

Using Bayesian Statistics to Model Binary Outcomes



Abstract

Bayesian statistics

- Method for reallocating probability given data
- Improves the original assumptions

Our project

- Brief overview on Bayesian statistics
- We use simple examples to showcase its importance
- We show that models can be improved based on data that we gather.

Models used

- Determining the bias of a coin
- Determining the likelihood of cystic fibrosis given a positive test result.

Flipping Coins

- Assume we have 100 coins
- 99 are fair, 1 is double sided
- We flip the coin once and get heads
- What are the chances of having a fair coin?
- Intuitively: 99%
- What if we use Bayes' theorem?

Applying Bayes' theorem to Coins

$T =$ We have a fair coin

$d =$ We flipped heads

- Initially : $p(T) = 99\%$
- $p(d|T) = 0.5$
- $p(d) = 0.99 * 0.5 + 0.01 * 1 = .505$

Apply Bayes' theorem:

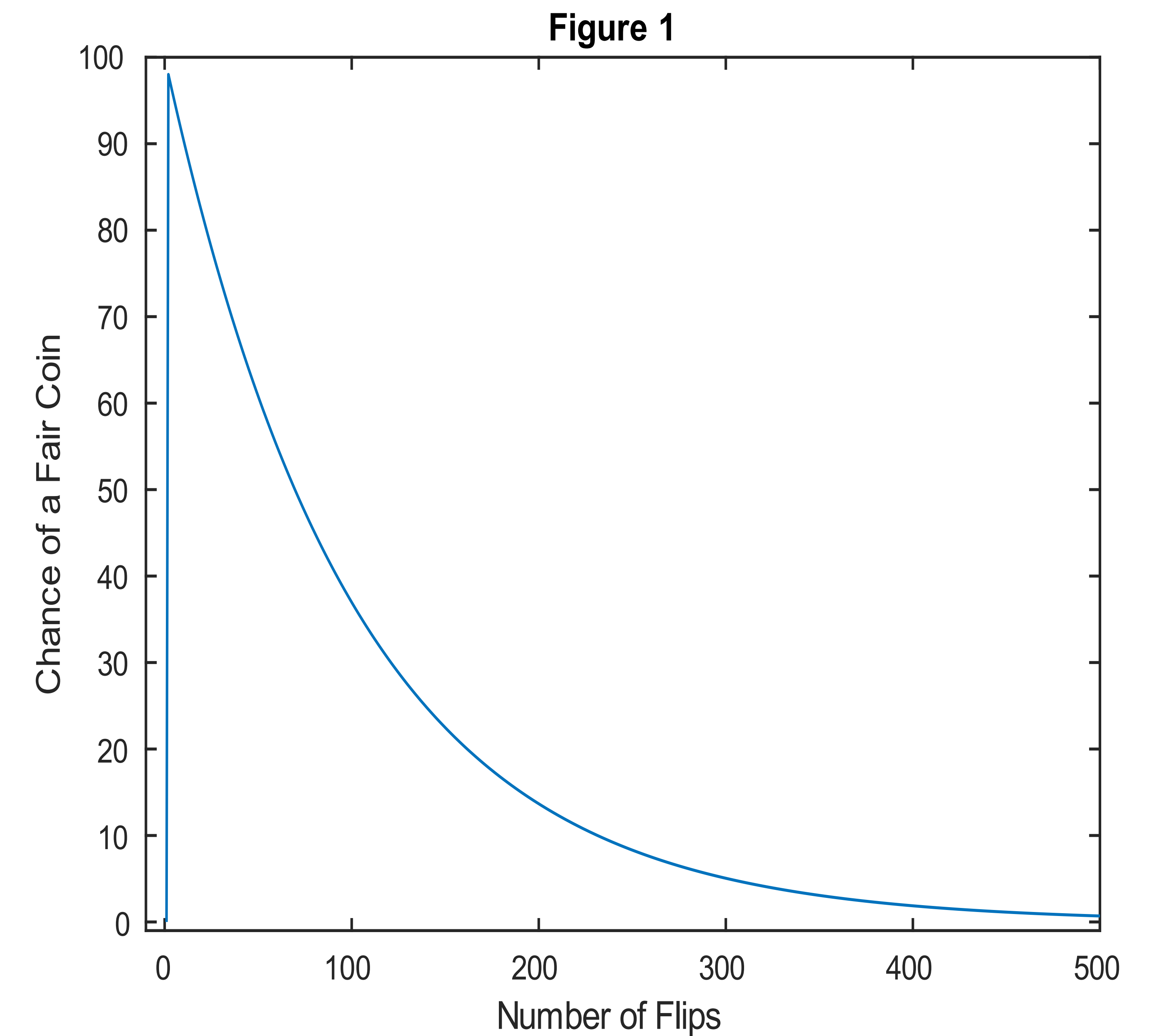
$$p(T|d) = \frac{0.5 * .99}{.505} = 0.98 = 98\%$$

- Does not match intuitive result.
- Flipping the coin again, we get heads. What is the new probability?
- Our new $p(T) = 0.98$

$$p(T|d) = \frac{0.5 * .98}{.505} = 0.97 = 97\%$$

The more heads that are obtained when flipping, the more likely it is that the coin is biased.

J. Bayes.



Application to Cystic Fibrosis Screenings

$T =$ The patient has cystic fibrosis

$d =$ A positive test result

(Assay of trypsinogen concentrations)

- Initially $p(T) = 1$ in 4000 = 0.025%
 - $p(d|T) = 0.99$
- $p(d) = 2.5^{-4} * 0.99 + 0.99975 * 0.01 = 0.10245$

Apply Bayes' Theorem

$$p(T|d) = \frac{0.99 * 2.5 * 10^{-4}}{0.10245} = 0.0024 = 0.24 \%$$

This illustrates why medical test with even the slightest rate of false positives are performed multiple times or in conjunction with another test

It is also important to remember that a patient exhibiting any symptoms of the disease would dramatically increase our $p(T)$ value

Bayes' Theorem

Likelihood

How likely is the data given that the theory is true?

Prior

How likely was the theory before observing the data?

$$p(T|d) = \frac{p(d|T) * p(T)}{p(d)}$$

Posterior

How likely is our theory given the data?

Marginal

How likely is the data to occur under all possible theories? How many times does the data occur?

References

- Hammond, K. B., Abman, S. H., Sokol, R. J., & Accurso, F. J. (1991). Efficacy of Statewide Neonatal Screening for Cystic Fibrosis by Assay of Trypsinogen Concentrations. *New England Journal of Medicine*, 325(11), 769-774. doi:10.1056/nejm1991091232511045
- Kruschke, J. K. (2015). *Doing Bayesian data analysis: A tutorial with R, JAGS, and Stan*. Amsterdam: Elsevier/AP.

Acknowledgments

We would like to thank Dr. Hurtado Rúa for her guidance and the Chose Ohio First program for giving us the opportunity to pursue this project.