DATA MINING 2 Odds and Log Odds

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Contains edited slides from StatQuest



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Visually, we have 5 games total...

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...and 4 of which my team will lose.

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Alternatively, we can write this as a fraction... $\frac{1}{4} = 0.25$

...if we do the math, we see that the odds are 0.25 that my team will win the game.

Here's another example: You might say that the odds in favor of my team winning the game are 5 to 3:

Alternatively, we can write this as a fraction... $\frac{5}{3} = 1.7$ see that 1.7 that

...if we do the math, we see that the odds are 1.7 that my team will win the game.

Note: Odds are not probabilities!!!

Odds vs Probability

The odds are the ratio of something happening (i.e. my team winning)...

...to something not happening (i.e. my team **not winning**).



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Probability is the ratio of something happening (i.e. my team **winning**)...

...to *everything* that could happen (i.e. my team **winning** *and* **losing**).



Odds vs Probability

In the previous example, the odds in favor of my team **winning** the game are 5 to 3...

5

3

5 8

...here's the math...

...however, the probability of my team winning is the number of games they win (5) divided by the total number of games they play (8)...

Odds from Probabilities

 $\frac{5}{3} = 1.7$ In the last example we saw that the odds of winning are 1.7...

Odds from Probabilities $= \frac{5}{3} = 1.7$ $= \frac{5}{8} = 0.625$...and the probability of winning is 0.625.

Odds from Probabilities $=\frac{5}{3}=1.7$ $=\frac{5}{8}=0.625$ $=\frac{3}{8} = 0.375$...the probability of losing is 0.375

Odds from Probabilities $=\frac{5}{3}=1.7$ $=\frac{5}{8}=0.625$ $\frac{3}{8} = 0.375$

NOTE: We could also calculate the probability of **losing** as:

1 - the probability of winning



Odds from Probabilities



Now let's take the ratio of the probability of winning to the probability of losing...

The ratio of the probability of winning...

...to the probability of losing





odds = p/(1-p)

We can see that the worse my team is, the odds of winning get closer and closer to 0.





In other words, if the odds are *against* my team **winning**, then they will be between 0 and 1.



We can see that the better my team is, the odds of winning start at 1 and just go up and up.





In other words, if the odds are *for* my team **winning**, then they will be between 1 and infinity!



Another way to look at this is with a number line...



The odds of my team **losing** go from 0 to 1...





The asymmetry makes it difficult to compare the odds for or against my team winning.



Log of the Odds For example if the odds are ... but if the odds are in against 1 to 6, then the favor 6 to 1, then the odds odds are 1/6 = 0.17... are 6/1 = 6!....to infinity and beyond! 2 3 5 6 0 4

Taking the log() of the odds (i.e. log(odds)) solves this problem by making everything symmetrical.







Using the log function, the distance from the origin (or 0) is the same for 1 to 6 and 6 to 1 odds.





Earlier we saw that odds can be calculated from counts...



Odds and Log Odds $= \frac{5}{3} = 1.$

$$\log(\text{odds}) = \log(\frac{5}{3}) = \log(\frac{p}{(1-p)}) = \log(1.7) = 0.53$$

...and that means we can calculate the log of the odds with counts or probabilities - either way, we'll get the same value.



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Odds and Log Odds $= \frac{5}{3} = 1.7$

⁷ log(odds) = log(
$$\frac{5}{3}$$
) = log($\frac{p}{(1-p)}$) = log(1.7) =0.53

NOTE: The log of the ratio of the probabilities is called the **logit** function and forms the basis for logistic regression.



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- Odds are the ratio of something happening to something not happening
- Log odds are the log of the odds
- What's the big deal?

To show you what the big deal is all about, if I pick pairs of random numbers that add up to 100 (for example) and use them to calculate the log(odds) and draw a histogram...



...the histogram is in the shape of a normal distribution!



This makes the log(odds) useful for solving certain statistics problems - specifically ones where we are trying to determine probabilities about win/lose, or yes/no, or true/false types of situations.





When people say "odds ratio", they are talking about a "**ratio of odds**".





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Odds Ratios







Log of Odds Ratios













We can use an "odds ratio" to determine if there is a relationship between the mutated gene and cancer.

If someone has the mutated gene, are the odds higher that they will get cancer?















... and the odds ratio tells us that



...and the log(odds ratio) is 1.93.

What does all this mean?





...larger values mean that the mutated gene is a good predictor of cancer. Smaller values mean that the mutated gene is not a good predictor of cancer.

Odds Again

 Given some event with probability p of being 1, the odds of that event are given by:

odds = p / (1-p)

- When we go from Normal to High, the odds of being Sick triple:
- Odds ratio: 0.293/0.111 = 2.64
- 2.64 times more likely to be Sick with high values

| | Sick | | | |
|-------|--------|-----|------|-------|
| | | Yes | No | Total |
| Value | Normal | 402 | 3614 | 4016 |
| | High | 101 | 345 | 446 |
| | Total | 503 | 3959 | 4462 |

Sick

The odds of being sick if you have a Normal value are:

- Odds(Sick|Normal) = P(sick)/1-P(sick) =
 - = (402/4016) / (1 (402/4016))
 - = 0.1001 / 0.8889 = 0.111

The odds of being not sick with a Normal value is the reciprocal:

Odds(not Sick|Normal) = 0.8999/0.1001 = 8.99

For the High value we have

- Odds(Sick|High) = 101/345 = 0.293
- Odds(not Sick|High) = 345/101 = 3.416

Logit Transform

- The logit is the natural log of the odds
- logit(p) = ln(odds) = ln(p/(1-p))



References

• Regression. Appendix D. Introduction to Data Mining.

