



# A practical introduction to Bayesian Inference

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Prof. Gwendolyn Eadie

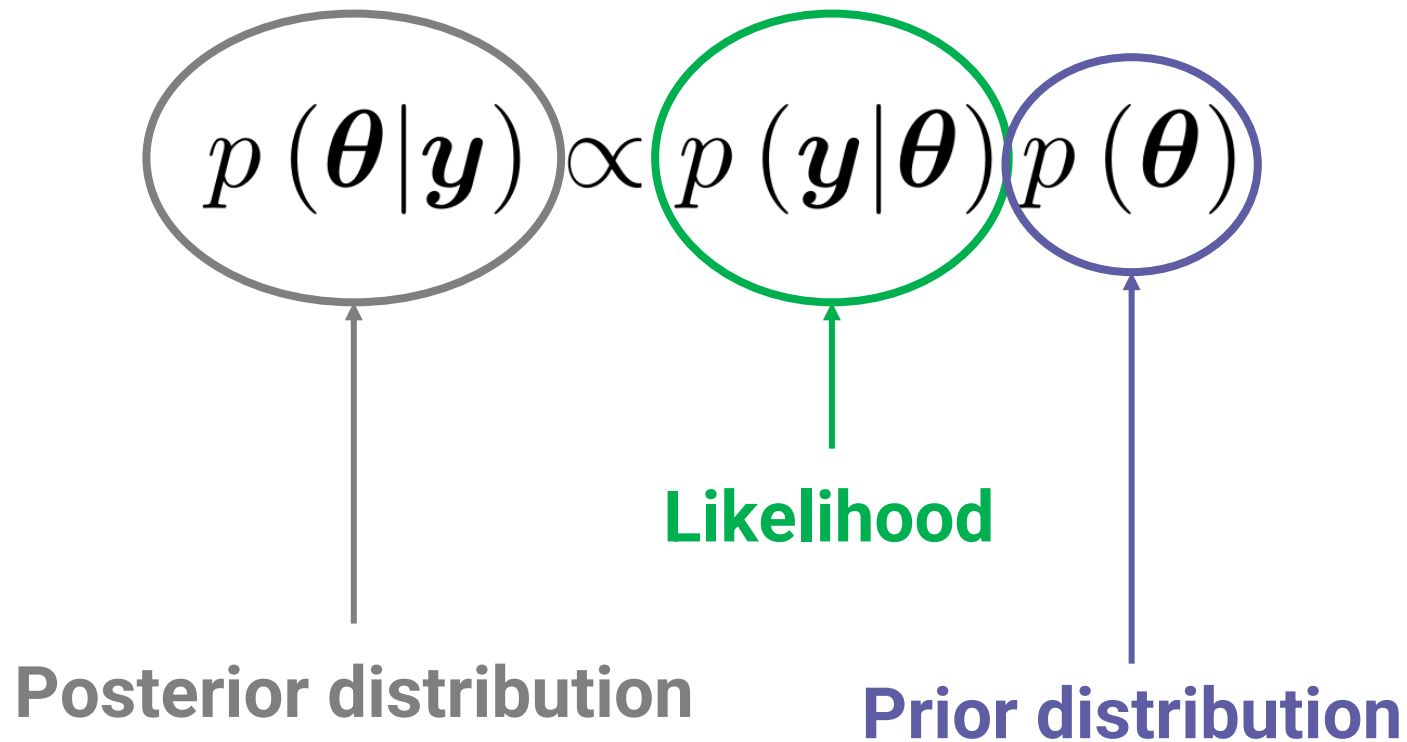
UofT Statistical Sciences Research Program



# Bayes' Theorem

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

- $\theta \rightarrow$  vector of model parameters
- $y \rightarrow$  data



# Bayesian Inference

- Parameters are not fixed
- The quantity of interest is the distribution of the parameters
- We want to calculate exactly or, if we can't, estimate the *posterior distribution*

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

- Computational methods for estimating the posterior
  - Grid of  $\theta$ 's  $\rightarrow$  evaluate the posterior at every point, given the data (\$\$\$)
  - Drawing samples from the posterior
    - E.g., Markov Chain Monte Carlo, Hamilton Monte Carlo, Gibbs Sampling, or some other sampling method
  - Variational Bayes
  - Other methods

# Today's activity: Bayesian inference with m&m's

- Find the **posterior distribution** for the percentage of **blue** m&m's<sup>®</sup> made at the factory — *using Bayes' theorem and 15 candies from a bag of m&m's.*

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

*No googling for answers!  
That spoils the fun.*



# Initial exploratory questions

- What kind of data are the colours of m&m's?
  - Numerical
  - Categorical
  - Continuous
- How will you record the data?
- Will you sample with replacement or without?
- How might we model the probability of drawing a blue m&m?

# What's (y)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 (or at least think about) what your prior distribution for  $\theta$  looks like

**Sketch of a prior distribution for  
the percentage of blue m&m's made at the factory**

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

$y \rightarrow$  # of successes (blue m&m's<sup>®</sup>)

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

Binomial Distribution:

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

Prior knowledge

- Outlined/sketch previously
- For convenience, use the conjugate prior
- find parameter values that make the conjugate prior look like our prior distribution sketch.

Beta Distribution:

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



Binomial Distribution:

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

Beta distribution:

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

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

# In-Class Activity

- Go to the R markdown or Jupyter notebook on Github
  1. write a function for the prior that takes  $\alpha$  and  $\beta$  as inputs
  2. Write code to plot the prior distribution
  3. Choose parameters to define the prior distribution
    - a) adjust the  $\alpha$  and  $\beta$  hyperparameters that best match your sketched prior distribution
  4. Calculate the mean and variance of the prior you chose

# Exercise

- Gather some data!
- Open your bag of m&m's
- Take out 15 , and record how many colours you have of each
- Perform Bayesian inference to predict the percentage of blue m&m's made at the factory



Please help us collect data!  
[bit.ly/mmcounts](https://bit.ly/mmcounts)



# Think-Pair-Share

- 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?

# Pool the data from the entire class

- How would you expect the posterior to change given more data?
- **Let's pool the data and find out!**
  - (What assumptions are we making here?)
- Plot the posterior again
- Compare to prior