



Data Analytics for Data Scientists

Design of Experiments (DoE)

Suggested solutions for Exercise 09: Factorial Designs

2025

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Suggested solution 01

Properties of designs

There are full factorial and fractional factorial designs.

Please give one example of a full factorial design and one example of a fractional factorial design and explain the advantages and disadvantages of each.

Suggested answers to the questions

Full factorial design	Fractional factorial design
Description : A full factorial design is an experimental setup in which all possible combinations of factor levels are tested. For example , if there are three factors, A, B, and C, each with two levels (low and high), a full factorial design would include 2 x 2 x 2 = 8 experimental runs: (A low, B low, C low), (A low, B low, C high), (A low, B high, C low), (A low, B high, C high), (A high, B low, C low), (A high, B low, C high), (A high, B low, C high), (A high, B high, C	Description : A fractional factorial design is a type of experimental design in which only a subset of all possible combinations of factor levels is tested. For example , if there are three factors, A, B, and C, each with two levels, instead of running all 2 ³ = 8 combinations, a fractional factorial design might include only 4 selected combinations to reduce the experimental burden: (A low, B low, C low), (A low, B high, C high), (A high, B low, C high), and (A high, B high, C low).
low), and (A high, B high, C high).	(Tingh, Dingh, Ciow).
 Advantages: Allows the estimation of all main effects and interaction effects between factors. Provides a complete picture of how all factors influence the response variable. Suitable for studying complex systems where interactions between factors are important. 	 Advantages: More efficient in terms of the number of runs required, especially when there are many factors. Useful when there are resource constraints or when a preliminary study is being conducted to identify significant factors. Can provide valuable information with fewer experimental runs if the assumption of negligible interactions holds.

Disadvantages:

- Requires a large number of experimental runs, especially as the number of factors or levels increases.
- Can be very resource-intensive in terms of time, cost, and materials.

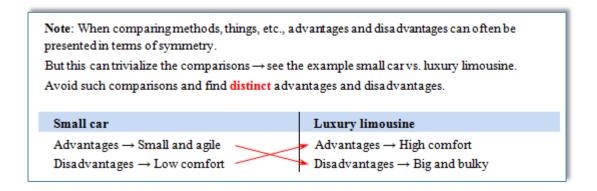
Disadvantages:

- Does not allow estimation of all interactions between factors, as only a subset of combinations is tested.
- Risk of confounding effects, where some factor effects are indistinguishable from others, making interpretation more difficult.
- May not provide a complete understanding of the system if interactions are actually significant.

Note

The note in "Task 01" in the "Tasks for Exercise 09: Factorial Designs"—see below — was, in this case, a trap intended to make you think.

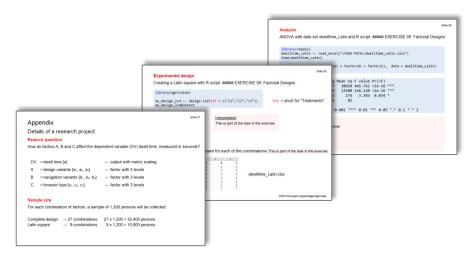
The differences are precisely symmetrical in the case of full factorial vs. fractional factorial.



Suggested solution 02

Research project – Latin square with R & Analysis

In the appendix of the lecture notes you will find details of a research project:



First, read through all the 3 slides in the appendix.

Use R to create the missing Latin square on Slide 28 and interpret its output. Use R to replicate the ANOVA shown on Slide 29 and interpret its output.

Use key words to interpret the Latin square and the ANOVA.

Describe the results – insert also the R-code, the R-output and if necessary, R-plots.

Suggested answers to the questions - Experimental design

```
library(agricolae)

my_design_lsd <- design.lsd(trt = c("c1","c2","c3"))
my_design_lsd$sketch</pre>
```

The column names of the Latin square ([,1] [,2] [,3]) correspond to the levels of factor A. The row names of the Latin square ([1,] [2,] [3,]) correspond to the levels of factor B. The content in the cells of the Latin square ("c1", "c2", "c3") correspond to the levels of factor C.

```
[,1] [,2] [,3] [,1] [,2] [,3] \rightarrow Levels of factor A [1,] [2,] [3,] \rightarrow Levels of factor B [2,] "c2" "c1" "c3" "c2" "c1","c2","c3" \rightarrow Levels of factor C
```

Suggested answers to the questions – Sampling

A sample with n = 1,200 is drawn for each of the combinations $[a_1, b_1, c_3], [a_2, b_1, c_2], \dots$

Suggested answers to the questions – Analysis

```
Df Sum Sq Mean Sq F value Pr(>F)
factor(A)
                          38028 465.761 <2e-16
               2
                  76057
factor(B)
               2
                  27159
                          13580 166.320 <2e-16 ***
factor(C)
                                  3.383
                                        0.034
               2
                    552
                            276
Residuals
           10338 844075
                             82
               0 (***, 0.001 (**, 0.05 (., 0.1 (, 1
Signif. codes:
```

There is a main effect of A (levels 1, 2, 3) on DV, F(2, 10338) = 465.761, p = .000. There is a main effect of B (levels 1, 2, 3) on DV, F(2, 10338) = 166.320, p = .000. There is a main effect of C (levels 1, 2, 3) on DV, F(2, 10338) = 3.383, p = .034.

<u>Note</u>

With p = .034 the factor C is relatively close to the significance limit of p = .050.