

Multiple logistic regression

GENERALIZED LINEAR MODELS IN R



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Instructor

Chapter overview

- Multiple logistic regression
- Formulas in R
- Model assumptions

Why multiple regression?

Problem: Multiple predictor variables. Which one should I include?

Solution: Include all of them using multiple regression.

Multiple predictor variables

- Simple linear models or simple GLM:
 - Limited to 1 Slope and 1 intercept
 - $y \sim \beta_0 + \beta_1 x + \epsilon$
- Multiple regression
 - Multiple slopes and intercepts:
 - $y \sim \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \beta_3 x_3 \dots + \epsilon$

Too much of a good thing

Theoretical maximum number of coefficients:

Maximum number of β s = Number of observations

Over-fitting:

Using too many predictors compared to number of samples

Practical maximum number of coefficients:

Number of $\beta \times 10 \approx$ Number of observations

Bus data: Two possible predictors

- With bus commuter data, 2 possible predictors
 - Number of days one commutes: `CommuteDay`
 - Distance of commute: `MilesOneWay`
- Possible to build a model with both

```
glm(Bus ~ CommuteDay + MilesOneWay, data = bus, family = 'binomial')
```

Summary of GLM with multiple predictors

Call:

```
glm(formula = Bus ~ CommuteDays + MilesOneWay, family = "binomial",  
     data = bus)
```

Deviance Residuals:

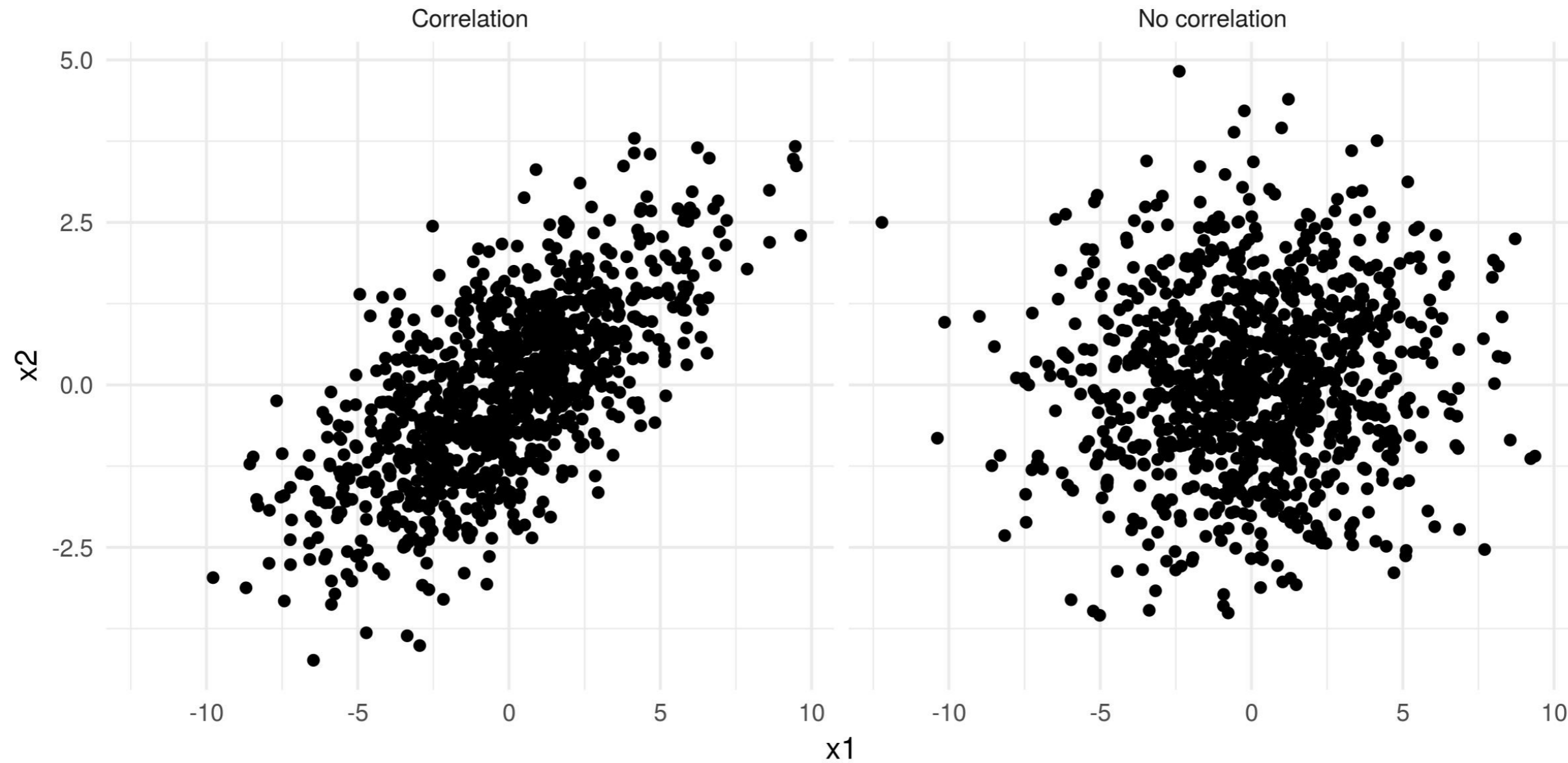
Min	1Q	Median	3Q	Max
-1.0732	-0.9035	-0.7816	1.3968	2.5066

Coefficients:

	Estimate	Std. Error	z value	Pr(> z)	
(Intercept)	-0.707515	0.119719	-5.910	3.42e-09	***
CommuteDays	0.066084	0.023181	2.851	0.00436	**
MilesOneWay	-0.059571	0.003218	-18.512	< 2e-16	***

#...

Correlation between predictors



Order of coefficients

No correlation between predictors

- Order not important
- $y \sim x_1 + x_2 + \epsilon \approx y \sim x_2 + x_1 + \epsilon$

Correlation between predictors

- Order may changes estimates
- $y \sim x_1 + x_2 + \epsilon \neq y \sim x_2 + x_1 + \epsilon$

Let's practice!

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Formulas in R

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Why care about formulas for multiple logistic regression?

- Formulas backbone of regression
- Tricky to figure out
- Understanding `model.matrix()` key

Slopes

- Estimates coefficient for continuous variable
 - e.g., `height = c(72.3, 21.1, 3.7, 1.0)`
- Formula also requires a global intercept
- Multiple slopes: Slope for each predictor

Intercepts

- Discrete groups used to predict
- factor or character in R: `fish = c("red", "blue")`
- Single intercept has two options:
 - Reference intercept + contrast: `y ~ x`
 - Intercept for each group: `y ~ x -1`

Multiple intercepts

- Estimates effect of each group compared to reference group
- The first group, alphabetically, in the factor
- Default has one reference group per variable
 - `y ~ x1 + x2`
- Can specify one group to estimate an intercept for all groups
 - `y ~ x1+ x2 - 1`
- First variable has intercept estimated for each group

Dummy variables

- Codes group membership
- Used under the hood (i.e., `model.matrix()`)
- 0s and 1s for each group
- Example input: `color = c("red", "blue")`
- Dummy variables for `y ~ colors` :
 - `intercept = c(1, 1)`
 - `blue = c(0, 1)`
- Dummy variables for `y ~ colors - 1` :
 - `red = c(1, 0)`
 - `blue = c(0, 1)`

model.matrix()

- `model.matrix()` does legwork for us
- Foundation for formulas in R

```
model.matrix( ~ colors)
```

```
(Intercept) colorsred
1           1           1
2           1           0
```

```
attr(,"assign")
```

```
[1] 0 1
```

```
attr(,"contrasts")
attr(,"contrasts")$colors
```

```
"contr.treatment"
```

- Order determined by factor order
- Change order change with Tidyverse or `factor()`

Factor vs numeric caveat

- R thinks variable is numeric
 - e.g., `month = c(1, 2, 3)`

```
month <- c(1, 2, 3)
model.matrix(~ month)
```

```
(Intercept) month
1           1     1
2           1     2
3           1     3
```

```
attr(,"assign")
```

```
0 1
```

- Need to specify factor or character
 - e.g., `month = factor(c(1, 2, 3))`

```
model.matrix(~ month)
```

```
(Intercept) month2 month3
1           1     0     0
2           1     1     0
3           1     0     1
```

```
attr(,"assign")
```

```
0 1 1
```

```
attr(,"contrasts")$month
```

```
"contr.treatment"
```

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Assumptions of multiple logistic regression

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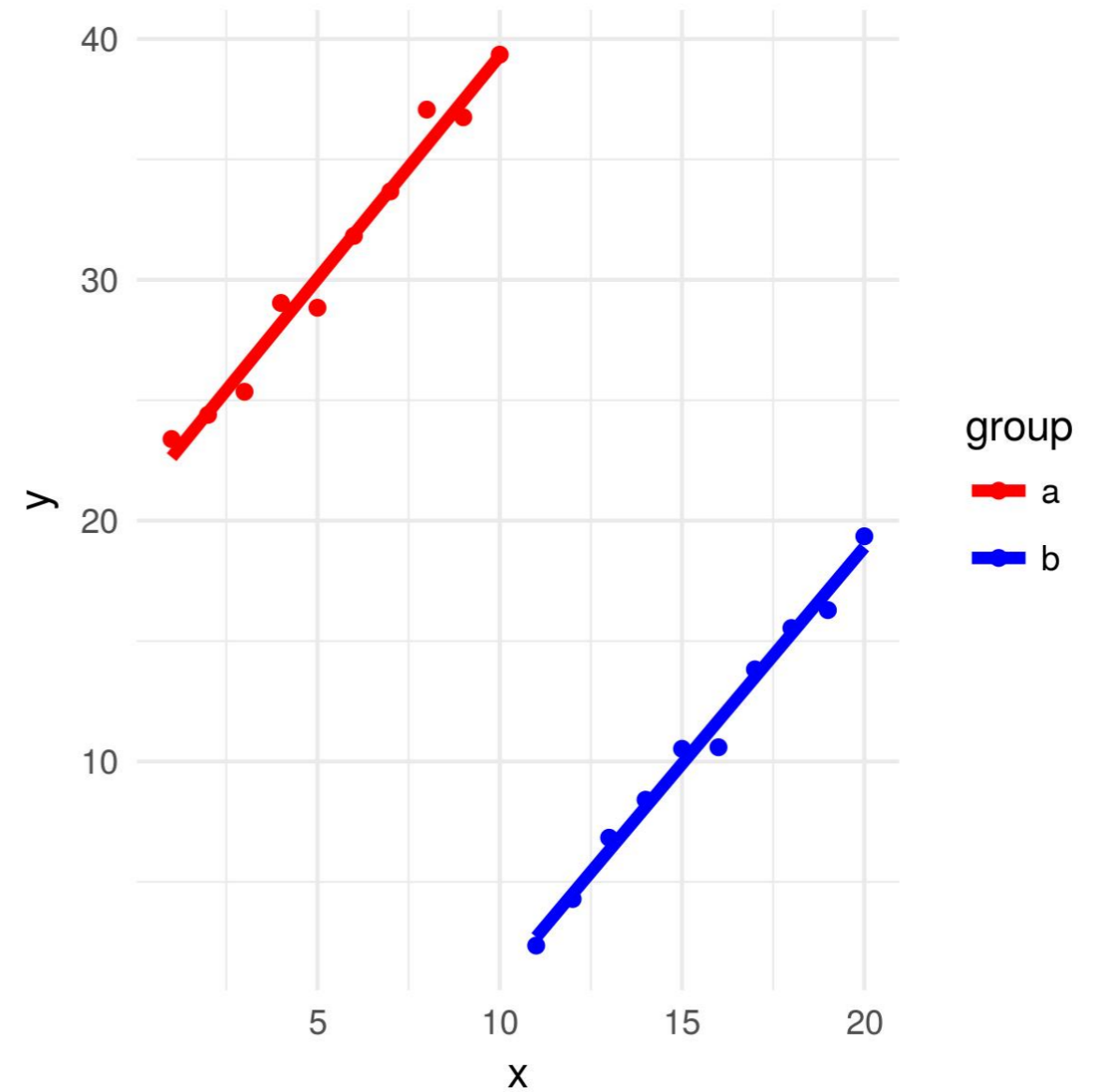
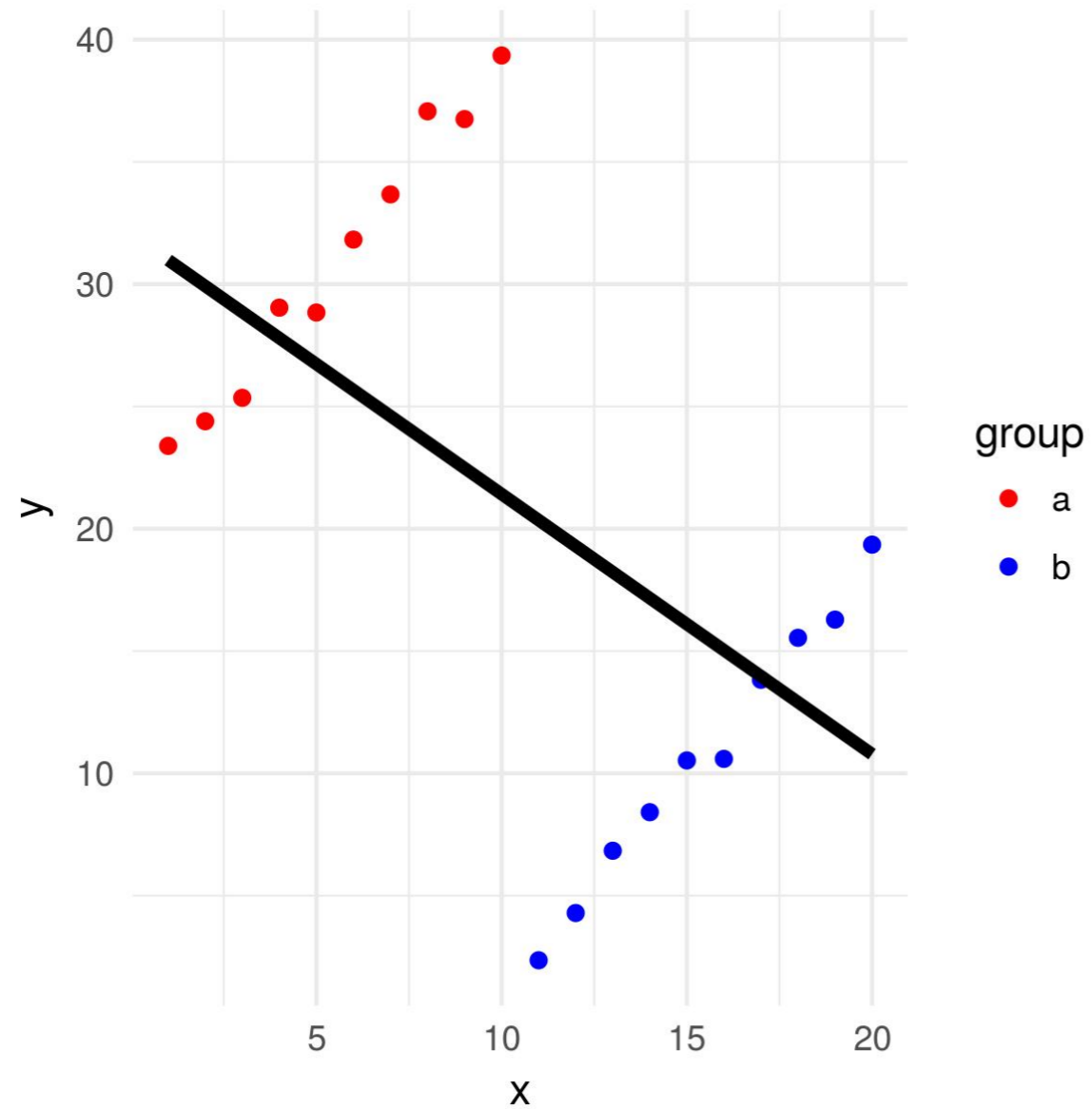


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Assumptions

- Limitations also apply to Poisson and other GLMs
- Important assumptions:
 - Simpson's paradox
 - Linear, monotonic
 - Independence
 - Overdispersion

Example Simpson's paradox



Simpson's paradox

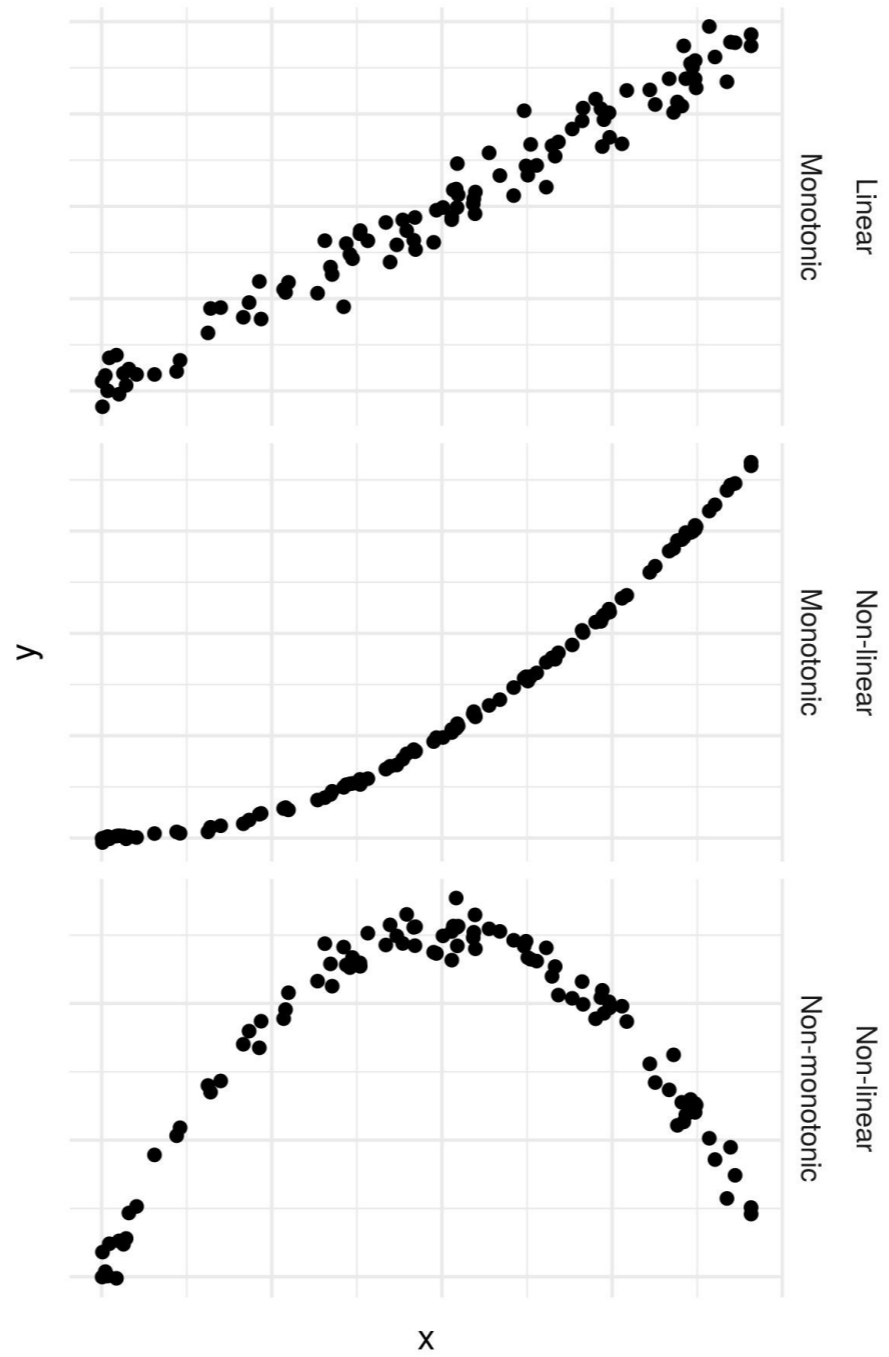
Key points

- Missing important predictor
- Inclusion changes outcome
- Easy to visualize with `lm()`

Simpson's paradox and admission data

Admissions data

- University of California Berkeley
- Graduate admission
- Rate of admission by department and gender
- Does bias exist?



Independence

Predictors

- If all independent, order has no effect on estimates
- If non-independent, order can change estimates

Response

- What is unit of focus?
- Individual, groups, group of groups?
- Test scores
 - Individual student?
 - Teacher? School? District?

Overdispersion

- Too many zeros or one (Binomial)
- Too many zeros, too large variance (Poisson)
- Variance changes
- Beyond scope of this course

Let's practice!

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Conclusion

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What you've learned

- How GLM extends LM:
 - Poisson Error term
 - Binomial Error term
- Understanding and plotting results
- GLM with multiple regression

Where to from here?

- DataCamp [Multiple \(linear\) regression course in R](#) (if you missed it)
- Extending to include random effects with [Hierarchical and mixed-effect models in R](#)
- Fit [generalized additive models in R](#) (GAMs) to non-linear models
- Decide what coefficients to use with model selection such as AIC
- Many other types of regression
- Searching and R packages documentation to learn more

Happy coding!

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