

Statistical Methods for Discrete Response, Time Series, and Panel Data (W271): Lab 2

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Strategic Placement of Products in Grocery Stores

Answer **Question 12 of chapter 3 (on page 189 and 190)** of Bilder and Loughin's *"Analysis of Categorical Data with R"*. Here is the background of this analysis, taken as an excerpt from this question:

In order to maximize sales, items within grocery stores are strategically placed to draw customer attention. This exercise examines one type of item—breakfast cereal. Typically, in large grocery stores, boxes of cereal are placed on sets of shelves located on one side of the aisle. By placing particular boxes of cereals on specific shelves, grocery stores may better attract customers to them. To investigate this further, a random sample of size 10 was taken from each of four shelves at a Dillons grocery store in Manhattan, KS. These data are given in the `cereal_dillons.csv` file. The response variable is the shelf number, which is numbered from bottom (1) to top (4), and the explanatory variables are the sugar, fat, and sodium content of the cereals.

Since this question has part a to h, please write down each of the questions in your report so that I can easily follow your answer.

In order to maximize sales, items within grocery stores are strategically placed to draw customer attention. This exercise examines one type of item—breakfast cereal. Typically, in large grocery stores, boxes of cereal are placed on sets of shelves located on one side of the aisle. By placing particular boxes of cereals on specific shelves, grocery stores may better attract customers to them. To investigate this further, a random sample of size 10 was taken from each of four shelves at a Dillons grocery store in Manhattan, KS. These data are given in the `cereal_dillons.csv` file. The response variable is the shelf number, which is numbered from bottom (1) to top (4), and the explanatory variables are the sugar, fat, and sodium content of the cereals. Using these data, complete the following:

```
cereal <- read.csv('cereal_dillons.csv', header=TRUE)
cereal
```

##	ID	Shelf	Cereal	size_g	sugar_g
## 1	1	1	Kellogg's Razzle Dazzle Rice Crispies	28	10
## 2	2	1	Post Toasties Corn Flakes	28	2
## 3	3	1	Kellogg's Corn Flakes	28	2
## 4	4	1	Food Club Toasted Oats	32	2
## 5	5	1	Frosted Cheerios	30	13
## 6	6	1	Food Club Frosted Flakes	31	11
## 7	7	1	Capn Crunch	27	12
## 8	8	1	Capn Crunch's Peanut Butter Crunch	27	9
## 9	9	1	Post Honeycomb	29	11
## 10	10	1	Food Club Crispy Rice	33	2
## 11	11	2	Rice Crispies Treats	30	9
## 12	12	2	Kellogg's Smacks	27	15
## 13	13	2	Kellogg's Froot Loops	32	15
## 14	14	2	Capn Crunch's Peanut Butter Crunch	27	9
## 15	15	2	Cinnamon Grahams	30	11
## 16	16	2	Marshmallow Blasted Froot Loops	30	16

## 17 17	2	Koala Coco Krunch	30	13
## 18 18	2	Food Club Toasted Oats	33	10
## 19 19	2	Cocoa Pebbles	29	13
## 20 20	2	Oreo O's	27	11
## 21 21	3	Food Club Raisin Bran	54	17
## 22 22	3	Post Honey Bunches of Oats	30	6
## 23 23	3	Rice Chex	31	2
## 24 24	3	Kellogg's Corn Pops	31	14
## 25 25	3	Post Morning Traditions - Raisin, Date, Pecan	54	14
## 26 26	3	Post Shredded Wheat Spoon Size	49	0
## 27 27	3	Basic 4	55	14
## 28 28	3	French Toast Crunch	30	12
## 29 29	3	Post Raisin Bran	59	20
## 30 30	3	Food Club Frosted Shreded Wheat	50	1
## 31 31	4	Total Raisin Bran	55	19
## 32 32	4	Food Club Wheat Crunch	60	6
## 33 33	4	Oatmeal Crisp Raisin	55	19
## 34 34	4	Food Club Bran Flakes	31	5
## 35 35	4	Cookie Crisp	30	12
## 36 36	4	Kellogg's All Bran Original	31	6
## 37 37	4	Food Club Low Fat Granola	55	14
## 38 38	4	Oatmeal Crisp Apple Cinnamon	55	19
## 39 39	4	Post Fruit and Fiber - Dates, Raisons, Walnuts	55	17
## 40 40	4	Total Corn Flakes	30	3
##	fat_g	sodium_mg		
## 1	0.0	170		
## 2	0.0	270		
## 3	0.0	300		
## 4	2.0	280		
## 5	1.0	210		
## 6	0.0	180		
## 7	1.5	200		
## 8	2.5	200		
## 9	0.5	220		
## 10	0.0	330		
## 11	1.5	190		
## 12	0.5	50		
## 13	1.0	150		
## 14	2.5	200		
## 15	1.0	230		
## 16	0.5	105		
## 17	1.0	170		
## 18	1.5	150		
## 19	1.0	160		
## 20	2.5	150		
## 21	1.0	280		
## 22	1.5	190		
## 23	0.0	290		

```
## 24  0.0      120
## 25  5.0      160
## 26  0.5       0
## 27  3.0     320
## 28  1.0     180
## 29  1.0     300
## 30  1.0       0
## 31  1.0     240
## 32  0.0     300
## 33  2.0     220
## 34  0.5     220
## 35  1.0     180
## 36  1.0      65
## 37  3.0     100
## 38  2.0     260
## 39  3.0     280
## 40  0.0     200
```

a

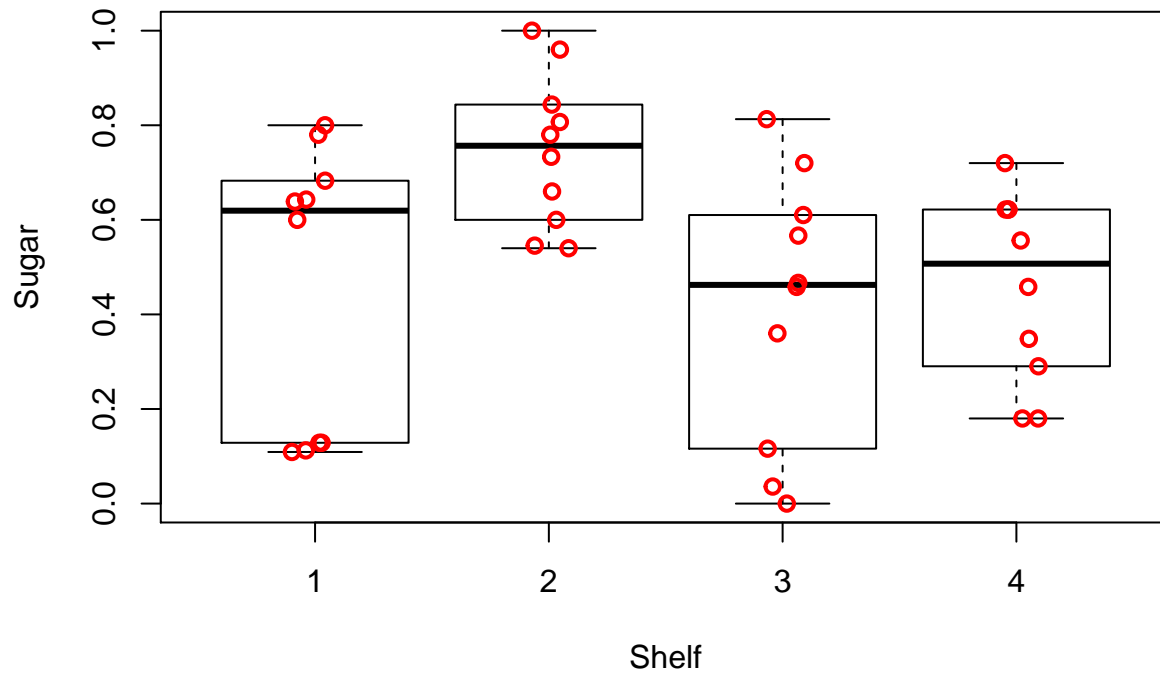
The explanatory variables need to be re-formatted before proceeding further. First, divide each explanatory variable by its serving size to account for the different serving sizes among the cereals. Second, re-scale each variable to be within 0 and 1.¹² Below is code we use to re-format the data after the data file is read into an object named `cereal`:

```
stand01 <- function(x) { (x - min(x))/(max(x) - min(x)) }
cereal2 <- data.frame(Shelf = cereal$Shelf, sugar =
  stand01(x = cereal$sugar_g/cereal$size_g), fat =
  stand01(x = cereal$fat_g/cereal$size_g), sodium =
  stand01(x = cereal$sodium_mg/cereal$size_g))
```

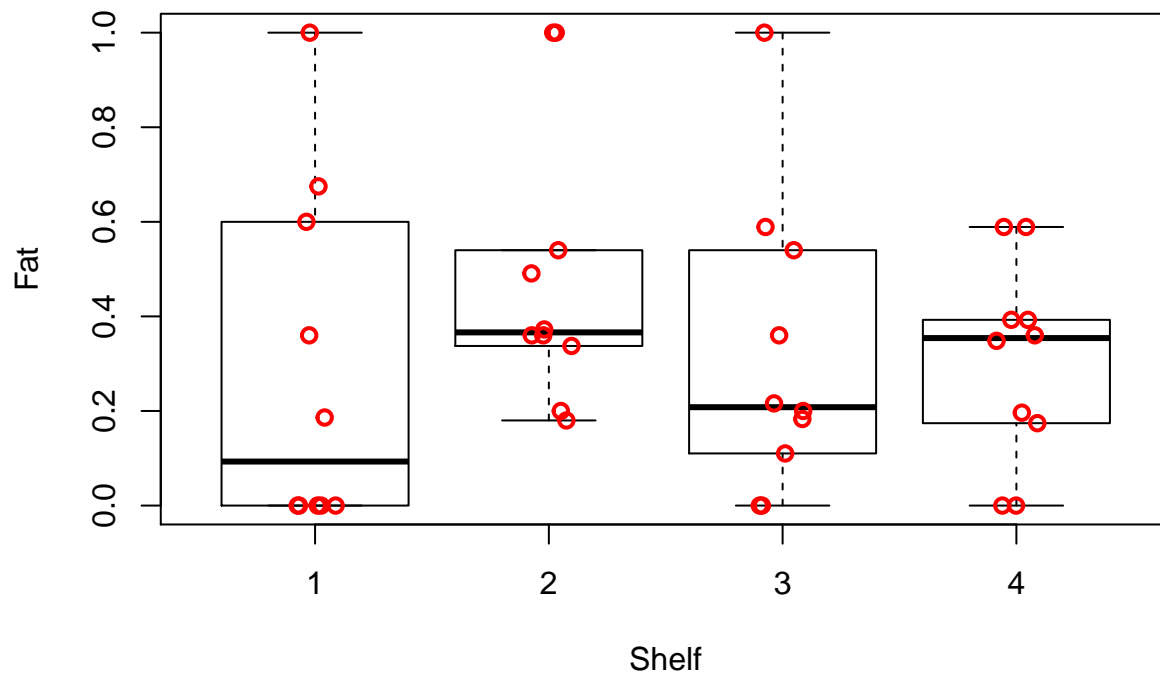
b

Construct side-by-side box plots with dot plots overlaid for each of the explanatory variables. Below is code that can be used for plots involving sugar:

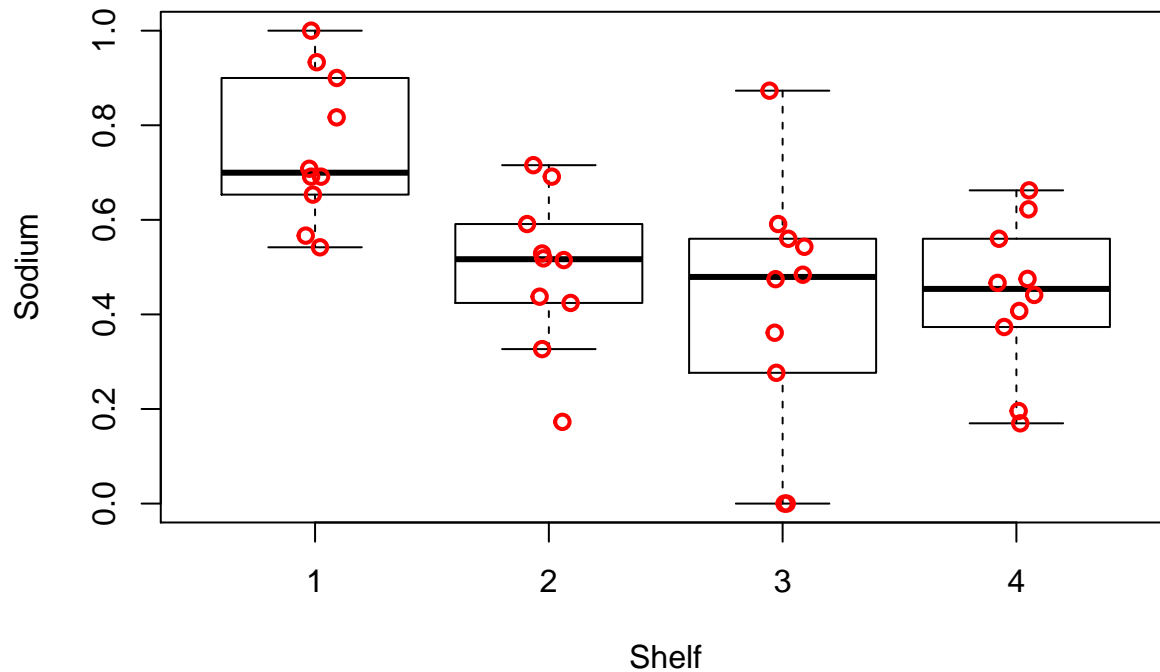
```
boxplot(formula = sugar ~ Shelf, data = cereal2, ylab = "Sugar", xlab = "Shelf", pars = list(
  stripchart(x = cereal2$sugar ~ cereal2$Shelf, lwd = 2, col = "red", method = "jitter", vertical
```



```
boxplot(formula = fat ~ Shelf, data = cereal2, ylab = "Fat", xlab = "Shelf", pars = list(outpch = 1, outcex = 1.2))
stripchart(x = cereal2$fat ~ cereal2$Shelf, lwd = 2, col = "red", method = "jitter", vertical = FALSE)
```



```
boxplot(formula = sodium ~ Shelf, data = cereal2, ylab = "Sodium", xlab = "Shelf", pars = list(outpch = 1, outcex = 1.2))
stripchart(x = cereal2$sodium ~ cereal2$Shelf, lwd = 2, col = "red", method = "jitter", vertical = FALSE)
```



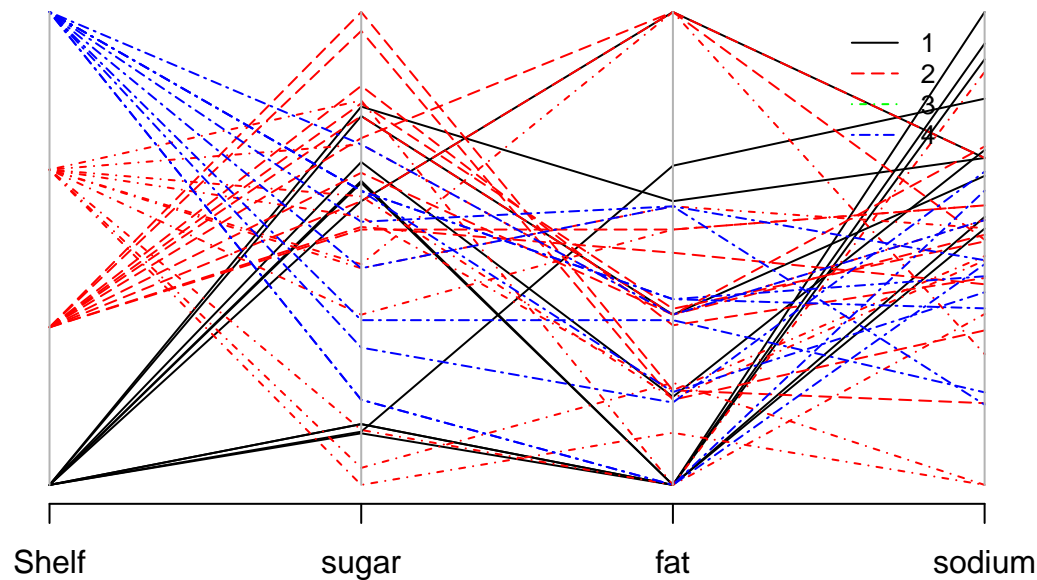
Also, construct a parallel coordinates plot for the explanatory variables and the shelf number. Discuss if possible content differences exist among the shelves.

```
library(package = MASS)
```

```
cereal2.colors<-ifelse(test = cereal2$Shelf==1, yes = "black",
  no = ifelse(test = cereal2$Shelf==2, yes = "red",
    ifelse(test = cereal2$Shelf==3, yes = "red", no = "blue")))

cereal2.lty<-ifelse(test = cereal2$Shelf==1, yes = "solid",
  no = ifelse(test = cereal2$Shelf==2, yes = "longdash",
    ifelse(test = cereal2$Shelf==3, yes = "dotdash", no = "twodash")))

parcoord(x = cereal2, col = cereal2.colors, lty = cereal2.lty) # Plot
legend(x = 3.5, y = 1, legend = c("1", "2", "3", "4"), lty = c("solid", "longdash", "dotdash",
  col=c("black", "red", "green", "blue"), cex=0.8, bty="n")
```



c

The response has values of 1, 2, 3, and 4. Under what setting would it be desirable to take into account ordinality. Do you think this occurs here?

It would be desirable to take into account ordinality when the variable has a natural ordering to their levels. In other words, if response levels can be arranged so that category 1 < category 2 < ... < category J in some conceptual scale of measurement (e.g., amount of agreement). Since the shelf has a natural ordering to their levels, bottom (1) to top (4), it would make sense to take into account ordinality.

d

Estimate a multinomial regression model with linear forms of the sugar, fat, and sodium variables. Perform LRTs to examine the importance of each explanatory variable.

```
library(package = MASS)
mod.fit.ord <- polr(formula = as.factor(Shelf) ~ sugar + fat + sodium, data = cereal2, method = "logistic")
summary(mod.fit.ord)
```

```
##
## Re-fitting to get Hessian
## Call:
## polr(formula = as.factor(Shelf) ~ sugar + fat + sodium, data = cereal2,
##       method = "logistic")
##
## Coefficients:
##           Value Std. Error  t value
## sugar  -1.61101    1.2830 -1.25565
```

```
## fat      -0.05123      0.9657 -0.05305
## sodium -4.85950      1.6302 -2.98094
##
## Intercepts:
##      Value   Std. Error t value
## 1|2 -4.7534   1.4837    -3.2037
## 2|3 -3.3435   1.3810    -2.4210
## 3|4 -1.9823   1.2867    -1.5407
##
## Residual Deviance: 98.52912
## AIC: 110.5291
```

```
library(package = car)
Anova(mod.fit.ord)
```

```
## Analysis of Deviance Table (Type II tests)
##
## Response: as.factor(Shelf)
##      LR Chisq Df Pr(>Chisq)
## sugar      1.6794 1  0.1950069
## fat         0.0028 1  0.9577007
## sodium    11.5685 1  0.0006708 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

e

Show that there are no significant interactions among the explanatory variables (including an interaction among all three variables).

```
library(package = car)
mod.fit.ord2 <- polr(formula = as.factor(Shelf) ~ sugar + fat + sodium
                     + sugar:fat + sugar:sodium + fat:sodium
                     + sugar:fat:sodium
                     , data = cereal2, method = "logistic")
Anova(mod.fit.ord2)
```

```
## Analysis of Deviance Table (Type II tests)
##
## Response: as.factor(Shelf)
##      LR Chisq Df Pr(>Chisq)
## sugar      1.1760 1  0.2781685
## fat         0.0419 1  0.8377311
## sodium    11.1699 1  0.0008314 ***
## sugar:fat    0.1014 1  0.7501457
## sugar:sodium  0.3945 1  0.5299556
## fat:sodium    0.2607 1  0.6096643
## sugar:fat:sodium 0.1077 1  0.7427907
```



```
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

f

Kellogg's Apple Jacks (<http://www.applejacks.com>) is a cereal marketed toward children. For a serving size of 28 grams, its sugar content is 12 grams, fat content is 0.5 grams, and sodium content is 130 milligrams. Estimate the shelf probabilities for Apple Jacks.

```
summary(mod.fit.ord)
```

```
##
## Re-fitting to get Hessian
## Call:
## polr(formula = as.factor(Shelf) ~ sugar + fat + sodium, data = cereal2,
##       method = "logistic")
##
## Coefficients:
##              Value Std. Error  t value
## sugar  -1.61101      1.2830 -1.25565
## fat    -0.05123      0.9657 -0.05305
## sodium -4.85950      1.6302 -2.98094
##
## Intercepts:
##      Value   Std. Error t value
## 1|2 -4.7534   1.4837   -3.2037
## 2|3 -3.3435   1.3810   -2.4210
## 3|4 -1.9823   1.2867   -1.5407
##
## Residual Deviance: 98.52912
## AIC: 110.5291
```

```
predict(object = mod.fit.ord, newdata = data.frame(sugar = (((12/28) - min(cereal$sugar_g))/(max(cereal$sugar_g) - min(cereal$sugar_g))),
                                                    fat = (((0.5/28) - min(cereal$fat_g))/(max(cereal$fat_g) - min(cereal$fat_g))),
                                                    sodium = (((130/28) - min(cereal$sodium_mg))/(max(cereal$sodium_mg) - min(cereal$sodium_mg))))
```

```
##              1              2              3              4
## 0.009468087 0.028204574 0.094803728 0.867523610
```

g

Construct a plot similar to Figure 3.3 where the estimated probability for a shelf is on the y-axis and the sugar content is on the x-axis. Use the mean overall fat and sodium content as the corresponding variable values in the model. Interpret the plot with respect to sugar content.

h

Estimate odds ratios and calculate corresponding confidence intervals for each explanatory variable. Relate your interpretations back to the plots constructed for this exercise.