

# Semi-Latin Squares and their Extensions

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British Combinatorial Conference, Cardiff, July 2026

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In the case that  $k = 1$ , we just get a Latin square of order  $n$ ,  
so we always assume that  $k > 1$ .

# An example of a semi-Latin square

|     |               |     |     |               |     |     |               |     |     |               |     |     |               |     |
|-----|---------------|-----|-----|---------------|-----|-----|---------------|-----|-----|---------------|-----|-----|---------------|-----|
| $A$ | $\alpha$      | $a$ | $B$ | $\beta$       | $b$ | $C$ | $\gamma$      | $c$ | $D$ | $\delta$      | $d$ | $E$ | $\varepsilon$ | $e$ |
| $E$ | $\delta$      | $c$ | $A$ | $\varepsilon$ | $d$ | $B$ | $\alpha$      | $e$ | $C$ | $\beta$       | $a$ | $D$ | $\gamma$      | $b$ |
| $D$ | $\beta$       | $e$ | $E$ | $\gamma$      | $a$ | $A$ | $\delta$      | $b$ | $B$ | $\varepsilon$ | $c$ | $C$ | $\alpha$      | $d$ |
| $C$ | $\varepsilon$ | $b$ | $D$ | $\alpha$      | $c$ | $E$ | $\beta$       | $d$ | $A$ | $\gamma$      | $e$ | $B$ | $\delta$      | $a$ |
| $B$ | $\gamma$      | $d$ | $C$ | $\delta$      | $e$ | $D$ | $\varepsilon$ | $a$ | $E$ | $\alpha$      | $b$ | $A$ | $\beta$       | $c$ |

# An example of a semi-Latin square

|                       |                       |                       |                       |                       |
|-----------------------|-----------------------|-----------------------|-----------------------|-----------------------|
| $A$ $\alpha$ $a$      | $B$ $\beta$ $b$       | $C$ $\gamma$ $c$      | $D$ $\delta$ $d$      | $E$ $\varepsilon$ $e$ |
| $E$ $\delta$ $c$      | $A$ $\varepsilon$ $d$ | $B$ $\alpha$ $e$      | $C$ $\beta$ $a$       | $D$ $\gamma$ $b$      |
| $D$ $\beta$ $e$       | $E$ $\gamma$ $a$      | $A$ $\delta$ $b$      | $B$ $\varepsilon$ $c$ | $C$ $\alpha$ $d$      |
| $C$ $\varepsilon$ $b$ | $D$ $\alpha$ $c$      | $E$ $\beta$ $d$       | $A$ $\gamma$ $e$      | $B$ $\delta$ $a$      |
| $B$ $\gamma$ $d$      | $C$ $\delta$ $e$      | $D$ $\varepsilon$ $a$ | $E$ $\alpha$ $b$      | $A$ $\beta$ $c$       |

Here there are 5 rows, 5 columns, and 3 plots per cell.

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|                       |                       |                       |                       |                       |
|-----------------------|-----------------------|-----------------------|-----------------------|-----------------------|
| $A$ $\alpha$ $a$      | $B$ $\beta$ $b$       | $C$ $\gamma$ $c$      | $D$ $\delta$ $d$      | $E$ $\varepsilon$ $e$ |
| $E$ $\delta$ $c$      | $A$ $\varepsilon$ $d$ | $B$ $\alpha$ $e$      | $C$ $\beta$ $a$       | $D$ $\gamma$ $b$      |
| $D$ $\beta$ $e$       | $E$ $\gamma$ $a$      | $A$ $\delta$ $b$      | $B$ $\varepsilon$ $c$ | $C$ $\alpha$ $d$      |
| $C$ $\varepsilon$ $b$ | $D$ $\alpha$ $c$      | $E$ $\beta$ $d$       | $A$ $\gamma$ $e$      | $B$ $\delta$ $a$      |
| $B$ $\gamma$ $d$      | $C$ $\delta$ $e$      | $D$ $\varepsilon$ $a$ | $E$ $\alpha$ $b$      | $A$ $\beta$ $c$       |

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There are 15 treatments,  
each occurring once in each row and once in each column.

How should we analyse data from such an experiment?

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They effectively form a system of incomplete blocks, so the design needs to take account of this.

Once people started to include the cells in the data analysis, Frank Yates became enthusiastic about using semi-Latin squares.

# Randomization

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This randomization justifies the linear model whose variance-covariance matrix has the following eigenspaces (also called **strata**).

| Subspace                      | dimension    |
|-------------------------------|--------------|
| Mean                          | 1            |
| Rows                          | $n - 1$      |
| Columns                       | $n - 1$      |
| Cells within Rows and Columns | $(n - 1)^2$  |
| Plots within Cells            | $n^2(k - 1)$ |

Treatment contrasts are orthogonal to the first three strata, so any treatment contrast that is an eigenvector of the matrix of projection onto one of the last two strata is also an eigenvector of the matrix of projection onto the other.

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The design is called a **Trojan square** if it can be made by superposing the letters of  $k$  mutually orthogonal Latin squares, each with  $n$  different treatments.

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Our previous example with  $n = 5$  and  $k = 3$  is Trojan.

|                               |                               |                               |                               |                               |
|-------------------------------|-------------------------------|-------------------------------|-------------------------------|-------------------------------|
| $A \quad \alpha \quad a$      | $B \quad \beta \quad b$       | $C \quad \gamma \quad c$      | $D \quad \delta \quad d$      | $E \quad \varepsilon \quad e$ |
| $E \quad \delta \quad c$      | $A \quad \varepsilon \quad d$ | $B \quad \alpha \quad e$      | $C \quad \beta \quad a$       | $D \quad \gamma \quad b$      |
| $D \quad \beta \quad e$       | $E \quad \gamma \quad a$      | $A \quad \delta \quad b$      | $B \quad \varepsilon \quad c$ | $C \quad \alpha \quad d$      |
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| $B \quad \gamma \quad d$      | $C \quad \delta \quad e$      | $D \quad \varepsilon \quad a$ | $E \quad \alpha \quad b$      | $A \quad \beta \quad c$       |

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## Theorem

*If there is a Trojan square for the given values of  $n$  and  $k$ , then it is A-optimal.*

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## Theorem

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An  $s$ -fold **inflation** of a semi-Latin  $(n \times n)/k$  square is a semi-Latin  $(n \times n)/ks$  square obtained by replacing each treatment in the first semi-Latin square by  $s$  new treatments.

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## Theorem

*If there is a Trojan square of size  $(n \times n)/(n-1)$ , then all of its inflations are A-optimal.*

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Soicher proved that there exists a  $(6 \times 6)/5s$  uniform semi-Latin square whenever  $s \geq 2$ .

The first way that I extend the notion of semi-Latin squares is by relaxing the condition that each treatment occurs precisely once in each row and once in each column. We maintain the rules that

1. no treatment occurs more than once in any cell;
2. all treatments occur equally often in each row;
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RAB and Hervé Monod first did this in 2001 for an experiment where the experimental units were the half-leaves. The  $p$  plants can be thought of as columns, and the  $h$  leaf-heights can be thought of as rows. In this case,  $k = 2$ .

# An example from an experiment on plant disease

|          |          |          |          |          |          |          |          |          |          |          |          |          |          |          |          |          |
|----------|----------|----------|----------|----------|----------|----------|----------|----------|----------|----------|----------|----------|----------|----------|----------|----------|
| height 4 | <i>D</i> | <i>E</i> | <i>E</i> | <i>F</i> | <i>F</i> | <i>G</i> | <i>G</i> | <i>H</i> | <i>H</i> | <i>A</i> | <i>A</i> | <i>B</i> | <i>B</i> | <i>C</i> | <i>C</i> | <i>D</i> |
| height 3 | <i>A</i> | <i>G</i> | <i>B</i> | <i>H</i> | <i>C</i> | <i>A</i> | <i>D</i> | <i>B</i> | <i>E</i> | <i>C</i> | <i>F</i> | <i>D</i> | <i>G</i> | <i>E</i> | <i>H</i> | <i>F</i> |
| height 2 | <i>H</i> | <i>C</i> | <i>A</i> | <i>D</i> | <i>B</i> | <i>E</i> | <i>C</i> | <i>F</i> | <i>D</i> | <i>G</i> | <i>E</i> | <i>H</i> | <i>F</i> | <i>A</i> | <i>G</i> | <i>B</i> |
| height 1 | <i>B</i> | <i>F</i> | <i>C</i> | <i>G</i> | <i>D</i> | <i>H</i> | <i>E</i> | <i>A</i> | <i>F</i> | <i>B</i> | <i>G</i> | <i>C</i> | <i>H</i> | <i>D</i> | <i>A</i> | <i>E</i> |
|          | plant 1  | plant 2  | plant 3  | plant 4  | plant 5  | plant 6  | plant 7  | plant 8  |          |          |          |          |          |          |          |          |

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| height 4 | <i>D</i> | <i>E</i> | <i>E</i> | <i>F</i> | <i>F</i> | <i>G</i> | <i>G</i> | <i>H</i> | <i>H</i> | <i>A</i> | <i>A</i> | <i>B</i> | <i>B</i> | <i>C</i> | <i>C</i> | <i>D</i> |
| height 3 | <i>A</i> | <i>G</i> | <i>B</i> | <i>H</i> | <i>C</i> | <i>A</i> | <i>D</i> | <i>B</i> | <i>E</i> | <i>C</i> | <i>F</i> | <i>D</i> | <i>G</i> | <i>E</i> | <i>H</i> | <i>F</i> |
| height 2 | <i>H</i> | <i>C</i> | <i>A</i> | <i>D</i> | <i>B</i> | <i>E</i> | <i>C</i> | <i>F</i> | <i>D</i> | <i>G</i> | <i>E</i> | <i>H</i> | <i>F</i> | <i>A</i> | <i>G</i> | <i>B</i> |
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The treatment pairs  $\{A, E\}$ ,  $\{B, F\}$ ,  $\{C, G\}$  and  $\{D, H\}$ , each come in two different cells in the bottom row.

# An example from an experiment on plant disease

|          |          |          |          |          |          |          |          |          |          |          |          |          |          |          |          |          |
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| height 4 | <i>D</i> | <i>E</i> | <i>E</i> | <i>F</i> | <i>F</i> | <i>G</i> | <i>G</i> | <i>H</i> | <i>H</i> | <i>A</i> | <i>A</i> | <i>B</i> | <i>B</i> | <i>C</i> | <i>C</i> | <i>D</i> |
| height 3 | <i>A</i> | <i>G</i> | <i>B</i> | <i>H</i> | <i>C</i> | <i>A</i> | <i>D</i> | <i>B</i> | <i>E</i> | <i>C</i> | <i>F</i> | <i>D</i> | <i>G</i> | <i>E</i> | <i>H</i> | <i>F</i> |
| height 2 | <i>H</i> | <i>C</i> | <i>A</i> | <i>D</i> | <i>B</i> | <i>E</i> | <i>C</i> | <i>F</i> | <i>D</i> | <i>G</i> | <i>E</i> | <i>H</i> | <i>F</i> | <i>A</i> | <i>G</i> | <i>B</i> |
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The treatment pairs  $\{A, E\}$ ,  $\{B, F\}$ ,  $\{C, G\}$  and  $\{D, H\}$ , each come in two different cells in the bottom row.

Each remaining treatment pair occurs in exactly one cell.

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| height 3 | <i>A</i> | <i>G</i> | <i>B</i> | <i>H</i> | <i>C</i> | <i>A</i> | <i>D</i> | <i>B</i> | <i>E</i> | <i>C</i> | <i>F</i> | <i>D</i> | <i>G</i> | <i>E</i> | <i>H</i> | <i>F</i> |
| height 2 | <i>H</i> | <i>C</i> | <i>A</i> | <i>D</i> | <i>B</i> | <i>E</i> | <i>C</i> | <i>F</i> | <i>D</i> | <i>G</i> | <i>E</i> | <i>H</i> | <i>F</i> | <i>A</i> | <i>G</i> | <i>B</i> |
| height 1 | <i>B</i> | <i>F</i> | <i>C</i> | <i>G</i> | <i>D</i> | <i>H</i> | <i>E</i> | <i>A</i> | <i>F</i> | <i>B</i> | <i>G</i> | <i>C</i> | <i>H</i> | <i>D</i> | <i>A</i> | <i>E</i> |
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The treatment pairs  $\{A, E\}$ ,  $\{B, F\}$ ,  $\{C, G\}$  and  $\{D, H\}$ , each come in two different cells in the bottom row.

Each remaining treatment pair occurs in exactly one cell.

So this design is  $A$ -optimal.

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However, the definition of semi-Latin square restricts each treatment to occur precisely once in each row and once in each column.

# The trouble with names

Of course, in a semi-Latin rectangle there is no rule that says that the number of rows must be different from the number of columns.

However, the definition of semi-Latin square restricts each treatment to occur precisely once in each row and once in each column.

So if each treatment occurs  $s$  times in each row and  $s$  times in each column, we have to call it a **semi-Latin rectangle** if  $s \geq 2$ .

## A balanced semi-Latin rectangle

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|       |       |       |       |       |       |       |
|-------|-------|-------|-------|-------|-------|-------|
| 1 2 4 | 2 3 5 | 3 4 6 | 4 5 7 | 5 6 1 | 6 7 2 | 7 1 3 |
| 2 3 5 | 3 4 6 | 4 5 7 | 5 6 1 | 6 7 2 | 7 1 3 | 1 2 4 |
| 3 4 6 | 4 5 7 | 5 6 1 | 6 7 2 | 7 1 3 | 1 2 4 | 2 3 5 |
| 4 5 7 | 5 6 1 | 6 7 2 | 7 1 3 | 1 2 4 | 2 3 5 | 3 4 6 |
| 5 6 1 | 6 7 2 | 7 1 3 | 1 2 4 | 2 3 5 | 3 4 6 | 4 5 7 |
| 6 7 2 | 7 1 3 | 1 2 4 | 2 3 5 | 3 4 6 | 4 5 7 | 5 6 1 |
| 7 1 3 | 1 2 4 | 2 3 5 | 3 4 6 | 4 5 7 | 5 6 1 | 6 7 2 |

# Extended semi-Latin squares: the background

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Because North–South variation is usually greater per distance than East–West variation, statisticians in agricultural research in the UK decided in the 1980s that it is reasonable to divide the glasshouse into  $n$  rows and  $n$  columns, where each row is an East–West line of  $nk$  plots and each column consists of  $k$  contiguous North–South lines of plots.

# Extended semi-Latin squares: the background

In trials on glasshouse crops, there are usually pronounced row effects and column effects.

The glasshouse is typically rectangular, with its long axis East–West.

Because North–South variation is usually greater per distance than East–West variation, statisticians in agricultural research in the UK decided in the 1980s that it is reasonable to divide the glasshouse into  $n$  rows and  $n$  columns, where each row is an East–West line of  $nk$  plots and each column consists of  $k$  contiguous North–South lines of plots.

This view led to widespread use of semi-Latin squares in agricultural research.

# Extended semi-Latin squares: definition

The plot structure of an extended semi-Latin square has the form

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Now the  $n^2$  cells of size  $k$  and the  $nk$  lines of size  $n$  both form systems of incomplete-blocks, so we have to allow for both of them in our design, randomization and data analysis.

Here is an example with  $n = 4$  and  $k = 2$ .

|  |  |  |  |  |  |  |  |
|--|--|--|--|--|--|--|--|
|  |  |  |  |  |  |  |  |
|  |  |  |  |  |  |  |  |
|  |  |  |  |  |  |  |  |
|  |  |  |  |  |  |  |  |

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|   |   |   |   |   |   |   |   |
|---|---|---|---|---|---|---|---|
| * | * | + | + | † | † | £ | £ |
|   |   |   |   |   |   |   |   |
|   |   |   |   |   |   |   |   |
|   |   |   |   |   |   |   |   |

There are 4 rows, each containing 4 cells of 2 plots.

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|   |   |  |  |  |  |  |  |
|---|---|--|--|--|--|--|--|
| * | + |  |  |  |  |  |  |
| * | + |  |  |  |  |  |  |
| * | + |  |  |  |  |  |  |
| * | + |  |  |  |  |  |  |

There are 4 rows, each containing 4 cells of 2 plots.

There are 4 columns, each containing 2 lines of 4 plots.

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|--|--|--|--|--|--|--|--|
|  |  |  |  |  |  |  |  |
|  |  |  |  |  |  |  |  |
|  |  |  |  |  |  |  |  |
|  |  |  |  |  |  |  |  |

There are 4 rows, each containing 4 cells of 2 plots.

There are 4 columns, each containing 2 lines of 4 plots.

Within each column, cells are crossed with lines.

# Randomization

How should such a design be randomized?

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1. Randomize the order of the whole rows.
2. Randomize the order of the whole columns.
3. Within each whole column independently, randomize the order of the lines.

# Randomization

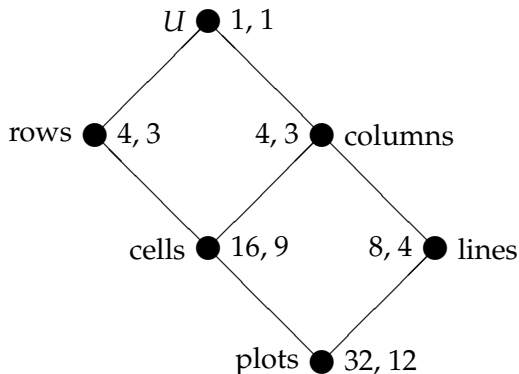
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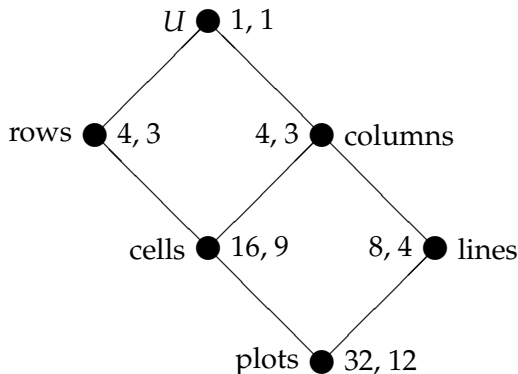
This randomization justifies the linear model whose variance-covariances matrix has the following eigenspaces (also called **strata**).

| Subspace                      | dimension         |
|-------------------------------|-------------------|
| Mean                          | 1                 |
| Rows                          | $n - 1$           |
| Columns                       | $n - 1$           |
| Lines within Columns          | $n(k - 1)$        |
| Cells within Rows and Columns | $(n - 1)^2$       |
| Plots within Cells and Lines  | $n(n - 1)(k - 1)$ |

# Hasse diagram of block factors in the example



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Here,  $U$  is the universal factor, which has a single level.  
For each factor, the first number shows its number of levels and  
the second number shows its number of degrees of freedom.

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RAB's approach is to use patterns and combinatorics. Since more information is lost to the block system with the smaller block size, it is better to start by finding the best design for that block system.

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| stratum | dim | $A$ | $B$           | $C$           | $AB$          | $AC$          | $BC$          | $ABC$         |
|---------|-----|-----|---------------|---------------|---------------|---------------|---------------|---------------|
| cells   | 9   | 0   | $\frac{1}{2}$ | $\frac{1}{2}$ | $\frac{1}{2}$ | $\frac{1}{2}$ | $\frac{1}{2}$ | $\frac{1}{2}$ |
| lines   | 4   |     |               |               |               |               |               |               |
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Next, consider the lines of size 4. There are two lines in each column. In any column, the factor  $A$  can be used to define the lines. If a factor is used like this in  $m$  columns, then its loss of information to lines is  $m/4$ .

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# Calculating the $A$ -criterion in the example

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Then the  $A$ -criterion is

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In fact, Emlyn's software shows that this design is  $A$ -optimal.

# Unrandomized version of the design in that example

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

Translation:

$$1 = 1 \quad 2 = c \quad 3 = b \quad 4 = bc \quad 5 = a \quad 6 = ab \quad 7 = ac \quad 8 = abc$$

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If they have random effects,  
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by estimating the sizes of these random effects  
and then using that information to estimate contrasts in all  
three strata  
and then combine these estimates.

This method works best when the matrices of projection onto all three strata have the same eigenvectors;  
in other words, we have general balance.

# References

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