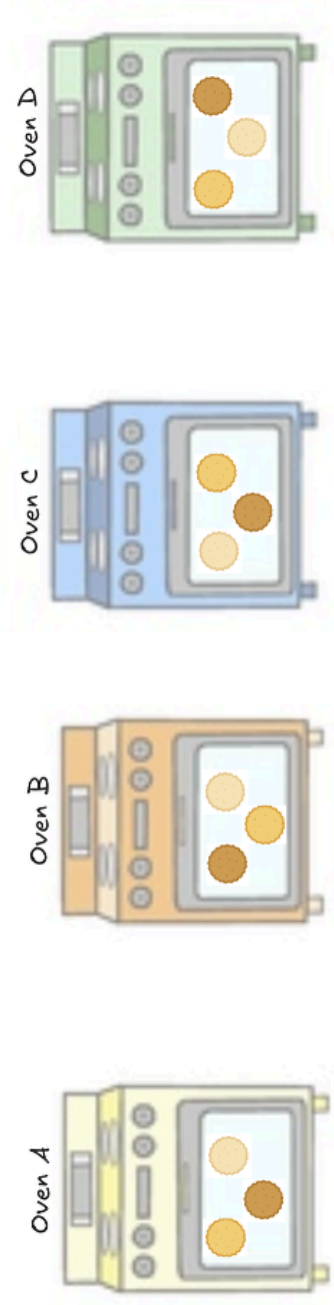


# Module 5: Randomized Complete Block Designs (RCBD)

Analyzing a RCBD

## Example 5.1: Cookie Recipes



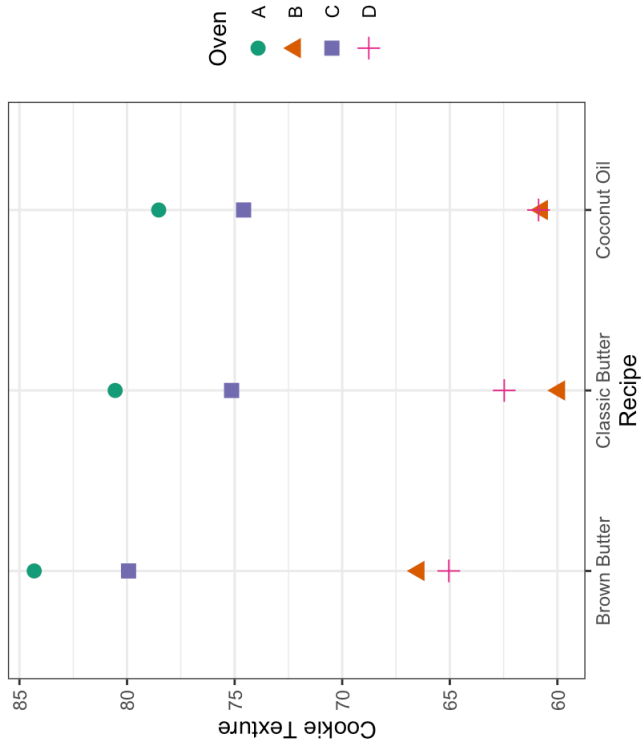
**Treatment Structure:** One-way with recipe (3-levels: classic butter, brown butter, coconut oil)

**Design Structure:** Recipe randomly assigned to tray (e.u.) in an RCBD with  $r = 4$  ovens (blocking factor). Texture recorded for each tray (m.u.).

# Cookie Recipes

```
1 cookie_data <- read_csv("data/cookie_rcbd_data.csv")
2 cookie_data
```

```
# A tibble: 12 × 3
  oven recipe texture
<chr> <chr> <dbl>
1 A Classic Butter 80.6
2 A Brown Butter 84.3
3 A Coconut Oil 78.5
4 B Classic Butter 59.9
5 B Brown Butter 66.4
6 B Coconut Oil 60.7
7 C Classic Butter 75.1
8 C Brown Butter 79.9
9 C Coconut Oil 74.6
10 D Classic Butter 62.5
11 D Brown Butter 65.0
12 D Coconut Oil 60.9
```



# What changes from a CRD?

If we analyze *ignoring* the effect of oven...

## Incorrect Analysis: CRD

$$y_{ij} = \mu + \tau_i + \epsilon_{ij} \text{ where } \epsilon_{ij} \text{ iid } \sim N(0, \sigma^2)$$

```
1 oven_crd_mod <- lm(texture ~ recipe, data = cookie_data)
2 anova(oven_crd_mod)
```

### Analysis of Variance Table

```
Response: texture
          Df Sum Sq Mean Sq F value Pr(>F)
recipe     2  63.82   31.910   0.3459  0.7166
Residuals  9  830.24   92.248
```

# Statistical Effects Model

$$y_{ij} = \mu + \tau_i + \rho_j + \epsilon_{ij} \text{ where } \epsilon_{ij} \text{ iid } \sim N(0, \sigma^2)$$

for  $i = 1, 2, 3$  and  $j = 1, 2, 3, 4$

where:

- $y_{ij}$  is the observed texture of the cookie tray baked in the  $j^{th}$  oven with the  $i^{th}$  recipe.
- $\mu$  is the overall mean texture
- $\tau_i$  is the effect of the  $i^{th}$  recipe
- $\epsilon_{ij}$  is the experimental error associated with the cookie tray baked in the  $j^{th}$  oven with the  $i^{th}$  recipe.

## What Blocking does to the variability

$$SST = SST_{rt} + SS_{Blk} + SSE$$

- $SST = \sum_i \sum_j (y_{ij} - \bar{y}_{..})^2$
- $SST_{rt} = r \sum_i (\bar{y}_{i.} - \bar{y}_{..})^2$
- $SS_{Blk} = t \sum_j (\bar{y}_{.j} - \bar{y}_{..})^2$
- $SSE = t \sum_j (y_{ij} - \bar{y}_{i.} - \bar{y}_{.j} + \bar{y}_{..})^2$

# ANOVA Table

Source of Variation	DF	SS	MS	F
Block	$r - 1$	SSBlk	MSBlk	MSBlk/MSE
Treatment	$t - 1$	SSTrt	MSTrt	MSTrt/MSE
Block x Treatment → error	$(r - 1)(t - 1)$	SSE	MSE → $\sigma^2$	

<b>Total (N = rt)</b>	<b>N - 1</b>	SST
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Blocking is a *design tool*, not usually a research question.

# Skeleton ANOVA

Source of Variation	DF
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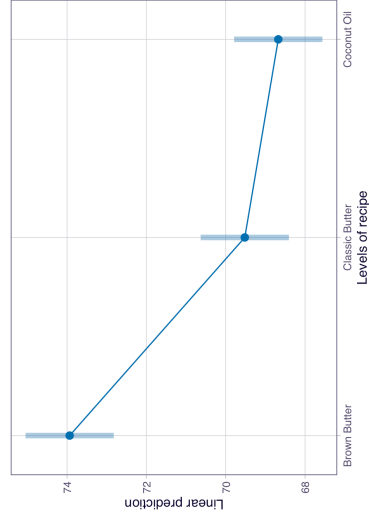
# R: RCBD Analysis

```
1 oven_rcbd_mod <- lm(texture ~ recipe + oven, data = cookie_data)
2 anova(oven_rcbd_mod)
```

## Analysis of Variance Table

```
Response: texture
Df Sum Sq Mean Sq F value    Pr(>F)
recipe 2  63.82  31.910  38.607 0.0003749 ***
oven   3 825.28 275.092 332.829 4.651e-07 ***
Residuals 6  4.96   0.827
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
1 emmip(oven_rcbd_mod, ~ recipe, CI = T)
```



```
1 emmeans(oven_rcbd_mod, ~ recipe) |>
2 cld(Letters = LETTERS, decreasing = T, adjust = "tukey")
```

recipe	emmean	SE	df	lower.CL	upper.CL	.group
Brown Butter	73.9	0.455	6	72.4	75.4	A
Classic Butter	69.5	0.455	6	68.0	71.0	B
Coconut Oil	68.7	0.455	6	67.2	70.2	B

Results are averaged over the levels of: oven

Confidence level used: 0.95

P value adjustment: sidak method for 3 estimates

P value adjustment: tukey method for comparing a family of 3 estimates  
significance level used: alpha = 0.05

NOTE: If two or more means share the same grouping symbol,

then we cannot show them to be different.

But we also did not show them to be the same.

# JMP: RCBD Analysis

Model Specification

Select Columns

3 Columns

oven recipe texture

Fit Model

Pick Role Variables

Y

texture

optional

Weight

optional numeric

Freq

optional numeric

Validation

optional numeric

By

optional

Personality:

Standard Least Squares

Emphasis:

Effect Leverage

Run

Keep dialog open

Construct Model Effects

Add recipe oven

Cross

Nest

Macros

Degree 2

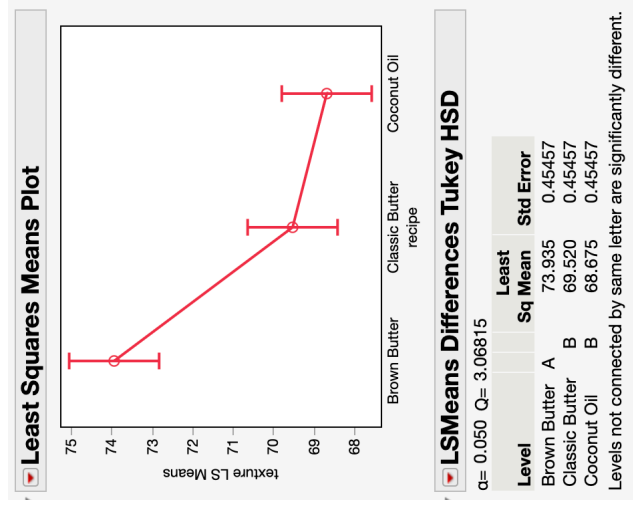
Attributes

Transform

No Intercept

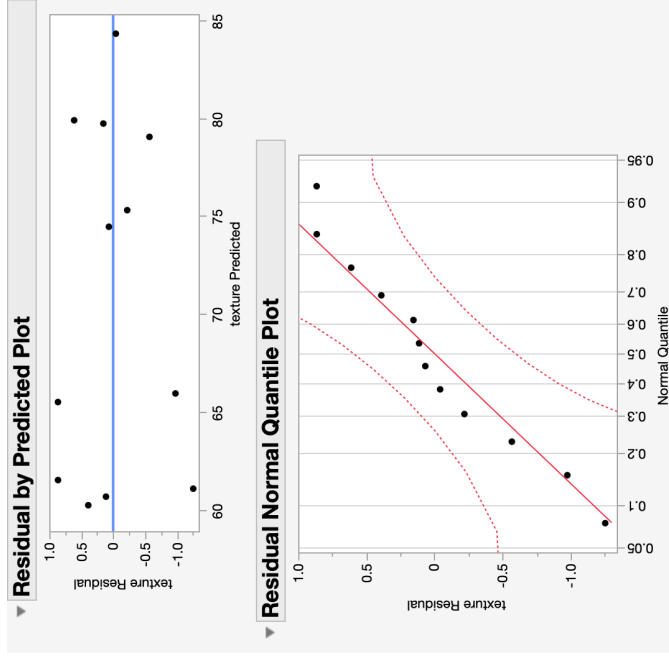
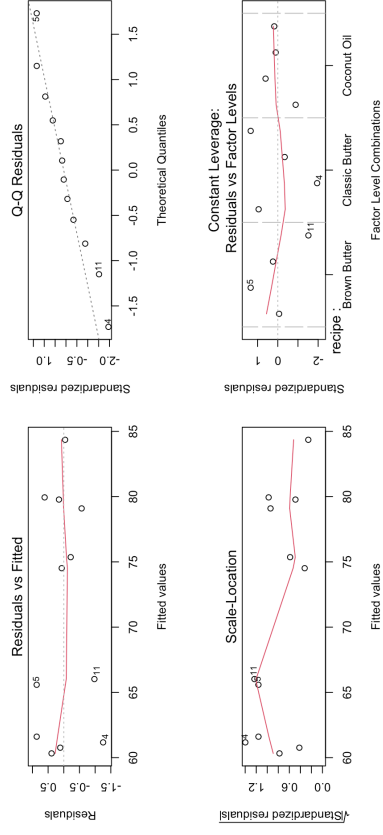
Analysis of Variance				
Source	DF	Sum of Squares	Mean Square	F Ratio
Model	5	889.09649	177.819	215.1405
Error	6	4.95916	0.827	Prob > F
C. Total	11	894.05565		<.0001*

Parameter Estimates				
Source	Nparm	DF	Sum of Squares	F Ratio
recipe	2	2	63.81924	38.6069
oven	3	3	825.27725	332.8295
				Prob > F
				0.0004*
				<.0001*



# Check Model Assumptions – $\epsilon_{ij} \text{iid} \sim N(0, \sigma^2)$

```
1 par(mfrow = c(2,2))
2 plot(oven_rcbd_mod)
```



# RCBD vs CRD

If we ignore ovens (CRD):

- Oven variability inflates MSE
- Harder to detect recipe differences

Analysis of Variance Table

```
Response: texture
Df Sum Sq Mean Sq F value Pr(>F)
recipe 2 63.82 31.910 0.3459 0.7166
Residuals 9 830.24 92.248
```

With RCBD:

- Oven variability is removed from error
- Smaller MSE

More power (*if blocks are different*)

Analysis of Variance Table

```
Response: texture
Df Sum Sq Mean Sq F value Pr(>F)
recipe 2 63.82 31.910 38.607 0.0003749 ***
oven 3 825.28 275.092 332.829 4.651e-07 ***
Residuals 6 4.96 0.827
---
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

# Why is this model additive? (no interaction)

- Each recipe appears *once* per oven
- No replication within recipe-oven combinations
- Interaction is not estimable

Recall: The experimental error comes from the block x treatment term.

