# Naïve Bayesians

**Back to Basics Series** 

16 Jan 2021

# Developing the Bayesian muscle to solve a wide range of problems

# **Naïve Bayesian Philosophy**

Intuitive (Visual)
Understanding of the
Bayesian Reasoning

Ability to model real world problems in a Bayesian Setting

Starting from Simple Probabilistic modelling

Adapting it in a a Bayesian setting
And moving towards ML models

Fluency in the Calculus of Bayesian Stats & ML model



#### **Season 2: Back to Basics**



# **Back to Basics**

		Canonical Problem	Applications
Ep 1	Bayes Theorem	There are 2 boxes from which cookies can be taken from. Box A and Box B. Box A contains 10 chocolate cookies, Box B contains 5 ginger cookies. Given that you get a chocolate cookie which box was it taken from?	The Shy Librarian Problem Naive Bayes algorithm
Ep 2	Problems with Binomial Likelihoods	You have 2 coins C1 and C2. p(heads for C1) = .7 & P(heads for C2) = 0.6 You flip the coin 10 times. What is the probability that the given coin you picked is C1 given you have 7 heads and 3 tails?	A/B Testing
Ep 4	Disease Detection	A particular disease affects 1% of the population. There is an imperfect test for this disease: The test gives a positive result for 90% of people who have the disease, and 5% of the people who are disease-free. Given a positive test result – what is the probability of having the disease?	COVID Tests (PCR & Antibody)! Fraud Detection
Ep 5	Naive Bayes Classification	Given these words occur in this text what's the probability it's spam?	
Ep 6	Gaussian Naive Bayes Classification	Given the weights and heights of basketball players, what's the probability that person a is a basketball player given weight = w and height = h?	Any Classification Problem

#### **Back to Basics**

Canonical Problem **Applications** Ep 7 Suppose tanks were given a serial number based on the order in which they were German Tank manufactured. Given that you've observed a tank with serial number "10", how Problem many tanks were actually manufactured in total? Waiting Times Suppose you need to gather 10 patients for a trial. Each signup happens at time Planning Trials t\_i (i=1, 10). How long do you have to wait after it took you 3 weeks to accrue 2 (Continuous Estimating Queues Distributions) signups?

# **Bayes Rule**

Posterior Likelihood Prior 
$$P(\theta \mid D) = P(D \mid \theta) P(\theta)$$

$$P(D)$$

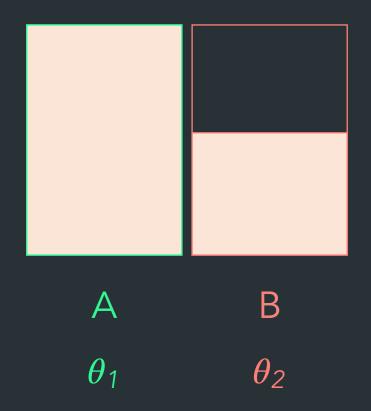
$$P(D)$$
Normalising Constant

#### **Bayes Rule**

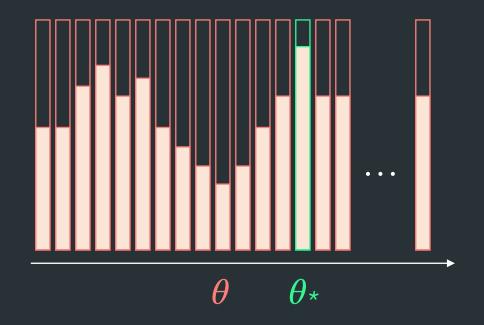
Posterior Likelihood Prior 
$$P(\theta_i \mid D) = P(D \mid \theta_i) P(\theta_i)$$

$$\sum_{all j} P(D \mid \theta_j) P(\theta_j)$$
Normalising Constant

# Visual representation of Bayes Rule for 2 Hypotheses



# What about continuous parameters?



Moving from a discrete set of hypothesis to a continuous parameters can be thought of having an infinite number of "boxes" to choose from

#### **Bayes Rule**

Posterior Likelihood Prior 
$$P(\theta \mid D) = P(D \mid \theta) P(\theta)$$

$$\int P(D \mid \theta') P(\theta') d\theta'$$
Normalising Constant

# Recap from Last Week | Canonical Problem

You have 2 coins  $C_1$  and  $C_2$ . p(heads for  $C_1$ ) = 0.7 p(heads for  $C_2$ ) = 0.6

You flip one of the coins 10 times and get 7 heads and 3 tails

What is the probability that the given coin you picked is  $C_1$ ?

#### **Canonical Problem**

You flip one of the coins 10 times and get 7 heads and 3 tails

What is the probability of getting a head in the next flip?

#### Real world Problems to Solve

- Comparing Amazon Reviews
  - Seller A has 7 positive ratings with 10 reviews
  - Seller B has 70 positive ratings with 100 reviews
  - Seller C has 140 positive ratings with 200 reviews

- A/B Testing:
  - Webpage A has 30% conversion rate (1200 users)
  - Webpage B has a 35% conversion rate (1300 users)

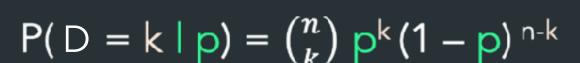
#### 7 out of 10 reviews are good what is the probability of getting a good review?

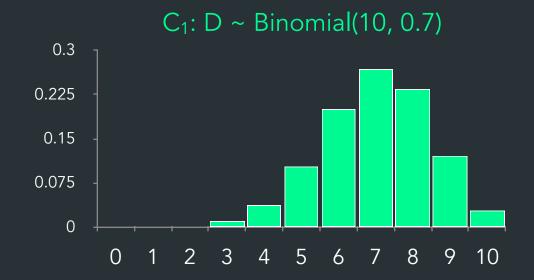
$$n = 10, k = 7$$
  
Expected value

- 7/10 = 70%
- How confident are you?
- What if 70 out of 100 reviews were positive?
- What if you knew that it was more likely for the data to tampered?

# What is the probability that the given coin you picked is C<sub>1</sub>?

- D ~ Binomial(n, p)
  - $0 \le D \le n$
  - n > 0
  - 0



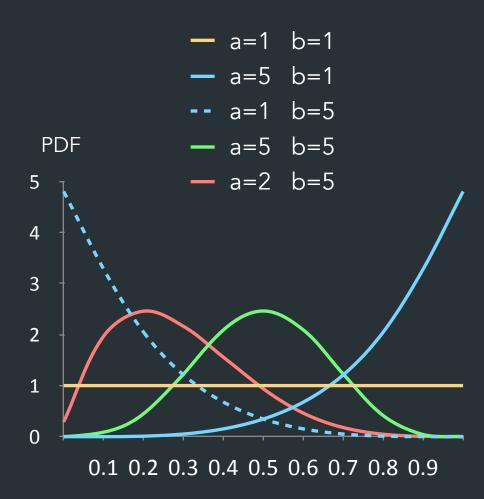


#### **Beta Distribution**

- $\theta \sim \text{Beta(a, b)}$ 
  - $0 \le \theta \le 1$
  - a > 0, b > 0
- Flexible family of distribution
- Conjugate prior to the Binomial

$$P(\theta) = c \theta^{a-1} (1 - \theta)^{b-1}$$

$$E(\theta) = \underbrace{a + b} \qquad Mode = \underbrace{a - 1}_{a + b - 2}$$

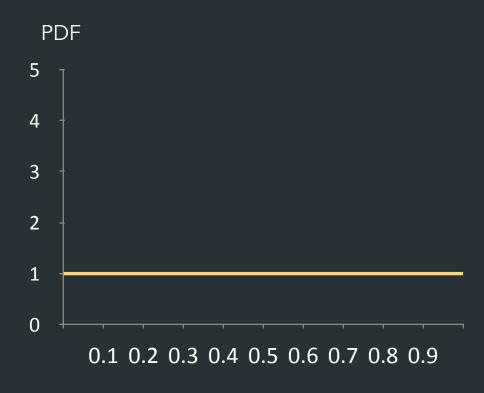


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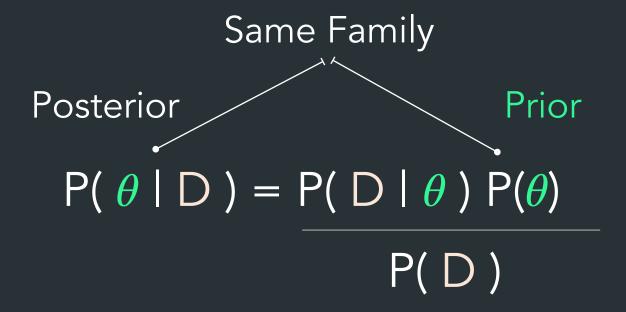
# **Bayes Rule**

Posterior Likelihood Prior 
$$P(\theta \mid D) = P(D \mid \theta) P(\theta)$$

$$P(D)$$

$$P(D)$$
Normalising Constant

# **Conjugate Prior**



- 1. Let p be probability of getting a head
  - Assume p follows a Beta(1, 1) distribution
- 2. Let D be the number of heads in a set of 10 flips
  - Assume D follows a Binomial(10, p) distribution

$$P(p|D) = P(D|p) P(p)$$

$$P(D)$$

#### **Bayes Rule**

Posterior Likelihood Prior 
$$P(\theta \mid D) = P(D \mid \theta) P(\theta)$$

$$\int P(D \mid \theta') P(\theta') d\theta'$$
Normalising Constant

$$P(p \mid D=k) \propto P(D=k \mid p) P(p)$$

$$\propto p^{k} (1-p)^{n-k} p^{a-1} (1-p)^{b-1}$$

$$\propto p^{a+k-1} (1-p)^{b+n-k-1}$$

$$= p^{a'-1} (1-p)^{b'-1}$$

$$= Beta(a', b')$$

$$= Beta(a+k, b+n-k)$$

$$= Beta(a', b')$$

$$= Beta(k+1, n-k+1) \quad a = 1, b=1$$

$$n = 10, k = 7$$
  
Expected value

- 7/10 = 70%
- How confident are you?
- What if 70 out of 100 reviews were positive?
- What if you knew that it was more likely for the data to tampered?

#### Before

$$n = 10, k = 7$$

Expected value

- 7/10 = 70%
- How confident are you?
- What if 70 out of 100 reviews were positive?
- What if you knew that it was more likely for the data to tampered?

#### After

- Start with Beta(1, 1)
- Update prior belief with Beta(1+7, 1+10-7)
- Expected value

$$= 8/12 = 67\%$$



# **Takeaways**

- Recap: Binomial distribution
- Continuous parameters / hypotheses can be estimated using Bayes Rule
- Conjugate priors allow us to analytically find posterior distributions
- Beta distribution is a conjugate prior for the Binomial distribution
- Estimate the probability of success of a coin flip using Bayes Rule

#### **Bernoulli Distribution**

- Y ~ Bernoulli(p)
  - X = 0, 1
  - 0
- Events with 1 Trial & 2 Possible Outcomes
- Examples
  - 1 coin flip
  - Answers to True/False Quiz
  - Voting in a 2-Candidate Election
  - Result of a COVID-19 Test

**Probability Mass Function (PMF)** 



