

Semi-Latin Squares and their Extensions

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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	α	a	B	β	b	C	γ	c	D	δ	d	E	ε	e
E	δ	c	A	ε	d	B	α	e	C	β	a	D	γ	b
D	β	e	E	γ	a	A	δ	b	B	ε	c	C	α	d
C	ε	b	D	α	c	E	β	d	A	γ	e	B	δ	a
B	γ	d	C	δ	e	D	ε	a	E	α	b	A	β	c

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D β e	E γ a	A δ b	B ε c	C α d
C ε b	D α c	E β d	A γ e	B δ a
B γ d	C δ e	D ε a	E α b	A β c

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

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C ε b	D α c	E β d	A γ e	B δ a
B γ d	C δ e	D ε a	E α b	A β c

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

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

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

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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	$n - 1$
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Cells within Rows and Columns	$(n - 1)^2$
Plots within Cells	$n^2(k - 1)$

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General balance is very useful if we want to combine treatment estimates using information from different strata.

Trojan squares

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

$A \quad \alpha \quad a$	$B \quad \beta \quad b$	$C \quad \gamma \quad c$	$D \quad \delta \quad d$	$E \quad \varepsilon \quad e$
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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 $s \geq 2$.

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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.

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

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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.

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*	*	+	+	†	†	£	£

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Within each column, cells are crossed with lines.

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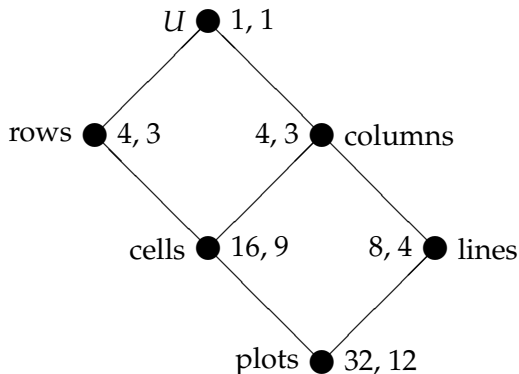
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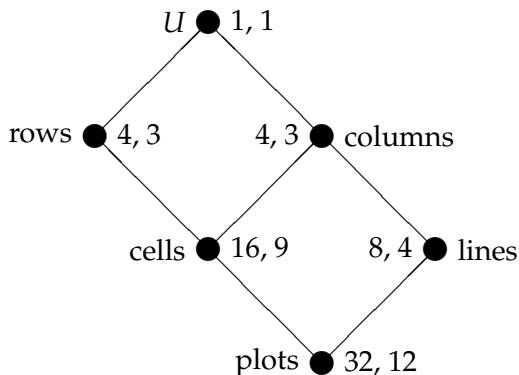
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Hasse diagram of block factors in the example



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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.

How can we find efficient extended semi-Latin squares?

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

E. R. Williams and R. A. Bailey:
Extended semi-Latin squares for use in field and glasshouse trials.

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