Exploring multivariable relationships

Prof. Maria Tackett



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Carbohydrates in Starbucks food

- Starbucks often displays the total calories in their food items but not the other nutritional information.
- Our goal is to analyze the relationship between the calories and total carbohydrates (carbs) in Starbucks food items, and assess if it differs based on the type of food item (bakery, salad, sandwich, etc.)
- We can use our analysis to estimate the total carbs using information about the total calories and type for a given food time



Starbucks data



Starbucks data

■ Observations: 77 Starbucks food items



Starbucks data

- Observations: 77 Starbucks food items
- Variables:
 - carb: Total carbohydrates (in grams)
 - calories: Total calories
 - **bakery**: 1: bakery food item, 0: other food type



Terminology

- carb is the response variable
 - variable whose variation we want to understand / variable we wish to predict
 - also known as outcome or dependent variable



Terminology

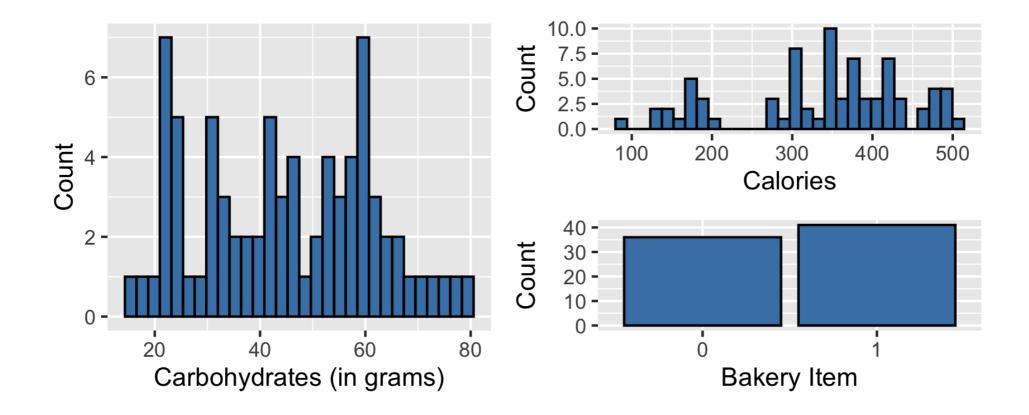
- carb is the response variable
 - variable whose variation we want to understand / variable we wish to predict
 - also known as outcome or dependent variable

- calories, bakery are the predictor variables
 - variables used to account for variation in the outcome
 - also known as explanatory, independent, or input variables



Let's look at the data

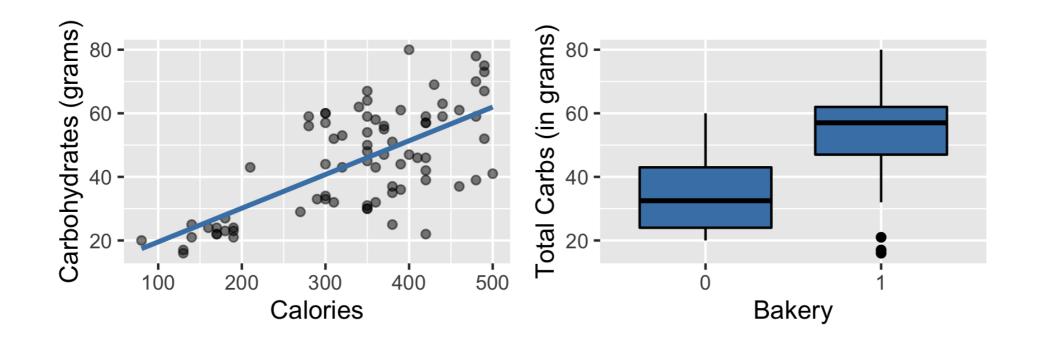
Plot Code





Response vs. Predictors

Plot Code





Model

carbs =
$$f$$
(calories, bakery) + ϵ

■ **Goal**: Determine *f*



Model

carbs =
$$f$$
(calories, bakery) + ϵ

- **Goal**: Determine *f*
- How do we determine f?
 - lacktriangle Make an assumption about the functional form f
 - Use the data to fit a model based on that form



Determine f

In general,

- 1) Choose the functional form of f, i.e. choose the appropriate model given the data
- \blacksquare Suppose f is a linear model

$$\hat{\mathbf{y}} = f(\mathbf{X}) = \beta_0 + \beta_1 x_1 + \dots + \beta_p x_p$$



Determine f

In general,

- 1) Choose the functional form of f, i.e. choose the appropriate model given the data
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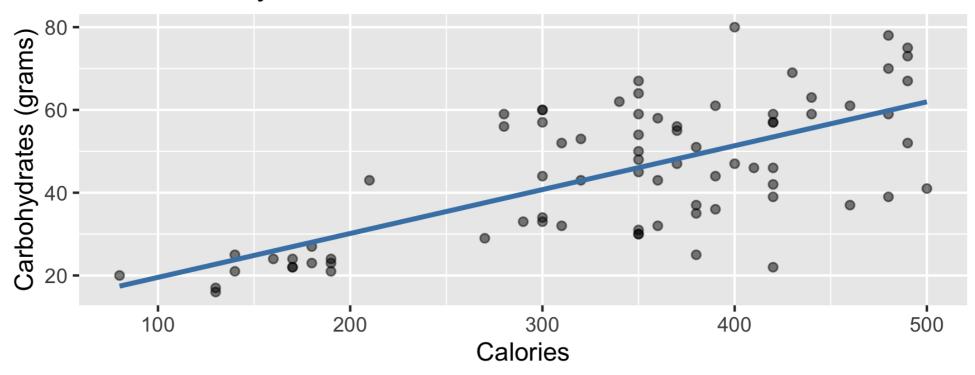
$$\hat{y} = f(\mathbf{X}) = \beta_0 + \beta_1 x_1 + \dots + \beta_p x_p$$

- 2) Use the data to fit (or train) the model, i.e **estimate the model parameters**
 - Estimate $\beta_0, \beta_1, \dots, \beta_p$



Carbs vs. Calories

Total Carbohydrates vs. Calories

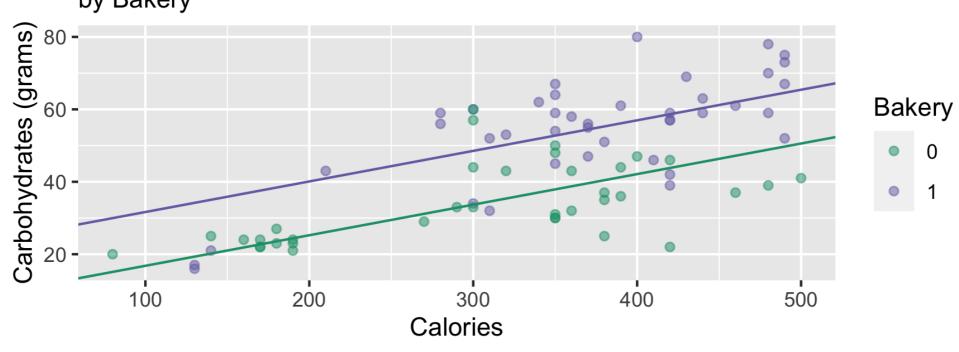




carbs =
$$\beta_0 + \beta_1$$
 calories + ϵ

Carbs vs. Calories + Bakery

Total Carbohydrates vs. Calories by Bakery

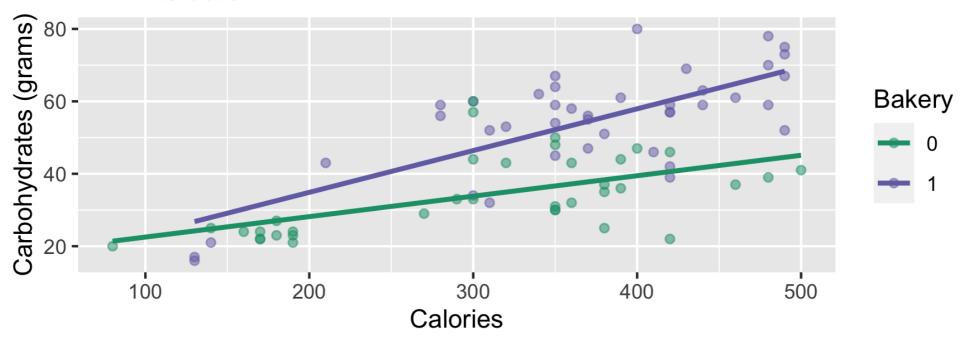




Carbs vs. Calories + Bakery (with interaction)

Total Carbohydrates vs. Calories

With Interaction





carbs = $\beta_0 + \beta_1$ calories + β_2 bakery + β_3 calories × bakery + ϵ

Code for plot on previous slide

```
ggplot(data = starbucks, aes(x = calories, y = carb, color = bakery)) +
geom_point(alpha = 0.5) +
geom_smooth(method = "lm", se = FALSE) +
labs(x = "Calories",
      y = "Carbohydrates (grams)",
      color = "Bakery",
      title = "Total Carbohydrates vs. Calories",
      subtitle = "With Interaction") +
scale_color_manual(values=c("#1B9E77", "#7570B3"))
```



Why?

carbs = $\beta_0 + \beta_1$ calories + β_2 bakery + β_3 calories × bakery + ϵ



Why?

carbs =
$$\beta_0 + \beta_1$$
 calories + β_2 bakery + β_3 calories × bakery + ϵ

Prediction:

What do we expect the total carbohydrates to be in a piece of Starbucks pumpkin bread, a bakery item that is 410 calories?



Why?

carbs =
$$\beta_0 + \beta_1$$
 calories + β_2 bakery + β_3 calories × bakery + ϵ

Prediction:

What do we expect the total carbohydrates to be in a piece of Starbucks pumpkin bread, a bakery item that is 410 calories?

Inference:

What is the relationship between the calories and total carbohydrates for bakery items at Starbucks? For non-bakery items?



Course Outline

Unit 1: Quantitative Response Variables

- Simple Linear Regression
- Multiple Linear Regression

Unit 3: Looking Ahead

- Log-linear regression
- Weighted least squares
- Communicating

- Unit 2: Categorical Response Variable
 - Logistic Regression
 - Multinomial Logistic Regression

