

ADA2: Class 08, Ch 05a Paired Experiments and Randomized Block Experiments

Randomized complete block design (RCBD)

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Menu items

A fast food franchise is test marketing 3 new menu items. To find out if they have the same popularity, 6 franchisee restaurants are randomly chosen for participation in the study. In accordance with the randomized block design, each restaurant will be test marketing all 3 new menu items. Furthermore, a restaurant will test market only one menu item per week, and it takes 3 weeks to test market all menu items. The testing order of the menu items for each restaurant is randomly assigned as well. The table represents the sales figures of the 3 new menu items in a restaurant after a week of test marketing.

Randomized Complete Block Design (RCBD)

Following the in-class assignment this week, perform a complete RCBD analysis.

1. (2 p) Reshape and plot the data, describe relationships of Sales between Items and Restaurants
2. (0 p) Fit model
3. (3 p) Assess model assumptions
4. (2 p) State and interpret the hypothesis test for difference in Item mean sales
5. (2 p) If appropriate, perform pairwise comparisons with Tukey HSD correction
6. (1 p) What is your recommendation to the Franchise?

Data

```
library(erikmisc)
library(tidyverse)

# read the data
dat_food <- read.table(text="
Restaurant Item1 Item2 Item3
A           31    27    24
B           31    28    31
C           45    29    46
D           21    18    48
E           42    36    46
F           32    17    40
", header = TRUE) %>%
  as_tibble()
```

(2 p) Reshape and plot the data, describe relationships of Sales between Items and Restaurants

The code below will get you started with reshaping the data. The rest is up to you!

```
dat_food_long <-
  dat_food %>%
  pivot_longer(
    cols      = starts_with("Item")
  , names_to  = "Item"
  , values_to = "Sales"
  ) %>%
  mutate(
    Item      = factor(Item)
  , Restaurant = factor(Restaurant)
  )

str(dat_food_long)
```

```
tibble [18 x 3] (S3: tbl_df/tbl/data.frame)
 $ Restaurant: Factor w/ 6 levels "A","B","C","D",...: 1 1 1 2 2 2 3 3 3 4 ...
 $ Item      : Factor w/ 3 levels "Item1","Item2",...: 1 2 3 1 2 3 1 2 3 1 ...
 $ Sales     : int [1:18] 31 27 24 31 28 31 45 29 46 21 ...
```

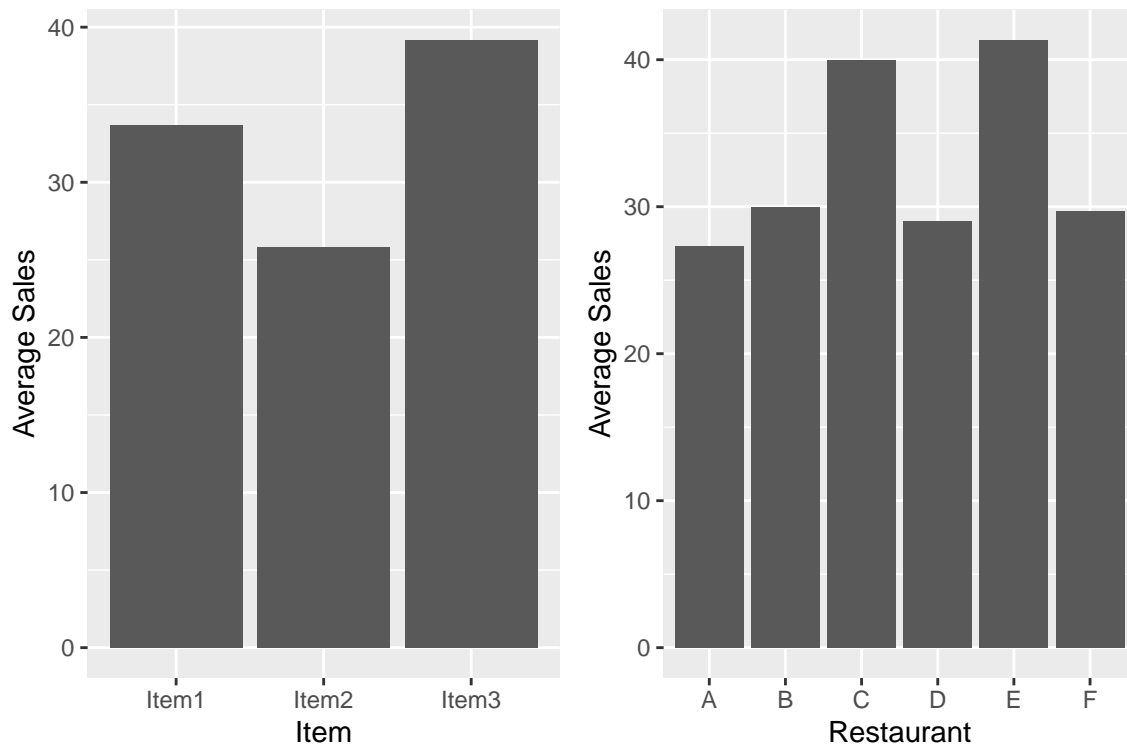
```
# Group means
m_dat_b <-
  dat_food_long %>%
  group_by(Item) %>%
  summarize(
    m = mean(Sales)
  )
m_dat_b
```

Item	m
Item1	33.66667
Item2	25.83333
Item3	39.16667

```
m_dat_c <-
  dat_food_long %>%
  group_by(Restaurant) %>%
  summarize(
    m = mean(Sales)
  )
m_dat_c
```

Restaurant	m
A	27.33333
B	30.00000
C	40.00000
D	29.00000
E	41.33333
F	29.66667

```
library(ggpubr)
gg1 <- ggplot(data = m_dat_b) +
  geom_bar(aes(x = Item, y = m), stat = "identity") +
  ylab("Average Sales")
gg2 <- ggplot(data = m_dat_c) +
  geom_bar(aes(x = Restaurant, y = m), stat = "identity") +
  ylab("Average Sales")
ggarrange(gg1, gg2, nrow = 1, label.x = "Average Sales")
```



Based on means alone, looks like Item 2 leads to lowest sales, Item 1 leads to a medium amount of sales, and Item 3 leads to the highest sales. Restaurants A, B, D, and F all appear to have the same sales, whereas restaurants C and E have higher sales, similar to one another.

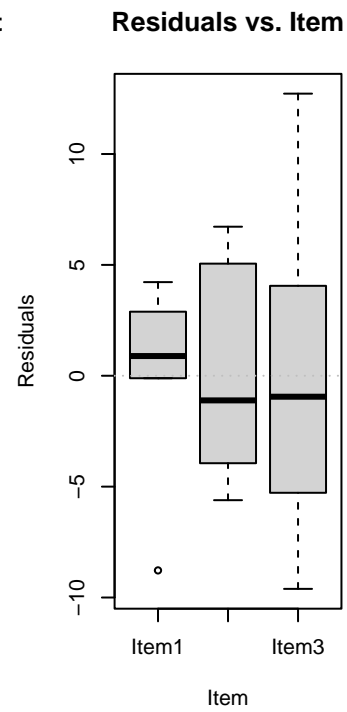
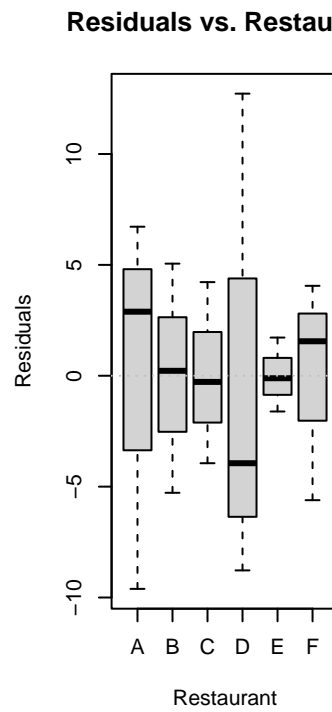
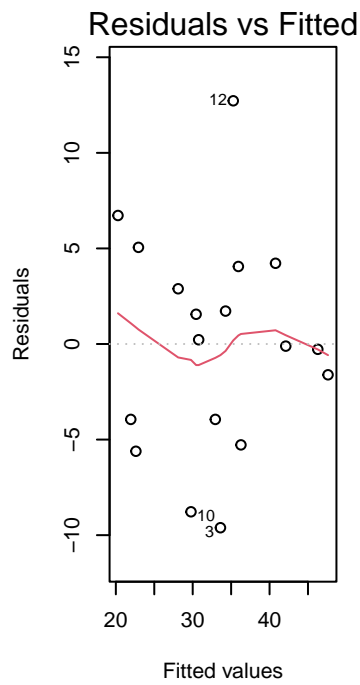
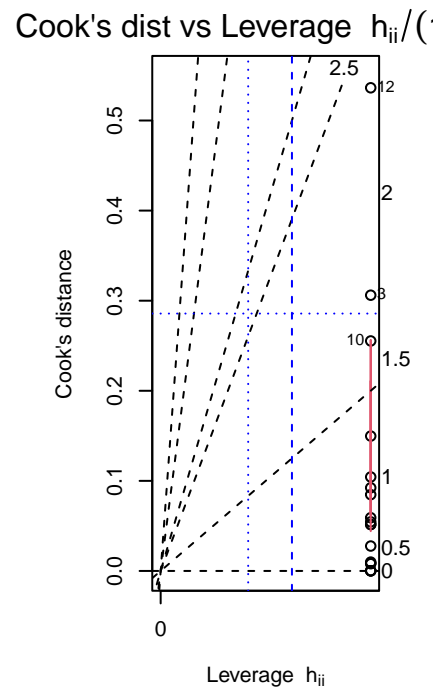
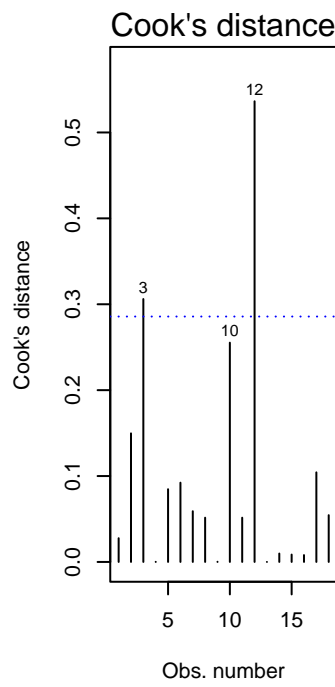
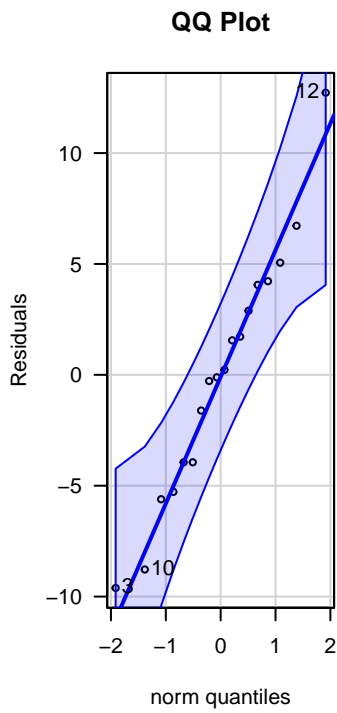
(0 p) Fit model

```
mod1 <- lm(Sales ~ Restaurant + Item, data = dat_food_long)
car::Anova(mod1)
```

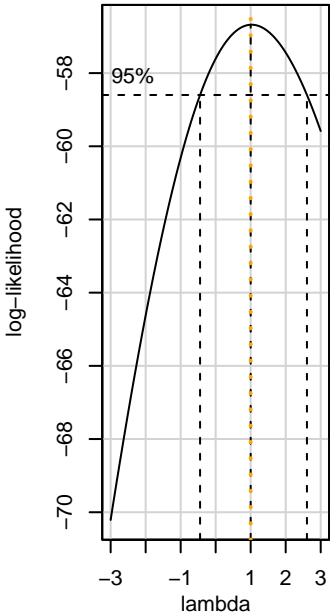
	Sum Sq	Df	F value	Pr(>F)
Restaurant	559.7778	5	2.060953	0.1546506
Item	538.7778	2	4.959092	0.0318971
Residuals	543.2222	10	NA	NA

(3 p) Assess model assumptions

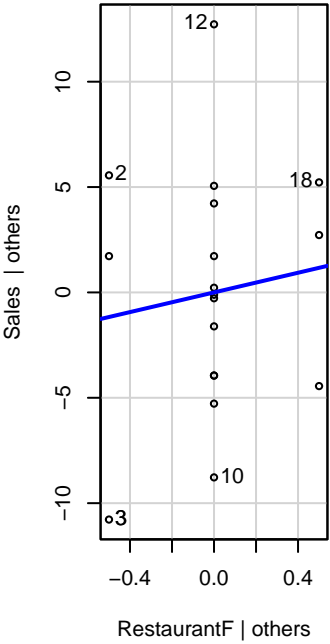
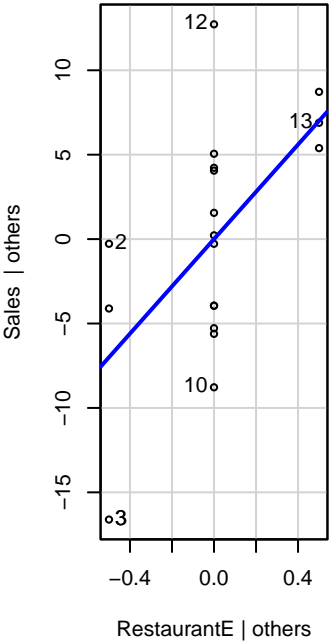
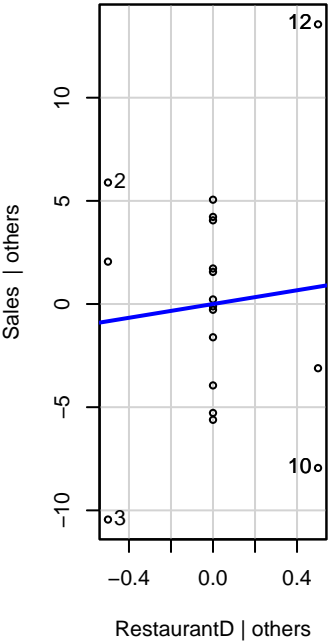
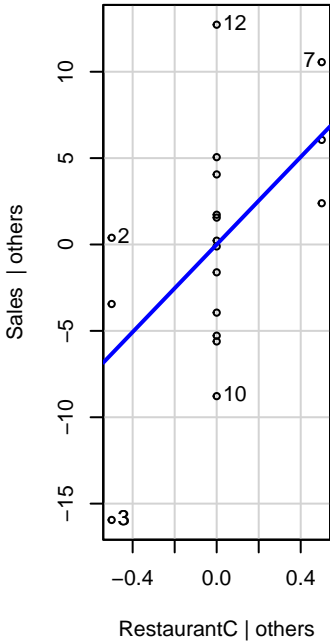
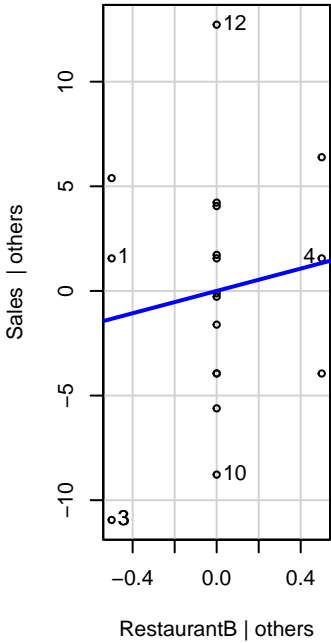
```
e_plot_lm_diagnostics(mod1, sw_plot_set = "simpleAV")
```

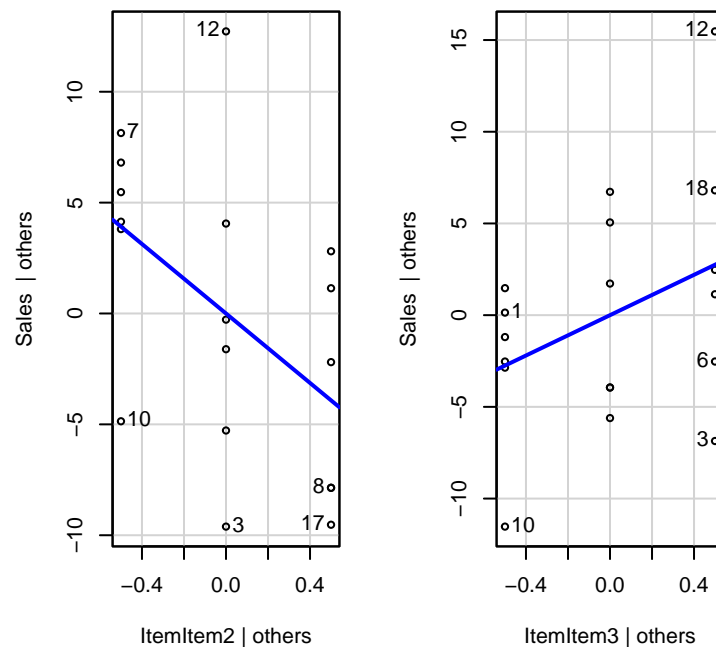


Box-Cox power transformati



lambda of 1 is none (y^1); 0 is $\log(y)$ of an





There are a few violations here. In particular, it looks like there may be an outlier (Observation 12) and heterogeneity of variance across both Items and Restaurants. The residuals do look normally distributed, and the Box-Cox profile indicates no transformation of Sales is warranted.

(2 p) State and interpret the hypothesis test for difference in Item mean sales

The hypothesis test is:

$H_0: \mu_1 = \mu_2 = \mu_3$ vs H_1 : not all means equal.

Rejection of this null hypothesis indicates that at least one pair of true means are not equal.

(2 p) If appropriate, perform pairwise comparisons with Tukey HSD correction

Yes, it's appropriate!

```
aov1 <- aov(Sales ~ Item + Restaurant, data = dat_food_long)
TukeyHSD(aov1)
```

```
Tukey multiple comparisons of means
95% family-wise confidence level
```

```
Fit: aov(formula = Sales ~ Item + Restaurant, data = dat_food_long)
```

```
$Item
      diff      lwr      upr    p adj
Item2-Item1 -7.833333 -19.498312  3.831645 0.2061328
Item3-Item1  5.500000  -6.164978 17.164978 0.4305650
Item3-Item2 13.333333   1.668355 24.998312 0.0262709
```

```
$Restaurant
      diff      lwr      upr    p adj
B-A    2.666667 -18.235335 23.568669 0.9971568
C-A   12.666667  -8.235335 33.568669 0.3559810
D-A    1.666667 -19.235335 22.568669 0.9997013
E-A   14.000000  -6.902002 34.902002 0.2676025
F-A    2.333333 -18.568669 23.235335 0.9984869
```

C-B	10.0000000	-10.902002	30.902002	0.5815802
D-B	-1.0000000	-21.902002	19.902002	0.9999758
E-B	11.3333333	-9.568669	32.235335	0.4620926
F-B	-0.3333333	-21.235335	20.568669	0.9999999
D-C	-11.0000000	-31.902002	9.902002	0.4909865
E-C	1.3333333	-19.568669	22.235335	0.9998999
F-C	-10.3333333	-31.235335	10.568669	0.5508770
E-D	12.3333333	-8.568669	33.235335	0.3809386
F-D	0.6666667	-20.235335	21.568669	0.9999968
F-E	-11.6666667	-32.568669	9.235335	0.4340602

The Tukey test shows a significant difference between the mean Sales for Item 3 and Item 2 at α 0.05, where Item 3 has average sales 13.33 higher than for Item 2, but the other two pairwise comparisons do not show significant differences.

We also see no significant differences for pairwise comparisons among any restaurants.

(1 p) What is your recommendation to the Franchise?

I recommend using Item 3, since it has the highest sales. Although Item 3 does not show a significantly higher mean than Item 1, this could simply be due to lack of power.