Using likelihoods to compare models

Prof. Maria Tackett

01.24.22

Click for PDF of slides



Announcements

- Week 03 reading:
 - BMLR: Chapter 3 Distribution Theory (for reference)
 - BMLR: Chapter 4 Poisson Regression
- Quiz 01 Tue, Jan 25 at 9am Thu, Jan 26 at 3:30pm (start of lab)



Quiz 01

- Open Jan 25 at 9am and must be completed by Thu, Jan 26 at 3:30pm
 - The quiz is not timed and will be administered in Gradescope.
- Covers
 - Syllabus
 - BMLR Chapters 1 2
 - Jan 05 Jan 24 lectures
- Fill in the blank, multiple choice, short answer questions
- Open book, open note, open internet (not crowd sourcing sites). You <u>cannot</u> discuss the quiz with anyone else. Please email me if you have questions.



Learning goals

- Use likelihood to compare models
- Activity: Exploring the response variable for mini-project 01



Recap



Fouls in college basketball games

The data set <u>04-refs.csv</u> includes 30 randomly selected NCAA men's basketball games played in the 2009 - 2010 season.

We will focus on the variables **foul1**, **foul2**, and **foul3**, which indicate which team had a foul called them for the 1st, 2nd, and 3rd fouls, respectively.

- **H**: Foul was called on the home team
- V: Foul was called on the visiting team

We are focusing on the first three fouls for this analysis, but this could easily be extended to include all fouls in a game.



Fouls in college basketball games

```
refs <- read_csv("data/04-refs.csv")
refs %>% slice(1:5) %>% kable()
```

game	date	visitor	hometeam	foul1	foul2	foul3
166	20100126	CLEM	ВС	V	V	V
224	20100224	DEPAUL	CIN	Н	Н	V
317	20100109	MARQET	NOVA	Н	Н	Н
214	20100228	MARQET	SETON	V	V	Н
278	20100128	SETON	SFL	Н	V	V

We will treat the games as independent in this analysis.



Likelihoods

A likelihood is a function that tells us how likely we are to observe our data for a given parameter value (or values).

Model 1 (Unconditional Model)

• p_H : probability of a foul being called on the home team

Model 2 (Conditional Model)

- $p_{H|N}$: Probability referees call foul on home team given there are equal numbers of fouls on the home and visiting teams
- $p_{H|HBias}$: Probability referees call foul on home team given there are more prior fouls on the home team
- $p_{H|VBias}$: Probability referees call foul on home team given there are more prior fouls on the visiting team



Likelihoods

A likelihood is a function that tells us how likely we are to observe our data for a given parameter value (or values).

Model 1 (Unconditional Model)

$$Lik(p_H) = p_H^{46} (1 - p_H)^{44}$$

Model 2 (Conditional Model)

$$Lik(p_{H|N}, p_{H|HBias}, p_{H|VBias}) = [(p_{H|N})^{25} (1 - p_{H|N})^{23} (p_{H|HBias})^{8}$$
$$(1 - p_{H|HBias})^{12} (p_{H|VBias})^{13} (1 - p_{H|VBias})^{9}]$$



Maximum likelihood estimates

The maximum likelihood estimate (MLE) is the value between 0 and 1 where we are most likely to see the observed data.

Model 1 (Unconditional Model)

•
$$\hat{p}_H = 46/90 = 0.511$$

Model 2 (Conditional Model)

•
$$\hat{p}_{H|N} = 25/48 = 0.521$$

•
$$\hat{p}_{H|HBias} = 8/20 = 0.4$$

•
$$\hat{p}_{H|VBias} = 13/22 = 0.591$$

- What is the probability the referees call a foul on the home team, assuming foul calls within a game are independent?
- Is there a tendency for the referees to call more fouls on the visiting team or home team?
- Is there a tendency for referees to call a foul on the team that already has more fouls?



MLEs for Model 2

<u>Click here</u> for details on finding MLEs for Model2



Model comparison



Model comparisons

- Nested models
- Non-nested models



Comparing nested models



Nested Models

Nested models: Models such that the parameters of the reduced model are a subset of the parameters for a larger model

Example:

Model A:
$$y = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \epsilon$$

Model B: $y = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \beta_3 x_3 + \beta_4 x_4 + \epsilon$

Model A is nested in Model B. We could use likelihoods to test whether it is useful to add x_3 and x_4 to the model.

$$H_0: \beta_3 = \beta_4 = 0$$

 H_a : at least one β_i is not equal to 0



Nested models

Another way to think about nested models: Parameters in larger model can be equated to get the simpler model or if some parameters can be set to constants

Example:

Model 1: p_H

Model 2: $p_{H|N}$, $p_{H|HBias}$, $p_{H|VBias}$

Model 1 is nested in Model 2. The parameters $p_{H|N}$, $p_{H|HBias}$, and $p_{H|VBias}$ can be set equal to p_H to get Model 1.

 $H_0: p_{H|N} = p_{H|HBias} = p_{H|VBias} = p_H$

 H_a : At least one of $p_{H|N}$, $p_{H|HBias}$, $p_{H|VBias}$ differs from the others



Steps to compare models

- 1 Find the MLEs for each model.
- 2 Plug the MLEs into the log-likelihood function for each model to get the maximum value of the log-likelihood for each model.
- 3 Find the difference in the maximum log-likelihoods
- 4 Use the Likelihood Ratio Test to determine if the difference is statistically significant



Steps 1 - 2

Find the MLEs for each model and plug them into the log-likelihood functions.

Model 1:

• $\hat{p}_H = 46/90 = 0.511$

```
loglik1 <- function(ph){
  log(ph^46 * (1 - ph)^44)
}
loglik1(46/90)</pre>
```

[1] -62.36102

Model 2

- $\hat{p}_{H|N} = 25/48 = 0.521$
- $\hat{p}_{H|HBias} = 8/20 = 0.4$
- $\hat{p}_{H|VBias} = 13/22 = 0.591$

```
## [1] -61.57319
```



Step 3

Find the difference in the log-likelihoods

```
(diff <- loglik2(25/48, 8/20, 13/22) - loglik1(46/90))
## [1] 0.7878318
```

Is the difference in the maximum log-likelihoods statistically significant?



Likelihood Ratio Test

Test statistic

 $LRT = 2[\max\{\log(Lik(\text{larger model}))\} - \max\{\log(Lik(\text{reduced model}))\}]$

$$= 2 \log \left(\frac{\max\{(Lik(\text{larger model})\}}{\max\{(Lik(\text{reduced model})\}} \right)$$

LRT follows a χ^2 distribution where the degrees of freedom equal the difference in the number of parameters between the two models



Step 4

```
(LRT <- 2 * (loglik2(25/48, 8/20, 13/22) - loglik1(46/90)))
## [1] 1.575664
```

The test statistic follows a χ^2 distribution with 2 degrees of freedom. Therefore, the p-value is $P(\chi^2 > LRT)$.

```
pchisq(LRT, 2, lower.tail = FALSE)
```

```
## [1] 0.4548299
```

The p-value is very large, so we fail to reject H_0 . We do not have convincing evidence that the conditional model is an improvement over the unconditional model. Therefore, we can stick with the unconditional model.



Comparing non-nested models



Comparing non-nested models

```
AIC = -2(max log-likelihood) + 2p

BIC = -2(max log-likelihood) + plog(n)

(Model1_AIC <- 2 * loglik1(46/90) + 2

## [1] -122.722

## [1] -121.3208

(Model2_AIC <-2 * loglik2(25/48, 8/20,

## [1] -117.1464

## [1] -112.9428
```

Choose Model 1, the unconditional model, based on AIC and BIC



Looking ahead

- Likelihoods help us answer the question of how likely we are to observe the data given different parameters
- In this example, we did not consider covariates, so in practice the parameters we want to estimate will look more similar to this

$$p_{H} = \frac{e^{\beta_{0} + \beta_{1}x_{1} + \dots + \beta_{p}x_{p}}}{1 + e^{\beta_{0} + \beta_{1}x_{1} + \dots + \beta_{p}x_{p}}}$$

- Finding the MLE becomes much more complex and numerical methods may be required.
 - We will primarily rely on software to find the MLE, but the conceptual ideas will be the same



Response variable in mini-project 01



Activity Instructions

The goal of this activity is for your team to start exploring the response variable for your mini-project 01 analysis. The properties explored in this activity are ones you will consider throughout the semester as you decide which GLM is most appropriate for a given data set. Use <u>Table 3.1 in BMLR</u> for reference.

Write the following for the primary response variable in your analysis:

- What is the response variable? What is its definition?
- Is the response variable discrete or continuous?
- What possible values can it take? (not necessarily just the values in the data set)
- What is the name of the distribution the variable follows?
- What is/are the parameter(s) for this distribution? Estimate the parameters from the data.
- Visualize the distribution of the response variable. Is this what you expected? Why or why not?



Activity Instructions

<u>Click here</u> to put the answers on your team's slide.

You can add any analysis to the bottom of the **proposal.Rmd** document in your team's project repo.



Looking ahead

- Review <u>Chapter 3 Distribution Theory</u>
 - Use this chapter as a reference throughout the semester
- For next time <u>Chapter 4 Poisson Regression</u>



Acknowledgements

These slides are based on content in <u>BMLR Chapter 2 - Beyond Least Squares: Using Likelihoods</u>

