



# Intro to Bayesian Analysis

*1st IAA-CSIC Severo Ochoa School on Statistics, Data Mining, and Machine Learning*  
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***This exercise is from Eadie et al (2019), Journal of Statistics Education***

# Introducing Bayesian Analysis

*with a tasty example!*





What is the probability of drawing a **blue** m&m's<sup>®</sup> from an individual bag of m&m's<sup>®</sup>?

What is the percentage of **blue** m&m's<sup>®</sup> made at the factory?



What is the probability of drawing a **blue** m&m's<sup>®</sup> from an individual bag of m&m's<sup>®</sup>?

What is the percentage of **blue** m&m's<sup>®</sup> made at the factory?

**No online searches, that spoils the fun!**




**What do you think is the simplest way to estimate the number of blue m&m's made at the factory?**



**What do you think is the simplest way to estimate the number of blue m&m's made at the factory?**

**What are the pitfalls to this approach?**





**What if I gave you only one bag of  
m&m's?**

**What if I only let you use the first 10  
m&m's from that bag as your data?**



# Quick Review: Bayes' Theorem

maybe shoes  
get to go

$P(B)$   
(unknown)

$$P(A \cap B) = \frac{P(A \cap B)}{P(B)}$$

posterior

$$P(B|A) = \frac{P(A \cap B)}{P(A)}$$

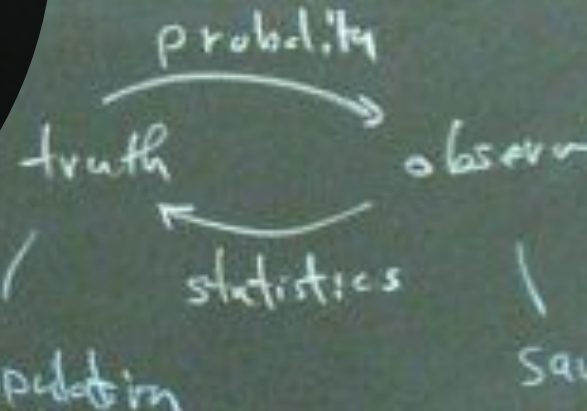
observe

likelihood

$$P(A|B) = \frac{P(A \cap B)}{P(B)}$$

$$P(A \cap B) = P(A|B) P(B)$$

$$P(B|A) = \frac{P(B \cap A)}{P(A)}$$



# Bayes' Theorem

$$p(\boldsymbol{\theta}|\boldsymbol{y}) = \frac{p(\boldsymbol{y}|\boldsymbol{\theta}) p(\boldsymbol{\theta})}{p(\boldsymbol{y})}$$

$\boldsymbol{\theta} \rightarrow$  model parameters

$\boldsymbol{y} \rightarrow$  data

# Bayes' Theorem

$$p(\boldsymbol{\theta}|\mathbf{y}) \propto p(\mathbf{y}|\boldsymbol{\theta}) p(\boldsymbol{\theta})$$

$\boldsymbol{\theta} \rightarrow$  model parameters

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# Bayes' Theorem

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

$\theta \rightarrow$  model parameters

$y \rightarrow$  data

Posterior distribution

# Bayes' Theorem

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

Likelihood

Posterior distribution

# Bayes' Theorem

The diagram illustrates Bayes' Theorem with the equation  $p(\theta|y) \propto p(y|\theta)p(\theta)$ . The terms are enclosed in colored ovals: a red oval for  $p(\theta|y)$ , a blue oval for  $p(y|\theta)$ , and a green oval for  $p(\theta)$ . Arrows point from labels below to these terms: a red arrow from 'Posterior distribution' to  $p(\theta|y)$ , a blue arrow from 'Likelihood' to  $p(y|\theta)$ , and a green arrow from 'Prior distribution' to  $p(\theta)$ .

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

Posterior distribution

Likelihood

Prior distribution

# m&m's<sup>®</sup> Activity



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

# m&m's<sup>®</sup> Activity



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

Find the posterior distribution for the percentage of **blue** m&m's<sup>®</sup> made at the factory – *using Bayes' theorem and one bag of m&m's.*



# m&m's<sup>®</sup> Activity



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

$\theta \rightarrow$  What does this represent?

$y \rightarrow$  What does this represent?

# m&m's<sup>®</sup> Activity



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

$\theta \rightarrow$  percentage of **blue**  
m&m's<sup>®</sup> made at the factory

$y \rightarrow$  data (m&m's<sup>®</sup>)

# Initial Questions (3 minutes)



1. What kind of data are m&m's?
  - a. Numerical
  - b. Categorical
  - c. Continuous
2. How will you record the data?
3. Will you sample with replacement or without?

# Initial Questions



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# Initial Questions



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2. How will you record the data? Table (all colours)
3. Will you sample with replacement or without?



## What's the sampling distribution? (3 min)

*i.e. How might we model the probability of drawing a blue m&m?*

**Likelihood** of drawing **y** blue  
**m&m's**<sup>®</sup> given **n** trials:

**y** → # of successes (blue **m&m**<sup>®</sup>)

**n-y** → # of failures (not a blue **m&m**<sup>®</sup>)

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



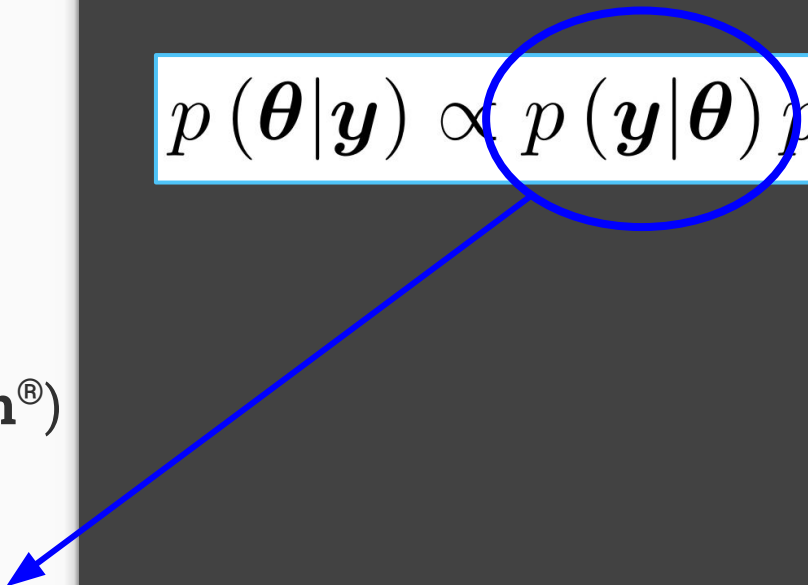
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Binomial Distribution:

$$p(\mathbf{y}|\boldsymbol{\theta}) \propto \boldsymbol{\theta}^y (1-\boldsymbol{\theta})^{n-y}$$

$$p(\boldsymbol{\theta}|\mathbf{y}) \propto p(\mathbf{y}|\boldsymbol{\theta}) p(\boldsymbol{\theta})$$


**Likelihood** of drawing **y** blue  
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**Prior Distribution:**

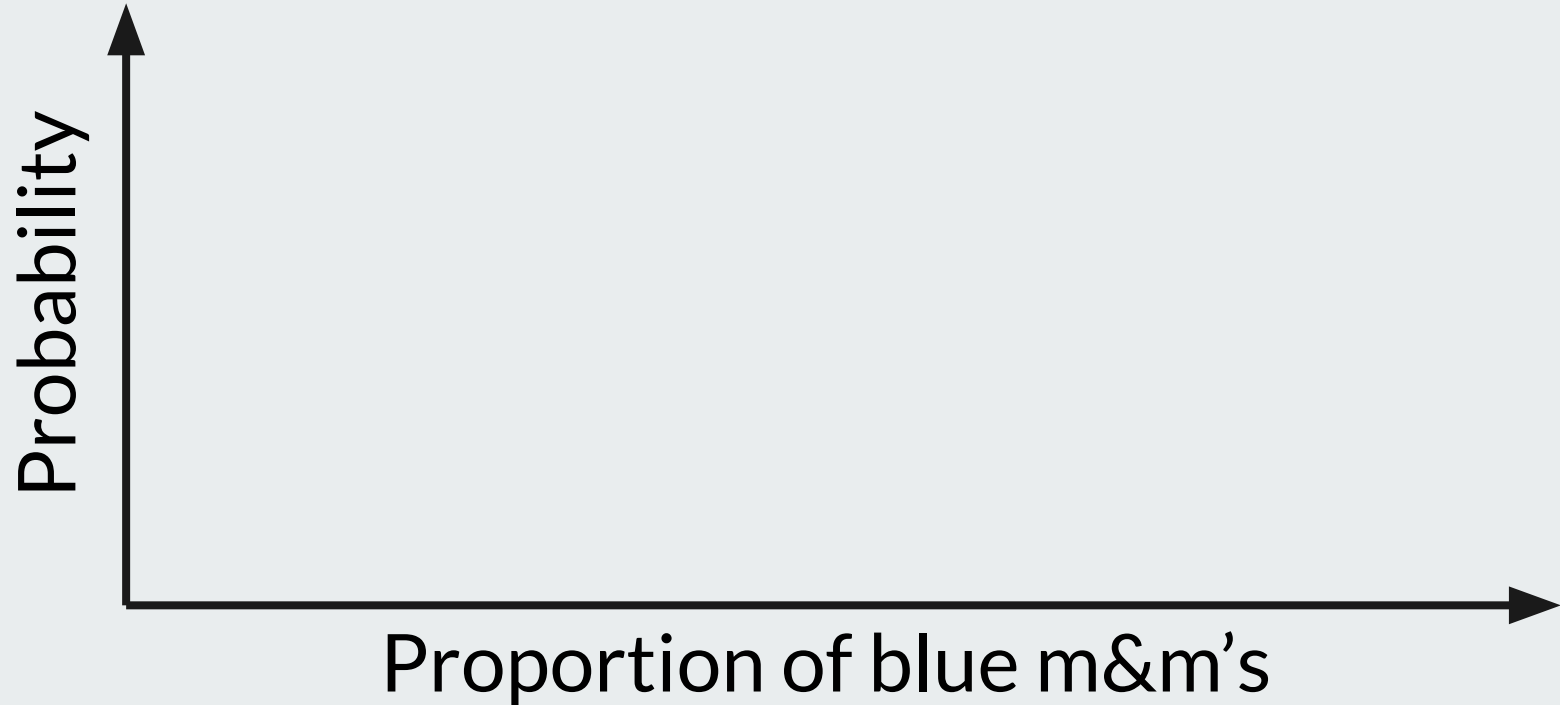
use prior knowledge about  
percentage of blue m&m's<sup>®</sup>

# What's our prior information?

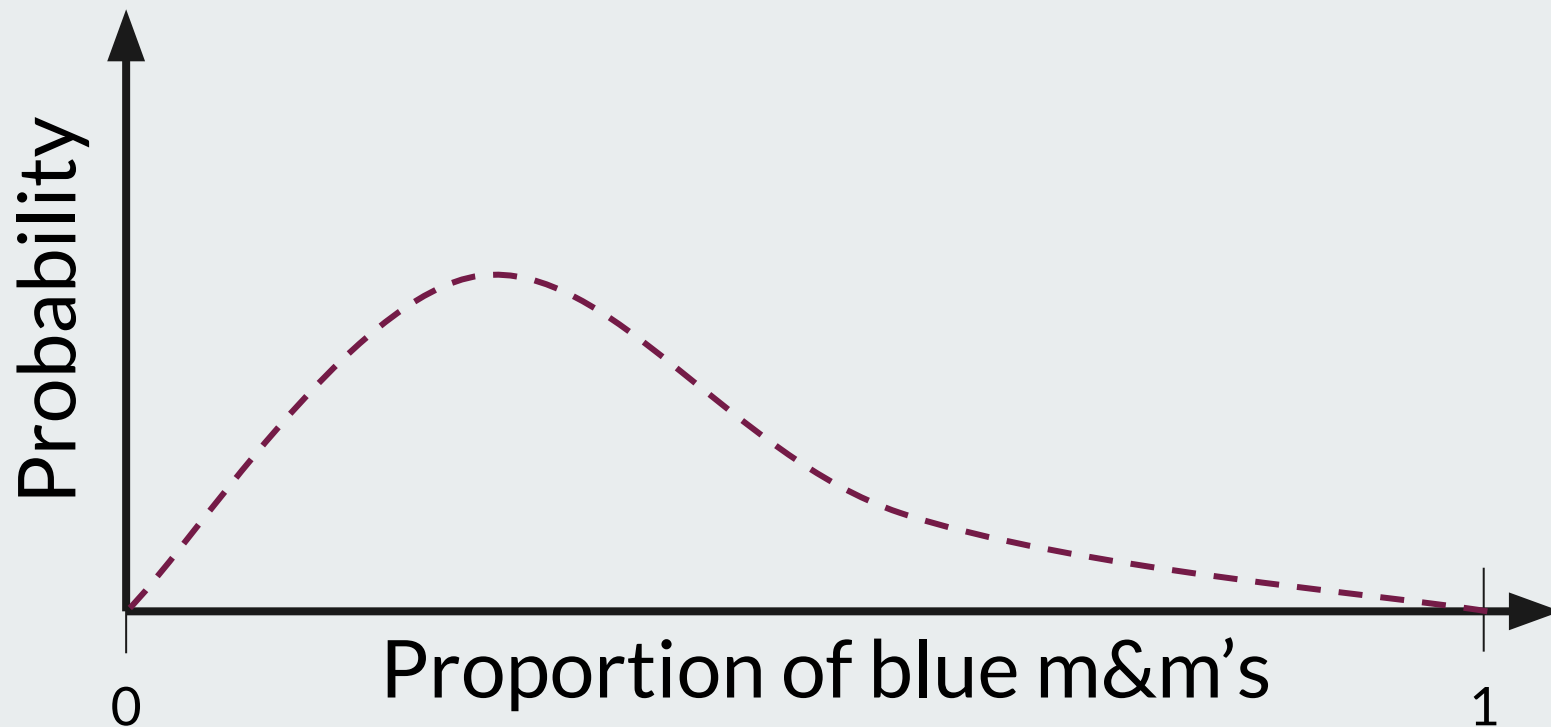


- How many different colours of m&m's are there?
- Do you think the m&m's are well-mixed before they go into a bag at the factory?
- What percentage of blue m&m's do you think are made at the factory?
- Do you think every bag will have the same percentage of blue m&m's?

# Sketch prior knowledge (5 min)



# Sketch prior knowledge



**Likelihood** of drawing **y** blue  
m&m's<sup>®</sup> given **n** trials:

**y** → # of successes (blue m&m's<sup>®</sup>)

**n-y** → # of failures (not a blue m&m's<sup>®</sup>)

Binomial Distribution:

$$p(\mathbf{y}|\boldsymbol{\theta}) \propto \boldsymbol{\theta}^y (1-\boldsymbol{\theta})^{n-y}$$

$$p(\boldsymbol{\theta}|\mathbf{y}) \propto p(\mathbf{y}|\boldsymbol{\theta}) p(\boldsymbol{\theta})$$

**Prior Distribution:**

use prior knowledge about  
percentage of blue m&m's<sup>®</sup>

→ **Conjugate prior**

Beta Distribution:

$$p(\boldsymbol{\theta}) \propto \boldsymbol{\theta}^{\alpha-1} (1-\boldsymbol{\theta})^{\beta-1}$$

# Prior Distribution hyperparameters

$\alpha = ?$

$\beta = ?$

$$p(\theta) \propto \theta^{\alpha-1} (1-\theta)^{\beta-1}$$

## Prior Distribution hyperparameters

$\alpha = ?$

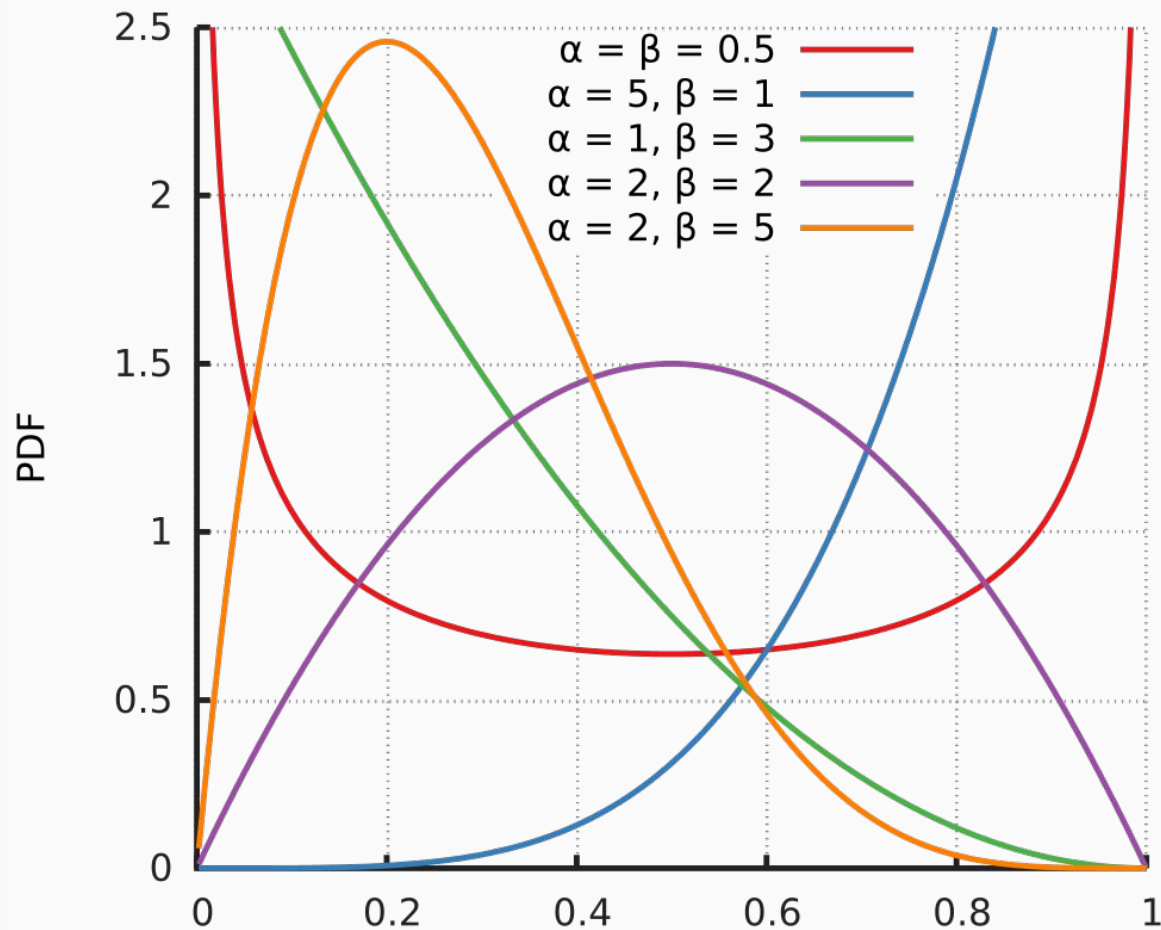
$\beta = ?$

$$p(\theta) \propto \theta^{\alpha-1} (1-\theta)^{\beta-1}$$

- Find values for the  $\alpha$  and  $\beta$  hyperparameters that best match your sketched prior distribution
- Some values to try:
  - when  $\alpha = \beta$
  - when  $\alpha = \beta = 1$



Here are some  
examples of beta  
distributions

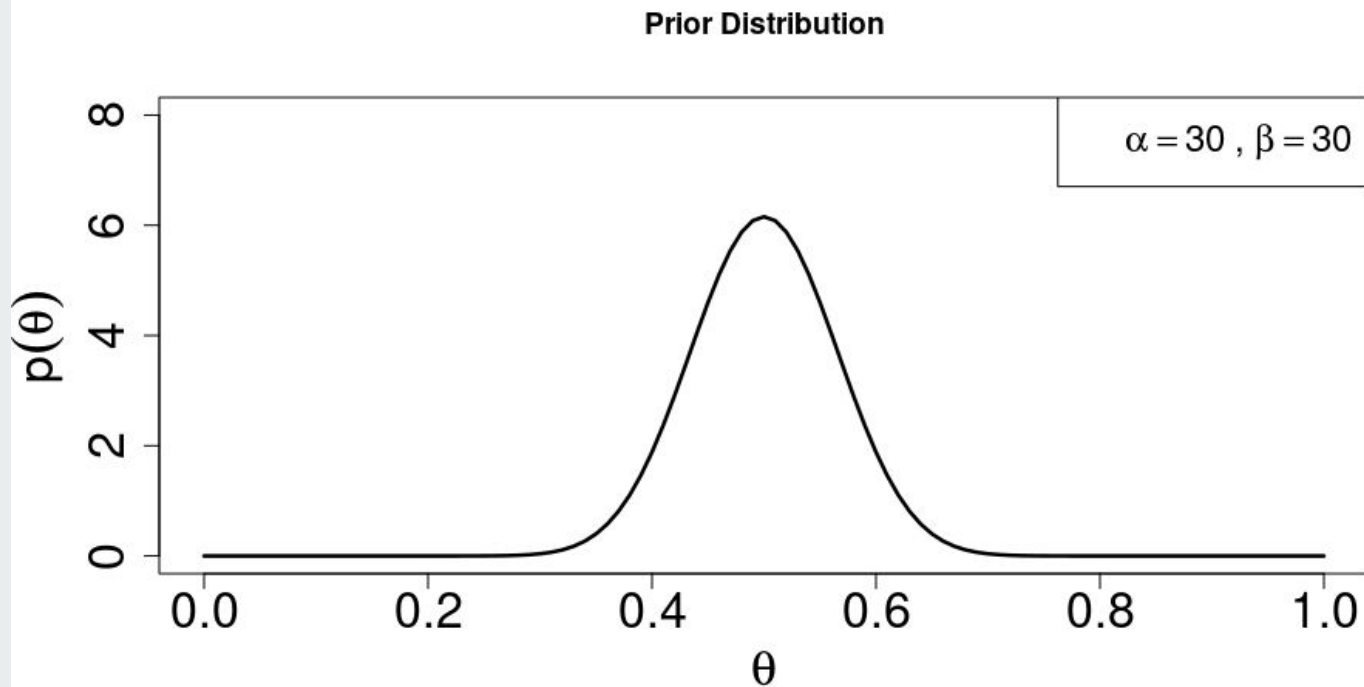


# In RStudio... (15 minutes)

1. Write a script to plot the beta distribution given the alpha and beta parameter values (*Hint*: check out ?Distributions)
2. What happens when:  $\alpha = \beta$  ?  $\alpha = \beta = 1$  ?
3. Find values for the  $\alpha$  and  $\beta$  hyperparameters that best match your sketched prior distribution

# Now you have a prior:

$$p(\theta) \propto \theta^{\alpha-1}(1-\theta)^{\beta-1}$$



$$p(\theta/y) \propto p(y/\theta) p(\theta)$$

Binomial Distribution:

$$p(\mathbf{y}|\theta) \propto \theta^y (1-\theta)^{n-y}$$


$$p(\theta|\mathbf{y}) \propto p(\mathbf{y}|\theta) p(\theta)$$

Binomial Distribution:

$$p(\mathbf{y}|\theta) \propto \theta^y (1-\theta)^{n-y}$$

Beta Distribution:

$$p(\theta) \propto \theta^{\alpha-1} (1-\theta)^{\beta-1}$$


$$p(\theta|\mathbf{y}) \propto p(\mathbf{y}|\theta) p(\theta)$$

Binomial Distribution:

$$p(\mathbf{y}|\theta) \propto \theta^y (1-\theta)^{n-y}$$

Beta Distribution:

$$p(\theta) \propto \theta^{\alpha-1} (1-\theta)^{\beta-1}$$


$$p(\theta|\mathbf{y}) \propto p(\mathbf{y}|\theta) p(\theta)$$

Simplify the expression

Binomial Distribution:

$$p(\mathbf{y}|\theta) \propto \theta^y (1-\theta)^{n-y}$$

Beta Distribution:

$$p(\theta) \propto \theta^{\alpha-1} (1-\theta)^{\beta-1}$$


$$p(\theta|\mathbf{y}) \propto p(\mathbf{y}|\theta) p(\theta)$$

$$p(\theta|\mathbf{y}) \propto \theta^{y+\alpha-1} (1-\theta)^{n-y+\beta-1}$$



What kind of distribution is this? (5 minutes)

$$p(\theta|y) \propto \theta^{y+\alpha-1} (1-\theta)^{n-y+\beta-1}$$

What kind of distribution is this?

$$p(\theta|y) \propto \theta^{y+\alpha-1} (1-\theta)^{n-y+\beta-1}$$

Write a script that will plot  $p(\theta|y)$   
given  $y$ ,  $n$ ,  $\alpha$ , and  $\beta$

Gather some data!



# Gather some data! (10 minutes)



- Take out the m&m's ( $n=?$ )
- Record number of each colour
- Plot the posterior

# Think-Pair-Share (10 minutes)

- Is the posterior distribution what you expected?
- Compare the posterior distribution to the prior distribution
- Is this the result you expected, given six different colours?  
Does this result tell you about the percentages of the other colours?
- How sensitive is the posterior to the prior distribution?

# Think-Pair-Share

- How would you expect the posterior to change given more data?

# Think-Pair-Share

- How would you expect the posterior to change given more data?
- Let's pool all the data to find out!
  - What assumptions are we making here?

# Let's try looking at the red ones



- How many were red in the bag?
- Calculate the posterior for the red ones



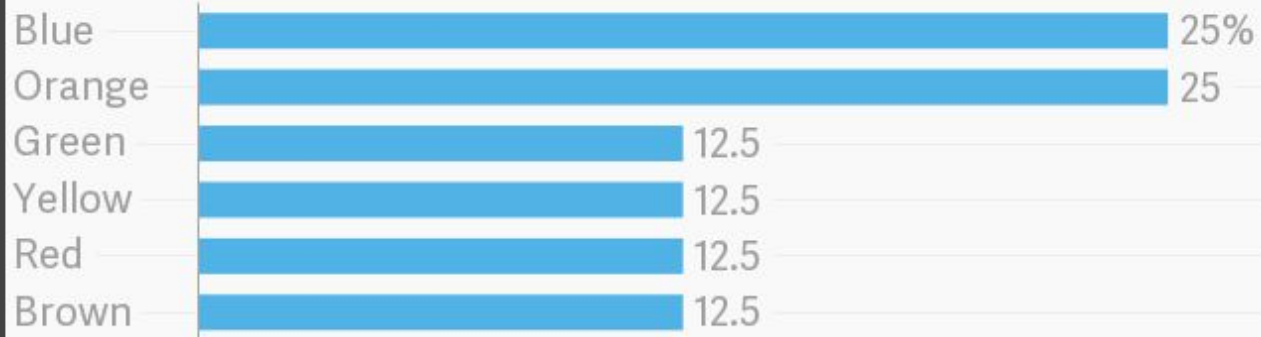
*Surprise twist!*

*Surprise twist!*

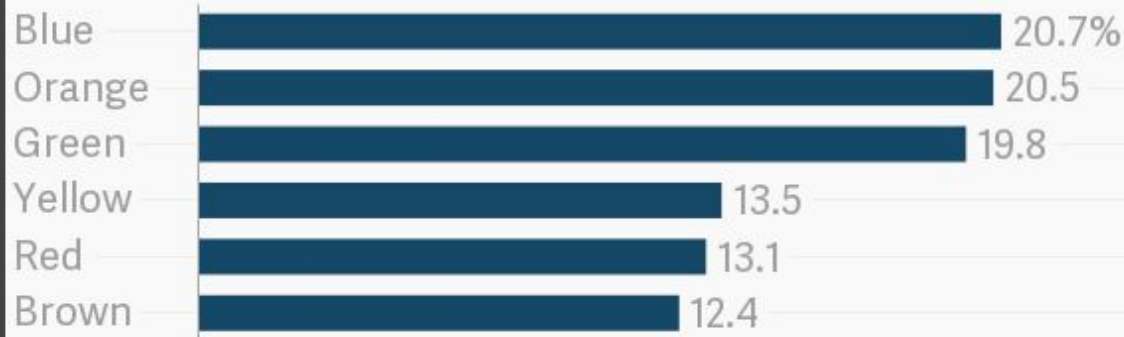
Different factories make  
different colour  
distributions of m&m's !

## M&Ms color distribution, c. 2017

### New Jersey factory



### Tennessee factory

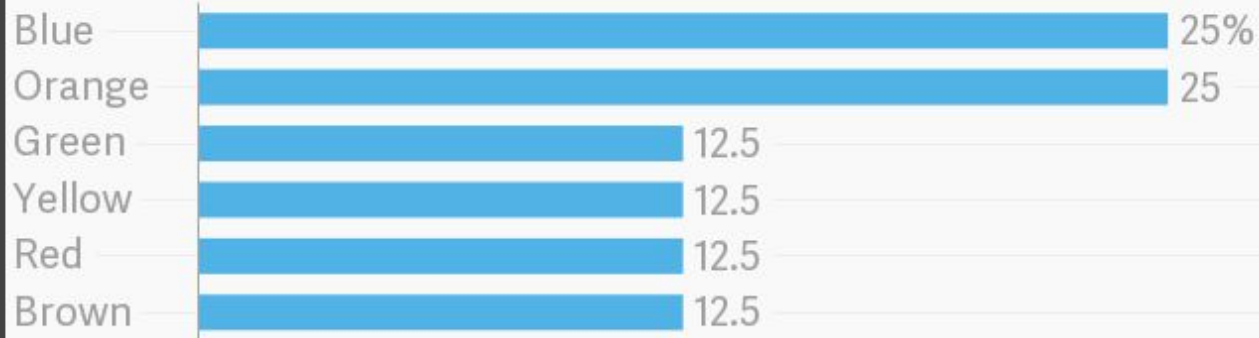


# Which factory did our m&m's come from?

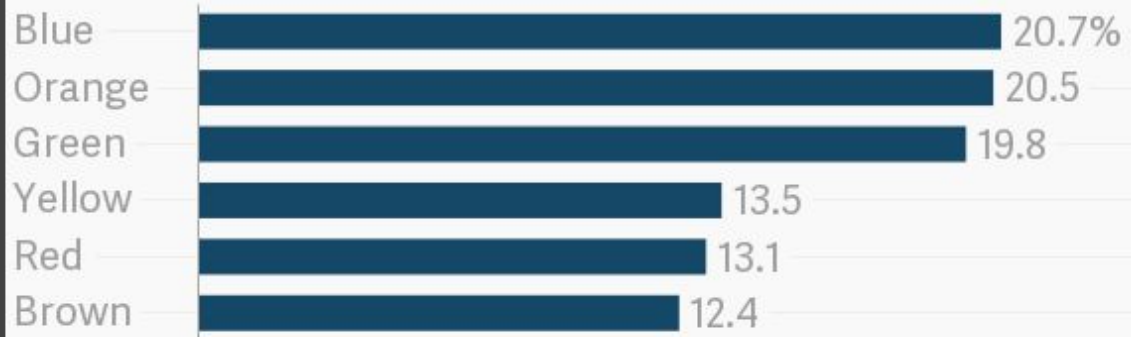


M&Ms color distribution, c. 2017

## New Jersey factory



## Tennessee factory



New Jersey = HKP

Tennessee = CLV



**Thomas Bayes  
(1701-1761)**



Image: Evan-Amos, Vanamo Media

What if we wanted to model all the colours simultaneously?

What distribution might we use for the likelihood?



Image: Evan-Amos, Vanamo Media

Wikipedia (different notation than we've been using)

What if we wanted to model all the colours simultaneously?

What distribution might we use for the likelihood?

Multinomial Distribution

$$f(x_1, \dots, x_k; p_1, \dots, p_k) = \frac{\Gamma(\sum_i x_i + 1)}{\prod_i \Gamma(x_i + 1)} \prod_{i=1}^k p_i^{x_i}.$$



# Conjugate prior to the multinomial distribution is the Dirichlet distribution

Wikipedia (different  
notation than we've  
been using)

$$\frac{1}{B(\boldsymbol{\alpha})} \prod_{i=1}^K x_i^{\alpha_i - 1}$$

$$B(\boldsymbol{\alpha}) = \frac{\prod_{i=1}^K \Gamma(\alpha_i)}{\Gamma(\sum_{i=1}^K \alpha_i)}$$

$$\boldsymbol{\alpha} = (\alpha_1, \dots, \alpha_K)$$

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# Toy problem



I hand out bags of m&m's, but the lot codes are hidden on the back.

# Toy problem



I hand out bags of m&m's, but the lot codes are hidden on the back.

I tell you that some people have m&m's from New Jersey and others from Tennessee.

# Toy problem



I hand out bags of m&m's, but the lot codes are hidden on the back.

I tell you that some people have m&m's from New Jersey and others from Tennessee.

But I don't tell you what proportion are from NJ!

# Develop a model that will tell you



1. Which factory your m&m's bag came from
2. The proportion of bags from NJ
3. The colour distributions produced at each factory

# Hierarchical Bayesian Model

let  $c_b = \#$  of each colour of m&m's in bag  $b$

$$c_b \sim \text{Multinomial}(\beta_f | z_b)$$

colour dist. produced  
at factory  $f$

$$\vec{\beta}_1 = (\beta_{1, \text{red}}, \beta_{1, \text{blue}}, \beta_{1, \text{green}}, \dots)$$

$$\vec{\beta}_2 = (\beta_{2, \text{red}}, \beta_{2, \text{blue}}, \beta_{2, \text{green}}, \dots)$$

$$\beta_f \sim \text{Dirichlet}(\eta) \rightarrow \vec{\eta} = (\eta_{\text{red}}, \eta_{\text{blue}}, \eta_{\text{green}}, \dots)$$

latent variable that  
assigns bag  $b$  to a factory  $f$

$$z_b \sim \text{Bernoulli}(\theta)$$

% of bags from NJ

$$\theta \sim \text{Dirichlet}(\alpha)$$

hyperparameter

(Picture taken of my  
notes, after class... need  
to make nicer slides of  
this!)

You may visit the link below to input data from  
an entire bag of plain m&m's.

This will help us keep track of the colour distributions of  
m&m's for future classes

*(please make sure to note the factory code!)*

<http://bit.ly/MandMsCounts>