13} = 0$. Then $C^{***} = \begin{bmatrix} 0 & 2 & 2 \\ 0 & 0 & c_{23} \end{bmatrix}$ and

| <i>ACPM</i> for C | | | <i>ACPM</i> for C^{***} | | |
|--|-------|-------|--|-------|-------|
| E_1 | E_2 | E_3 | E_1 | E_2 | E_3 |
| $\begin{bmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & c_{23} \end{bmatrix}$ | | | $\begin{bmatrix} 0 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 2 & c_{23} \end{bmatrix}$ | | |

The detection vectors for C^{***} can be obtained by permuting the second column and the third column of the detection vectors for C . Similarly, this holds for $c_{12} = 2$ and $c_{13} = 0$.

b). Let $c_{12} = 1$ and $c_{13} = 2$. Then we have

| <i>ACPM</i> for C | | | <i>ACPM</i> for C^{***} | | |
|--|-------|-------|--|-------|-------|
| E_1 | E_2 | E_3 | E_1 | E_2 | E_3 |
| $\begin{bmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 2 & c_{23} \end{bmatrix}$ | | | $\begin{bmatrix} 0 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 1 & c_{23} \end{bmatrix}$ | | |

Both matrices produce the same detection vectors. This holds for $c_{12} = 2$ and $c_{13} = 1$. Similar arguments hold for $c_{12} = 0$.

(Case 2). c_{12} and c_{22} are 1 or 2. Suppose that $c_{12} = c_{22}$. Then we have

| <i>ACPM</i> for C | | | <i>ACPM</i> for C^{***} | | |
|---|-------|-------|---|-------|-------|
| E_1 | E_2 | E_3 | E_1 | E_2 | E_3 |
| $\begin{bmatrix} 0 & 0 & 0 \\ 0 & c_{12} & c_{12} \\ 0 & c_{13} & c_{23} \end{bmatrix}$ | | | $\begin{bmatrix} 0 & 0 & 0 \\ 0 & -c_{12} & -c_{12} \\ 0 & -c_{12} + c_{13} & -c_{12} + c_{23} \end{bmatrix}$ | | |

It is clear that if $c_{13}=0$ or 1 then for the effect E_2 the detection vector for C^{***} can be obtained by permuting the second column and the third column of the detection vectors for C . If $c_{13}=2$ then the detection vectors are the same for both cases.

Suppose that $c_{12} \neq c_{22}$. Then we have

$$\begin{array}{ccc}
 \text{ACPM for } C & & \text{ACPM for } C^{***} \\
 \begin{array}{ccc} E_1 & E_2 & E_3 \\ \begin{bmatrix} 0 & 0 & 0 \\ 0 & c_{12} & c_{22} \\ 0 & c_{13} & c_{23} \end{bmatrix} \end{array} & & \begin{array}{ccc} E_1 & E_2 & E_3 \\ \begin{bmatrix} 0 & 0 & 0 \\ 0 & -c_{12} & -c_{22} \\ 0 & -c_{12} - c_{13} & -c_{22} + c_{23} \end{bmatrix} \end{array}
 \end{array}$$

Obviously, if $c_{13}=0$ or 1, then for the effect E_2 the detection vector for C^{***} can be obtained by permuting the two columns of detection vector for C . If $c_{13}=2$ then the detection vectors are the same for both cases. The same argument holds for the effect E_3 with the various values of c_{23} .

The above arguments in Case 1 and Case 2 are true for any choice of \underline{v} and for any form of P^* .

This completes the proof.

Combining Lemmas 4, 5 and 6 the following theorem is obtained.

Theorem. Suppose that C matrix C_1 and C_2 are rxf matrices where C_1 and C_2 belong to the same equivalence class. Then the detection vectors for C_1 and C_2 are the same or permute each other.

4. Example

Consider a 3^4 factorial experiment for which it can be assumed that all three and four-factor interaction effects are negligible. The A matrix for this example will be taken as

$$A = \begin{bmatrix} 1 & 1 & 1 & 0 \\ 1 & 2 & 0 & 1 \end{bmatrix},$$

thus there are flats of size nine. The alias sets are

$$S_0 = \{\mu\},$$

$$S_1 = \{F_1, F_2, F_3, F_2 F_4^2, F_3 F_4\}, \quad S_2 = \{F_2, F_1 F_3, F_1 F_4, F_3 F_4^2\},$$

$$S_3 = \{F_3, F_1 F_2, F_1 F_4^2, F_2 F_4\}, \quad S_4 = \{F_4, F_1 F_2^2, F_1 F_3^2, F_2 F_3^2\}.$$

An example of a parallel flats fraction in 27 runs is given with

$$C = (\underline{C}_1, \underline{C}_2, \underline{C}_3) \text{ as } C = \begin{bmatrix} 0 & 0 & 1 \\ 0 & 1 & 2 \end{bmatrix}.$$

By choosing the main effect in each alias set as the identified effect, the *ACPM* are

$$\begin{array}{c}
 \begin{array}{cccc}
 F_1 & F_2 F_3 & F_2 F_4^2 & F_3 F_4^4 \\
 \begin{bmatrix} e & e & e & e \\ e & e & (021) & (012) \\ e & (021) & (012) & e \end{bmatrix}
 \end{array} &
 \begin{array}{cccc}
 F_2 & F_1 F_3 & F_1 F_4 & F_3 F_4^2 \\
 \begin{bmatrix} e & e & e & e \\ e & e & (012) & (021) \\ e & (021) & (021) & (021) \end{bmatrix}
 \end{array} \\
 \\
 \begin{array}{cccc}
 F_3 & F_1 F_2 & F_1 F_4^2 & F_2 F_4 \\
 \begin{bmatrix} e & e & e & e \\ e & e & (021) & (012) \\ e & (021) & e & (012) \end{bmatrix}
 \end{array} &
 \begin{array}{cccc}
 F_4 & F_1 F_2^2 & F_1 F_3^2 & F_2 F_3^2 \\
 \begin{bmatrix} e & e & e & e \\ e & (021) & (021) & (021) \\ e & (012) & e & (021) \end{bmatrix}
 \end{array}
 \end{array}$$

Table 1 shows the equivalence classes of *C* and Table 2 shows the detection vectors for this example.

TABLE 1. EQUIVALENCE CLASSES OF 0 MATRIX FOR 3⁴, DESIGN

| | | | | | |
|---------|-------|-------|-------|-------|-------|
| CLASS 1 | | | | | |
| 0 0 1 | 0 0 1 | 0 2 2 | 0 1 0 | 0 1 0 | 0 2 2 |
| 0 1 2 | 0 2 1 | 0 2 1 | 0 2 1 | 0 1 2 | 0 1 2 |
| 0 0 2 | 0 0 2 | 0 1 1 | 0 2 0 | 0 2 0 | 0 1 1 |
| 0 2 1 | 0 1 2 | 0 1 2 | 0 1 2 | 0 2 1 | 0 2 1 |
| CLASS 2 | | | | | |
| 0 1 2 | 0 2 1 | 0 2 1 | 0 2 1 | 0 1 2 | 0 1 2 |
| 0 0 1 | 0 0 1 | 0 2 2 | 0 1 0 | 0 1 0 | 0 2 2 |
| 0 2 1 | 0 1 2 | 0 1 2 | 0 1 2 | 0 2 1 | 0 2 1 |
| 0 0 2 | 0 0 2 | 0 1 1 | 0 2 0 | 0 2 0 | 0 1 1 |
| CLASS 3 | | | | | |
| 0 0 1 | 0 0 1 | 0 2 2 | 0 1 0 | 0 1 0 | 0 2 2 |
| 0 1 0 | 0 2 2 | 0 1 0 | 0 0 1 | 0 2 2 | 0 0 1 |
| 0 0 2 | 0 0 2 | 0 1 1 | 0 2 0 | 0 2 0 | 0 1 1 |
| 0 2 0 | 0 1 1 | 0 2 0 | 0 0 2 | 0 1 1 | 0 0 2 |
| CLASS 4 | | | | | |
| 0 0 1 | 0 0 1 | 0 2 2 | 0 1 0 | 0 1 0 | 0 2 2 |
| 0 1 1 | 0 2 0 | 0 0 2 | 0 1 1 | 0 0 2 | 0 2 0 |
| 0 0 2 | 0 0 2 | 0 1 1 | 0 2 0 | 0 2 0 | 0 1 1 |
| 0 2 2 | 0 1 0 | 0 0 1 | 0 2 2 | 0 0 1 | 0 1 0 |
| CLASS 5 | | | | | |
| 0 0 0 | 0 0 0 | | | | |
| 0 1 2 | 0 2 1 | | | | |
| CLASS 6 | | | | | |
| 0 1 2 | 0 2 1 | | | | |
| 0 0 0 | 0 0 0 | | | | |
| CLASS 7 | | | | | |
| 0 1 2 | 0 2 1 | | | | |
| 0 1 2 | 0 2 1 | | | | |
| CLASS 8 | | | | | |
| 0 1 2 | 0 2 1 | | | | |
| 0 2 1 | 0 1 2 | | | | |

TABLE 2. THE DETECTION MATRIX FOR THE 3^4 DESIGN

| | | P1 | | | P2 | | | P3 | | | P4 | | |
|----|-------|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|
| | | 1-2 | 1-3 | 2-3 | 1-2 | 1-3 | 2-3 | 1-2 | 1-3 | 2-3 | 1-2 | 1-3 | 2-3 |
| 1 | MAIN | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 2 | 12 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 1 | 1 | 1 | 1 | 1 |
| 3 | 13 | 0 | 0 | 0 | 0 | 1 | 1 | 0 | 0 | 0 | 1 | 0 | 1 |
| 4 | 14 | 0 | 0 | 0 | 1 | 1 | 1 | 1 | 0 | 1 | 0 | 0 | 0 |
| 5 | 23 | 0 | 1 | 1 | 0 | 0 | 0 | 0 | 0 | 0 | 1 | 1 | 0 |
| 6 | 24 | 1 | 1 | 1 | 0 | 0 | 0 | 1 | 1 | 0 | 0 | 0 | 0 |
| 7 | 34 | 1 | 0 | 1 | 1 | 1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 8 | 12 13 | 0 | 0 | 0 | 0 | 1 | 1 | 0 | 1 | 1 | 1 | 1 | 1 |
| 9 | 12 14 | 0 | 0 | 0 | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 |
| 10 | 12 23 | 0 | 1 | 1 | 0 | 0 | 0 | 0 | 1 | 1 | 1 | 1 | 1 |
| 11 | 12 22 | 1 | 1 | 1 | 0 | 0 | 0 | 1 | 1 | 1 | 1 | 1 | 1 |
| 12 | 12 34 | 1 | 0 | 1 | 1 | 1 | 0 | 0 | 1 | 1 | 1 | 1 | 1 |
| 13 | 13 14 | 0 | 0 | 0 | 1 | 1 | 1 | 1 | 0 | 1 | 1 | 0 | 1 |
| 14 | 13 23 | 0 | 1 | 1 | 0 | 1 | 1 | 0 | 0 | 0 | 1 | 1 | 1 |
| 15 | 13 24 | 1 | 1 | 1 | 0 | 1 | 1 | 1 | 1 | 0 | 1 | 0 | 1 |
| 16 | 13 34 | 1 | 0 | 1 | 1 | 1 | 1 | 0 | 0 | 0 | 1 | 0 | 1 |
| 17 | 14 23 | 0 | 1 | 1 | 1 | 1 | 1 | 1 | 0 | 1 | 1 | 1 | 0 |
| 18 | 14 24 | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 0 | 0 | 0 |
| 19 | 14 34 | 1 | 0 | 1 | 1 | 1 | 1 | 1 | 0 | 1 | 0 | 0 | 0 |
| 20 | 23 24 | 1 | 1 | 1 | 0 | 0 | 0 | 1 | 1 | 0 | 1 | 1 | 0 |
| 21 | 23 34 | 1 | 1 | 1 | 1 | 1 | 0 | 0 | 0 | 0 | 1 | 1 | 0 |
| 22 | 24 34 | 1 | 1 | 1 | 1 | 1 | 0 | 1 | 1 | 0 | 0 | 0 | 0 |

COLUMN 4-8 DENOTE THE SUBSCRIPTS OF TWO-FACTOR INTERACTIONS

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