

# Bayes' Theorem



Rev. Thomas Bayes  
1702-1761

# Conditional Probability

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$$\Pr(A \mid B) = \Pr(A, B) / \Pr(B)$$

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## **BAYES RULE**

# False Positives

- If a patient has a disease the test returns a positive 99% of the time
- If a patient does not have the disease, the test returns positive 5% of the time
- 0.1% of the population has the disease
- What is the probability that someone who tested positive has the disease?

Suppose  $A$  = has disease

$B$  = tested positive

$$P(A|B) = \frac{P(B|A)P(A)}{P(B|A)P(A) + P(B|\text{not } A)P(\text{not } A)}$$

$$P(A|B) = \frac{0.99 \cdot 0.001}{0.99 \cdot 0.001 + 0.05 \cdot 0.999}$$

$$P(A|B) = \frac{0.00099}{0.00099 + 0.04995} \approx 0.019$$

DID THE SUN JUST EXPLODE?  
(IT'S NIGHT, SO WE'RE NOT SURE.)

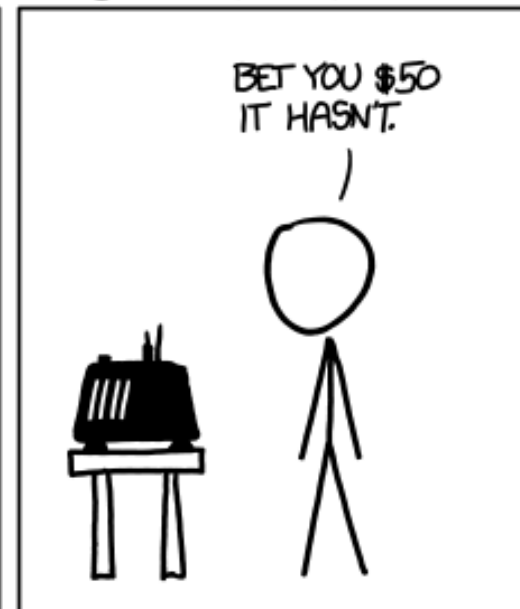


What is the  
probability  
that the sun  
exploded??

FREQUENTIST STATISTICIAN:



BAYESIAN STATISTICIAN:





# Bayes' Theorem

Posterior

Likelihood Prior

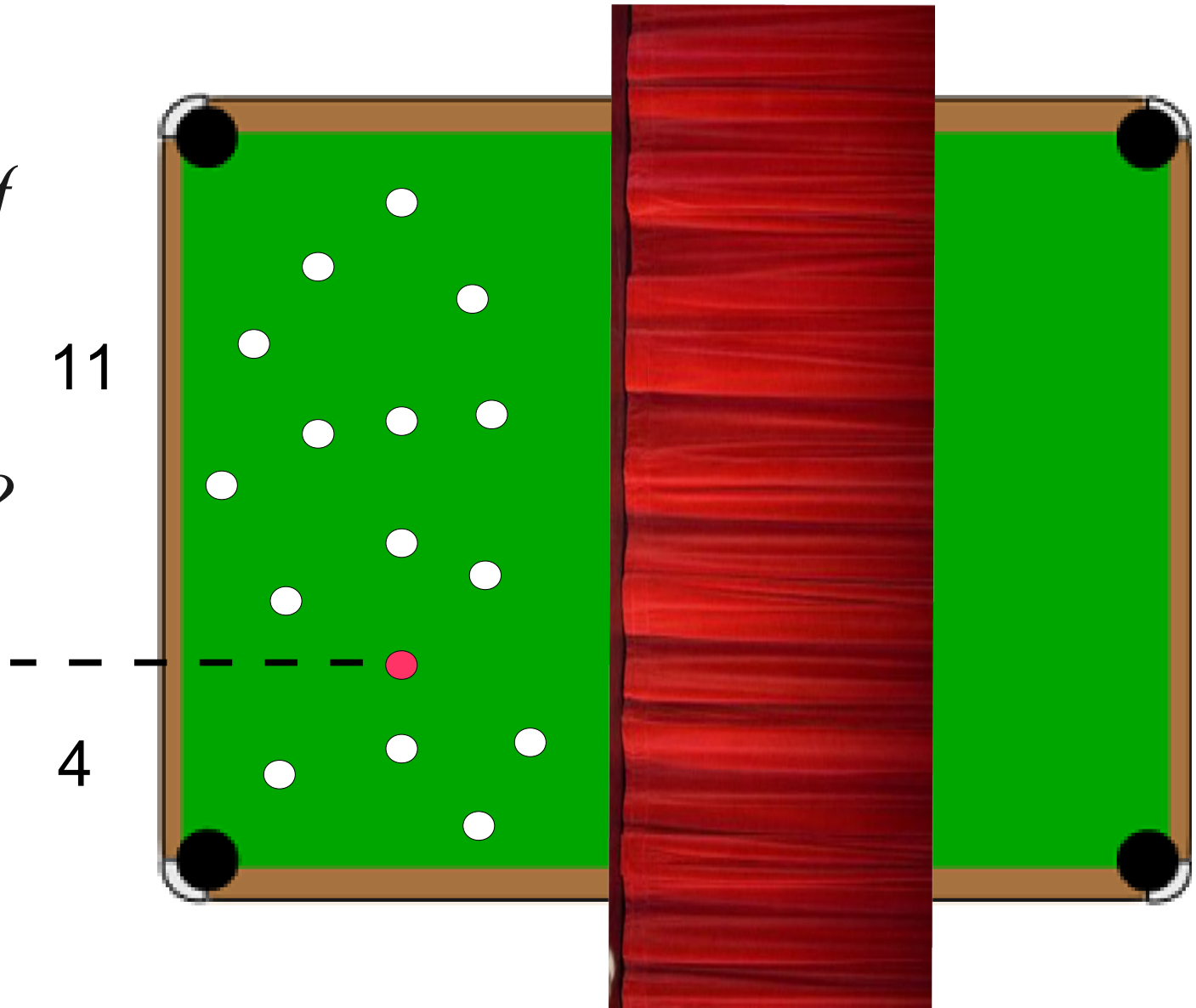
$$P(\theta|y) = \frac{P(y|\theta)P(\theta)}{P(y)}$$
$$= \frac{P(y|\theta)P(\theta)}{\int_{-\infty}^{\infty} P(y|\theta)P(\theta)d\theta}$$

# Bayes' Billiard Table

$$P(\theta) = \text{Unif}$$

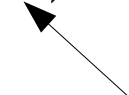


$$P(\theta|y) = ??$$



$$P(\theta|y) \propto P(y|\theta) P(\theta)$$

Unif(0,1)



What is  $P(y | \theta)$ ?

$$L = P(y|\theta) = \text{Binom}(y|n, \theta)$$

$$P(\theta|y) = \frac{\text{Binom}(y|n, \theta) \text{Unif}(\theta|0,1)}{\int_0^1 \text{Binom}(y|n, \theta) \text{Unif}(\theta|0,1)}$$

$$P(\theta|y) = \frac{\binom{n}{y} \theta^y (1-\theta)^{n-y} \cdot 1}{\int_0^1 \binom{n}{y} \theta^y (1-\theta)^{n-y} \cdot 1}$$

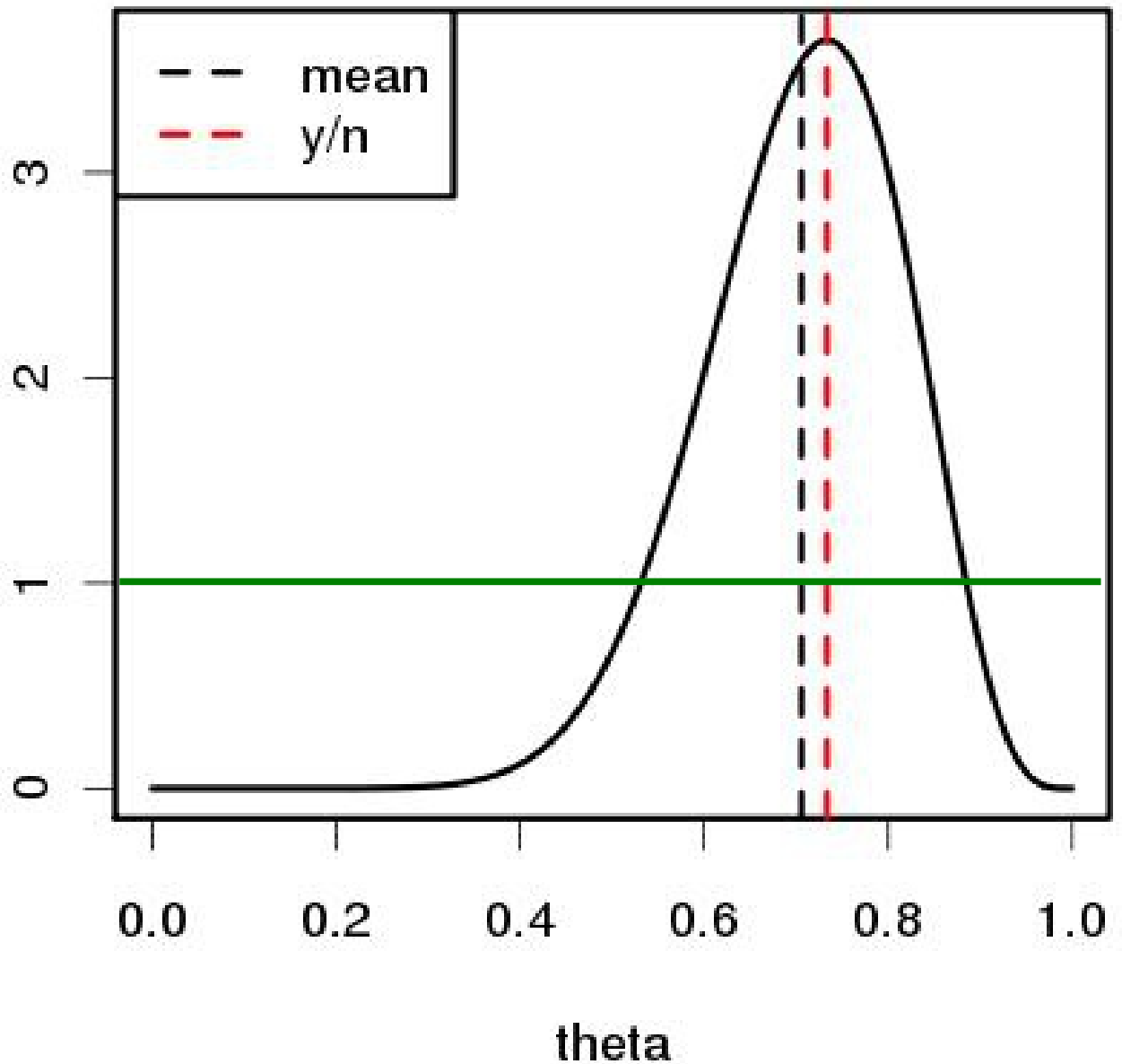
$$P(\theta|y) = \frac{\theta^y (1-\theta)^{n-y}}{\int_0^1 \theta^y (1-\theta)^{n-y}}$$

Mean? Variance? Confidence Interval?

$$Beta(x|\alpha, \beta) = \frac{x^{\alpha-1} (1-x)^{\beta-1}}{\int_0^1 x^{\alpha-1} (1-x)^{\beta-1}}$$

$$P(\theta|y) = Beta(x|y+1, n-y+1)$$

- Posterior is a PDF
- $\theta$  is a random variable
- Interested in full distribution

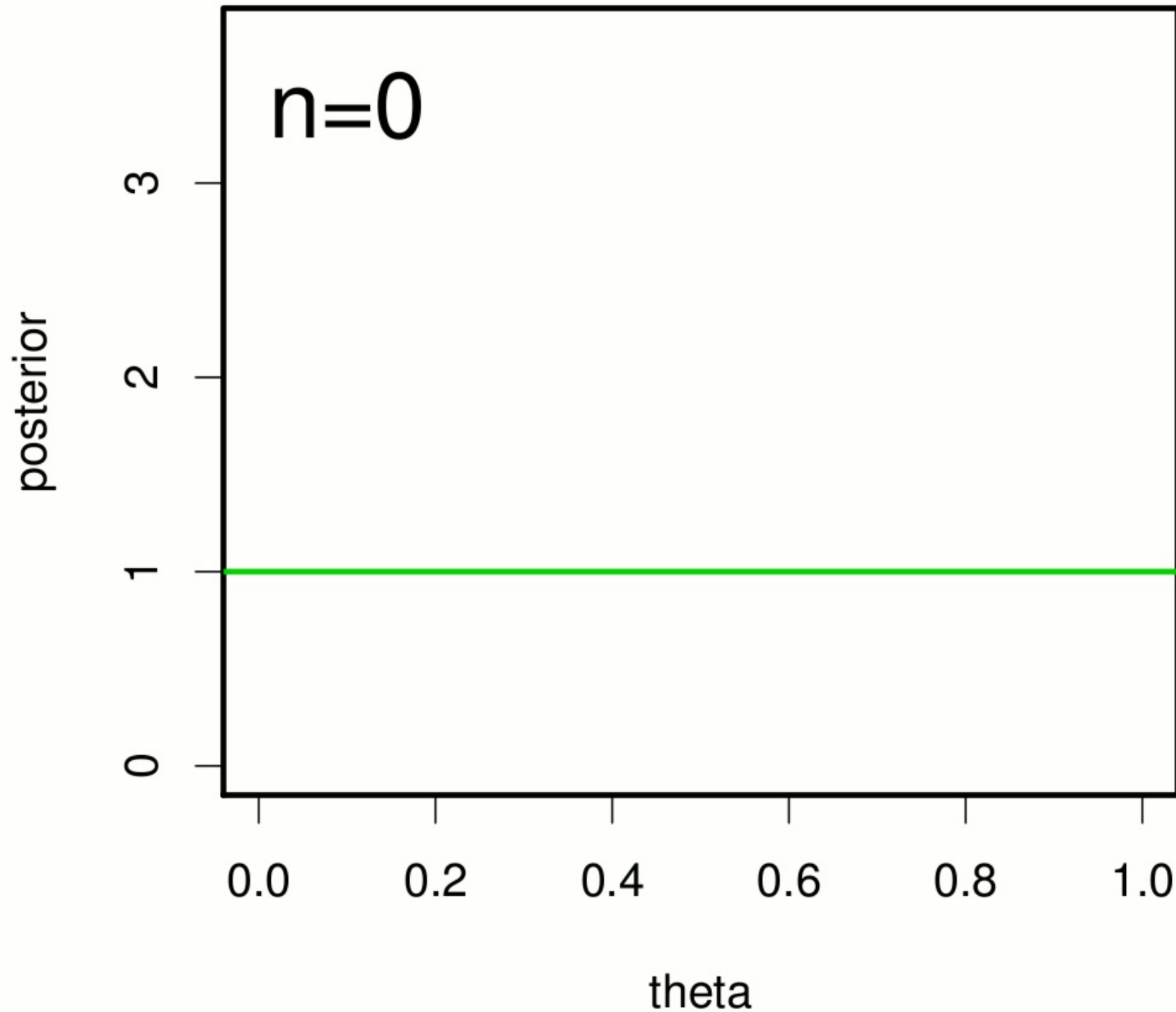


# Priors

- Makes it possible to calculate a posterior density of the model parameter rather than the likelihood of the data
- Provides a way of incorporating information that is external to the data set(s) at hand
- Inherently sequential

Previous Posterior = New Prior

# Data updates the prior





# Where do Priors come from?

- Uninformative / vague
  - Chosen to have minimal information content, allows the likelihood to dominate the analysis
- Previous analyses
  - Must be equivalent
  - Variance inflation
- “The literature”
  - Meta-analysis
- Expert knowledge

# Where do Priors come from?

- Uninformative / vague

- Ch
  - the

**Prior specification must be “blind” to the data in the analysis!!**

- Prev

- Mu

- Va

- “The

- Me

**No “double dipping” -- leads to falsely overconfident results**

- Expert knowledge

# How do I choose a prior PDF?

- Analogous to how we choose the data model
  - Range restrictions, shape, etc.
- Conjugacy
  - A prior is conjugate to the likelihood if the posterior PDF is in the same family as the prior
  - Allow for closed-form analytical solutions to either full posterior or (in multiparameter models) for the conditional distribution of that parameter.
  - Modern computational methods no longer require conjugacy

# Example: Tree mortality rate

- Data: observed  $n=4$  trees,  $y=1$  died this year

$$L = P(y|\theta) = \text{Binom}(y|\theta, n)$$

- Prior: last year observed  $n_0=2$  trees,  $y_0=1$  died

$$\text{Prior} = P(\theta) = \text{Beta}(\theta|y_0, n_0 - y_0)$$

$$P(\theta|y) \propto \text{Binom}(y|\theta, n) \text{Beta}(\theta|y_0, n_0 - y_0)$$

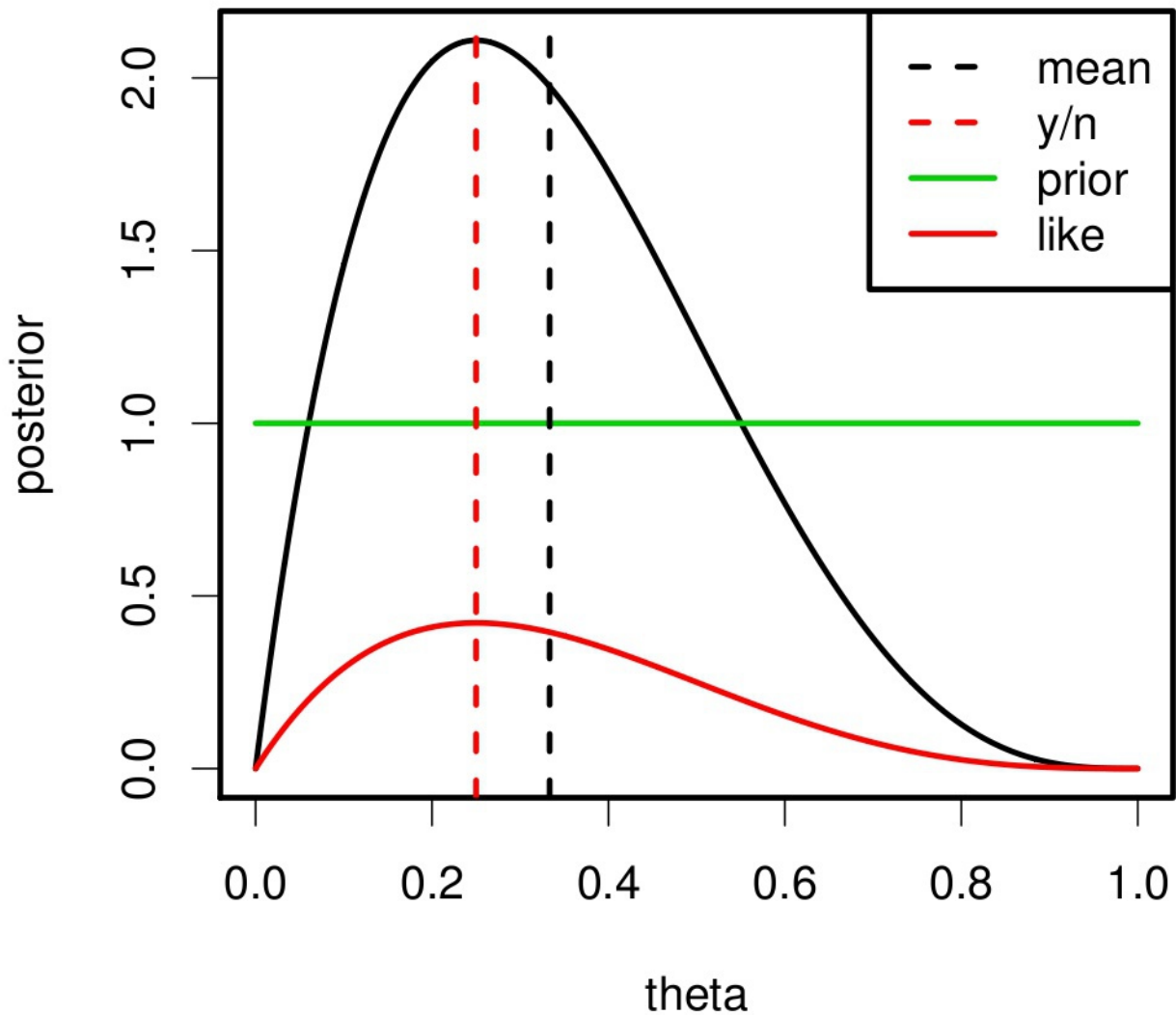
$$P(\theta|y) \propto \binom{n}{y} \theta^y (1-\theta)^{n-y} \times \frac{\theta^{y_0-1} (1-\theta)^{n_0-y_0-1}}{\text{B}(y_0, n_0 - y_0)}$$

$$P(\theta|y) \propto \theta^{y+y_0-1} (1-\theta)^{n-y+n_0-y_0-1}$$

$$P(\theta|y) = \text{Beta}(\theta|y + y_0, n + n_0 - y - y_0)$$

**Beta-Binomial Model**

$$P(\theta|y) = \text{Beta}(\theta|y + y_0, n + n_0 - y - y_0)$$
$$= \text{Beta}(\theta|2,4)$$



# How much impact does the prior have on the analysis?

