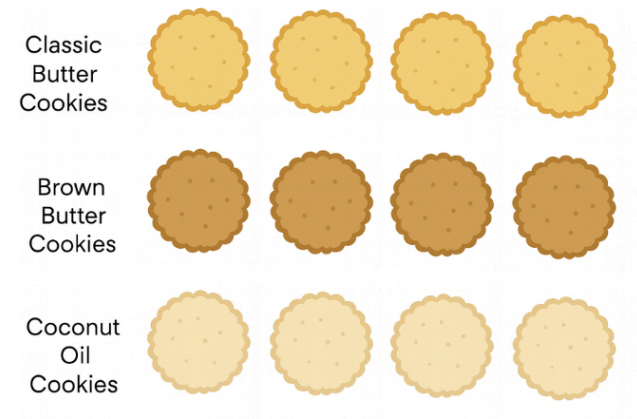


Module 5: Randomized Complete Block Designs (RCBD)

Designing a RCBD

Example 5.1: Cookies in Ovens

A bakery wants to compare three different cookie recipes for a new product line. The recipes differ in type of fat used: Classic Butter, Brown Butter, and Coconut Oil. Cookie quality is evaluated based on texture scores from a tasting panel.



Example 5.1: The Complication

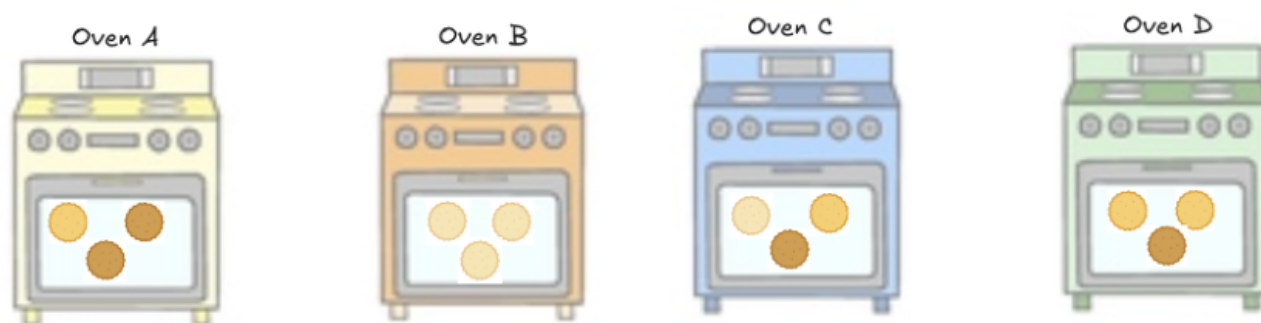
But...the bakery uses multiple ovens

The ovens are known to differ slightly in temperature calibration and air circulation. These differences can affect how cookies bake. For example, some ovens may run hotter, producing crispier cookies, while others bake more slowly and yield softer cookies. Although one oven may be relatively consistent, overall consistency between ovens is unlikely.



Suppose there were *no restrictions* on recipe randomization

Completely Randomized Design (CRD)



- Suppose Oven A runs slightly hotter than the others. How will this affect the results if that oven happens to get more Brown Butter trays?
- Suppose Oven B runs cooler and happens to bake most of the Coconut Oil trays. How will this affect the results?

Where the CRD breaks down

Key Idea: We can't separate recipe effects from oven effects.

CRD assumes:

- Experimental units are homogeneous

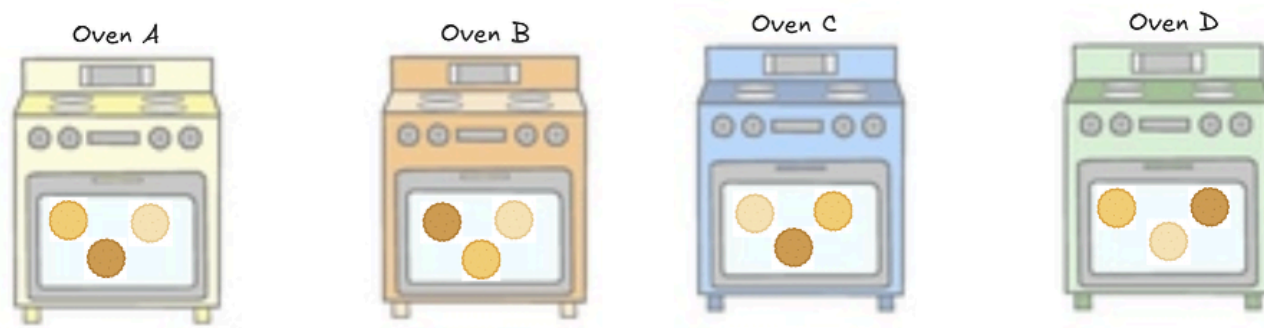
But here:

- Trays within an oven are similar
- Trays across ovens are *not* similar

The Solution: Block on oven

Randomized Complete Block Design (RCBD)

Why ovens? big source of variation, not scientifically interesting, known before the experiment



Key Idea: Block on what you can't control but can identify.

What 'Randomized Complete' means

Elements of a RCBD

- **Block:** This is a homogeneous group of experimental units. A RCBD consists of first sorting the experimental units into blocks.
- **Complete:** Each block consists of one complete replication of the set of treatments. Therefore, each treatment will show up once within each block.
- **Randomized:** The treatments are randomly assigned to experimental units separately within each block.

Why Block?

Advantages

- Effective blocking can lead to reduced experimental error and more precise estimates of treatment effects.
- The RCBD can accommodate any number of treatments and replications.
- The statistical analysis is relatively simple.

Disadvantages

- The degrees of freedom for experimental error are not as large as for a CRD.

Example 5.1: Study Structure

Treatment structure

One-way treatment structure with cookie recipes (3 levels – classic butter, brown butter, and coconut oil).

Experimental structure

Cookie recipe treatments are randomly assigned to trays (e.u.) in a RCBD (randomized complete block design) with $r = 4$ replications (blocks) where the blocking factor is oven. The texture score is recorded for each tray (m.u.).

Designing a RCBD

e.u. [Tray]	Block [Oven]			
	Oven A [runs hot]	Oven B [has a cool spot]	Oven C [dries cookies]	Oven D [new and even]
1				
2				
3				

R: Designing a RCBD

```
1 library(tidyverse)
2 library(edibble)
3
4 des <- design("Cookies in Ovens") |>
5   set_units(oven = c("Oven A", "Oven B", "Oven C", "Oven D"),
6             tray = nested_in(oven, 3)) |>
7   set_trts(recipe = c("Classic Butter", "Brown Butter", "Coconut Oil")) |>
8   allot_trts(recipe ~ tray) |>
9   assign_trts("random")
10
11 cookie_data <- serve_table(des)
12 cookie_data$texture <- NA
13 cookie_data
```

```
# Cookies in Ovens
# An edibble: 12 x 4
   oven   tray      recipe texture
  <U(4)> <U(12)>    <T(3)>
  <chr>  <chr>      <chr> <lgl>
1 Oven A tray01    Brown Butter NA
2 Oven A tray02    Classic Butter NA
3 Oven A tray03    Coconut Oil NA
4 Oven B tray04    Classic Butter NA
5 Oven B tray05    Coconut Oil NA
6 Oven B tray06    Brown Butter NA
7 Oven C tray07    Coconut Oil NA
8 Oven C tray08    Brown Butter NA
9 Oven C tray09    Classic Butter NA
10 Oven D tray10    Brown Butter NA
11 Oven D tray11    Classic Butter NA
12 Oven D tray12    Coconut Oil NA
```

R: Check your RCBD

Does each oven have 3 trays?

```
1 cookie_data |>  
2 count(oven)
```

```
# Cookies in Ovens  
# An edible: 4 x 2  
oven      n
```

```
<chr> <int>  
1 Oven A      3  
2 Oven B      3  
3 Oven C      3  
4 Oven D      3
```

Does each recipe appear r = 4 times?

```
1 cookie_data |>  
2 count(recipe)
```

```
# Cookies in Ovens  
# An edible: 3 x 2  
recipe     n
```

```
<chr>      <int>  
1 Brown Butter      4  
2 Classic Butter     4  
3 Coconut Oil        4
```

Does each recipe appear *once* in each oven?

```
1 cookie_data |>  
2 count(oven, recipe)
```

```
# Cookies in Ovens  
# An edible: 12 x 3  
oven  recipe      n
```

```
<chr> <chr>      <int>  
1 Oven A Brown Butter      1  
2 Oven A Classic Butter    1  
3 Oven A Coconut Oil       1  
4 Oven B Brown Butter      1  
5 Oven B Classic Butter    1  
6 Oven B Coconut Oil       1  
7 Oven C Brown Butter      1  
8 Oven C Classic Butter    1  
9 Oven C Coconut Oil       1  
10 Oven D Brown Butter     1  
11 Oven D Classic Butter    1  
12 Oven D Coconut Oil      1
```

JMP: Designing a RCBD

▼ DOE > Custom Design

▼ Custom Design

▼ Responses

Add Response ▼ Remove Number of Responses...

Response Name	Goal	Lower Limit	Upper Limit	Importance	Lower Detection Li
texture	Maximize ▼

▼ Factors

Add Factor ▼ Remove Add N Factors 1

Name	Role	Changes	Values	Units
▼ Recipe	Categorical	Easy ▼	Classic Butter Brown Butter Coconut Oil	
█ Oven	Blocking	Easy	1	

▼ Covariate/Candidate Runs

Select Covariate Factors Load a set of candidate runs for covariates from the current data table.

Specify Factors

Add a factor by clicking the Add Factor button. Double click on a factor name or level to edit it.

Continue

JMP: Designing a RCBD

Factors

Add Factor ▼ Remove Add N Factors

Name	Role	Changes	Values	Units		
▼ recipe	Categorical	Easy ▼	Classic Butter	Brown Butter	Coconut Oil	
oven	Blocking	Easy	1	2	3	4

Design

Run	recipe	oven
1	Classic Butter	2
2	Classic Butter	1
3	Coconut Oil	4
4	Classic Butter	4
5	Brown Butter	4
6	Classic Butter	3
7	Coconut Oil	1
8	Coconut Oil	3
9	Coconut Oil	2
10	Brown Butter	3
11	Brown Butter	2
12	Brown Butter	1

Design Generation

Number of Replicate Runs:

Number of Runs:

☐ Minimum 4

☐ Default 9

☒ User Specified

Make Design

Data Table Options

☐ Save X Matrix

☐ Simulate Responses

☐ Include Run Order Column

Run Order: Randomize within Blocks ▼

Make Table

JMP: Designing a RCBD

The screenshot shows the JMP Custom Design window. On the left, the 'Custom Design' tab is active, with a list of design steps: Design, Criterion, Model, Evaluate Design, Generalized Regression, and DOE Dialog. Below this, the 'Columns (3/0)' section shows three factors: 'recipe *' (red bar), 'oven *' (red bar), and 'texture *' (blue bar). A 'Filter' search box and a 'Tags' section are also visible. The main table displays 12 runs of the design, with columns for run number, recipe, oven, and texture. The design is a Randomized Complete Block Design (RCBD) with 3 treatments (recipe) and 4 replicates (oven).

		recipe	oven	texture		
1	Coconut Oil	1		•		
2	Classic Butter	1		•		
3	Brown Butter	1		•		
4	Brown Butter	2		•		
5	Coconut Oil	2		•		
6	Classic Butter	2		•		
7	Classic Butter	3		•		
8	Brown Butter	3		•		
9	Coconut Oil	3		•		
10	Classic Butter	4		•		
11	Coconut Oil	4		•		
12	Brown Butter	4		•		

Then use [Tables > Summary](#) to check RCBD.