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### Information Retrieval

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So much of life, it seems to me, is determined by pure randomness.

– Sidney Poitier.

God does not play dice with the universe.

- Einstein.



## The Law – Robert M. Coates

From the book, "The World of Mathematics – Volume IV".

#### **Triborough Bridge, NY, USA.** (aka Robert F. Kennedy Bridge)



Late 1940s, NY: No other bridge or main highway was affected, and though the two preceding nights had been equally balmy and moonlit, on both of these the bridge traffic had run close to **normal**. And then, one day...



It just looked as if **everybody** in Manhattan who owned a car had decided to drive out to Long Island that evening.

### No Reason!



**Sergeant**: "I kept askin' them" he said, "Is there night football somewhere that we don't know about? Is it the races you're goin' to?"

But the funny thing was half the time they'd be askin' me. "What's the crowd for, Mac?" they would say. And I'd just look at them. If normal things stop happening, if we lose regularities in life, our planet could become unlivable!

### Time for Action

• At this juncture, it was inevitable that Congress should be called on for action.



 Senator said, "You can control it". Re-education and reforms were decided upon. He said, (we need to lead people back to) "the basic regularities, the homely averageness of the American way of life".

### The Law of Large Numbers

Known as the Fundamental theorem of Probability

The average of the results obtained from a large number of trials should be close to the expected value, and will tend to become closer as more trials are performed.

### Expectation

- Roll a dice. Assume you may see 1 to 6 with equal probability.
- Expected Value = ?

$$\frac{1+2+3+4+5+6}{6} = 3.5$$

According to the law of large numbers, if a large number of six-sided dice are rolled, the average of their values (sometimes called the sample mean) is likely to be close to 3.5, with the precision increasing as more dice are rolled. – Wikipedia.

# A Minor Digression

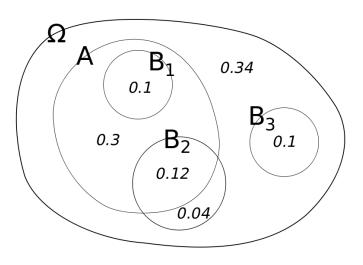
• What is the fundamental theorem of algebra?

# Quiz

- What is the fundamental theorem of algebra?
  - Loosely, "Every polynomial has root(s)"
  - More Precisely, "every non-constant singlevariable polynomial with complex coefficients has at least one complex root." [Source: Wikipedia].

### **Conditional Probability**

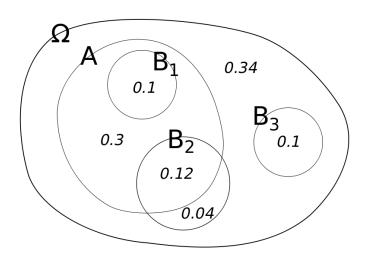
- P(A) = 0.52,
- P(B<sub>1</sub>) = 0.1, and so on as shown below
- What is  $P(A|B_1)$ ?
  - $P(A | B_1) = 1.$



Euler Diagram

### Quiz: Conditional Probability

• What is P(A|B<sub>2</sub>)?



Euler Diagram

• Developers in two companies are distributed as follows. Compute Joint Probabilities.

	Java	С	Total
Company-X	1	17	18
Company-Y	37	20	57
Total	38	37	75

	Java	С	Total
Company-X	0.013	0.227	0.24
Company-Y	??	??	??
Total	0.506	??	??

P(Company-X,Java) = 1/75 = 0.013

• Developers in two companies

	Java	С	Total		Java	С	Total
Company-X	1	17	18	Company-X	0.013	0.227	0.24
Company-Y	37	20	57	Company-Y	0.493	0.267	0.76
Total	38	37	75	Total	0.506	0.494	1

- Joint Probability P(Company-X, Java) = 0.013.
- P(Company-Y, Java) = 0.493
- Sometimes written as P(AB) or  $P(A \cap B)$

• Developers in two companies

	Java	С	Total
Company-X	1	17	18
Company-Y	37	20	57
Total	38	37	75

	Java	С	Total
Company-X	0.013	0.227	0.24
Company-Y	0.493	0.267	0.76
Total	0.506	0.494	1

- P(Company-Y|Java) = ??
- Is P(Company-Y|Java) == P(Java|Company-Y)?

• Developers in two companies

	Java	С	Total		Java	С	Total
Company-X	1	17	18	Company-X	0.013	0.227	0.24
Company-Y	37	20	57	Company-Y	0.493	0.267	0.76
Total	38	37	75	Total	0.506	0.494	1

- P(Company-Y|Java) = 37/38 = 0.974
- P(Java | Company-Y) = 37/57 = 0.649

### Odds

• Odds, O(A) = P(A)/P(A') = PA/(1-P(A))

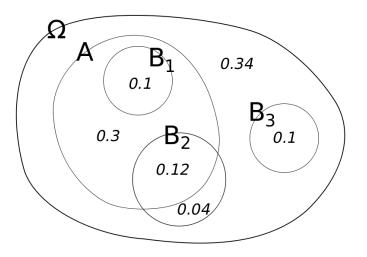
# Quiz

- What is the probability of getting a 5 when rolling a six sided die? Assume a fair die.
- What is the odds of the same event?

### Quiz: Conditional Probability

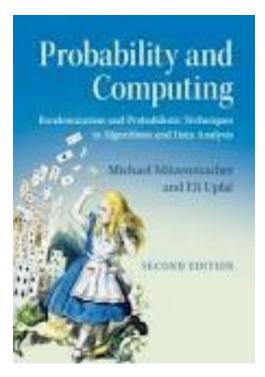
• What is P(A|B<sub>2</sub>)?

 $P(A|B_2) = 0.12/(0.12 + 0.04) = 0.75.$ 

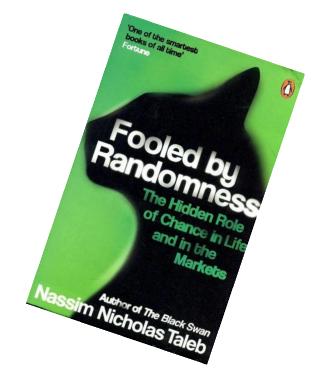


Euler Diagram

# Reading



Probability and Computing - Eli Upfal and Michael Mitzenmacher



Fooled by Randomness - Nassim Nicholas Taleb



Thomas Bayes, 1701 to 1761

### Bayesian Data Analysis and Beta Distribution

Venkatesh Vinayakarao

### Agenda

### Updating Beliefs using Probability Theory Role of Beta Distributions

Will Discuss
✓ Concepts
✓ Illustrations
✓ Intuitions
✓ Purpose
✓ Properties

#### Will not Discuss

- $\otimes$  Details
- $\odot$  Definitions
- $\odot$  Formalism
- $\odot$  Derivations
- $\otimes$  Proofs

## The Case of Coin Flips

General Assumption: If a coin is fair! Heads (H) and Tails (T) are equally likely. But, coin need not be fair  $\circledast$ 

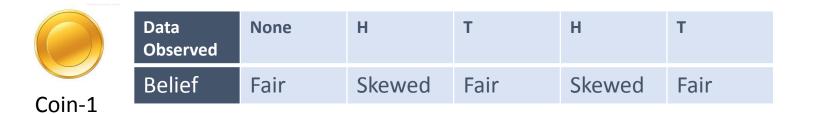


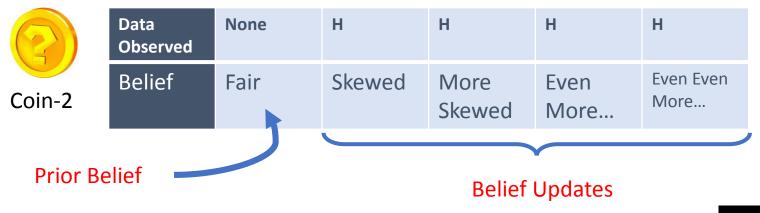
Experiment with **Coin - 1** HHHTTTHTHT Experiment with **Coin - 2** HHHHHHHHT

**Coin-1** more likely to be fair when compared to **coin-2**.

## Our Beliefs

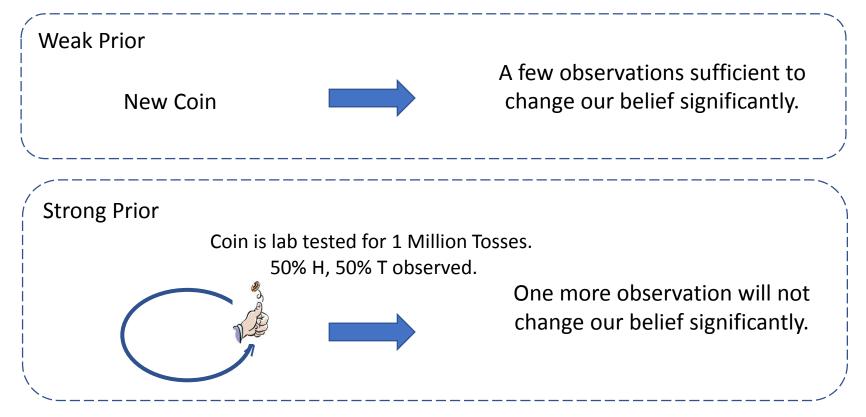
 Can we find a structured way to determine coin's nature?





### Priors

#### • Priors can be strong or weak



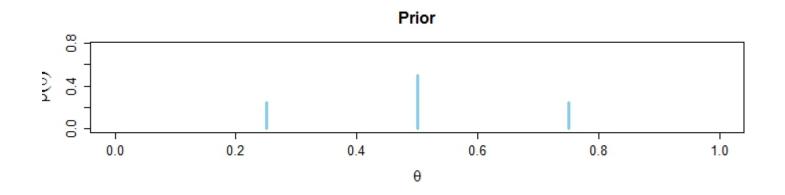
### HyperParameter

- Prior probability (of Heads) could be anything:
  - 0.5  $\rightarrow$  Fair Coin
  - 0.25  $\rightarrow$  Skewed towards Tails
  - 0.75  $\rightarrow$  Skewed towards Heads
  - 1  $\rightarrow$  Head is guaranteed!
  - 0  $\rightarrow$  Both sides are Tails.

We use  $\theta$  as a HyperParameter to visualize what happens for different values.

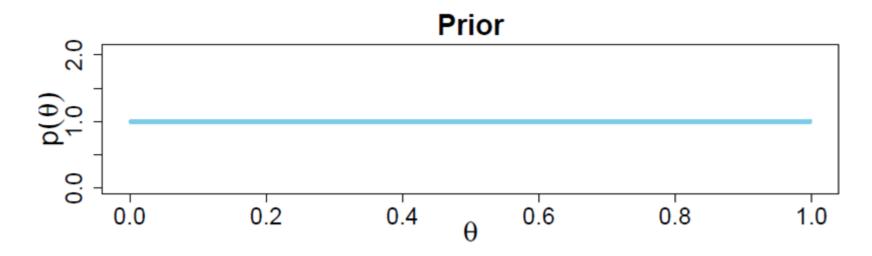
### World of Distributions

Discrete Distribution of Prior. Since I typically perceive coins as fair, Prior belief peaks at 0.5.



### Another Possibility

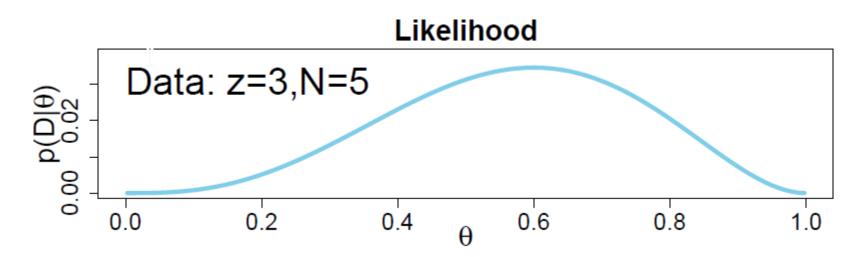
I may also choose to be unbiased! i.e.,  $\theta$  may take any value equally likely.



A Continuous Uniform Distribution!

### Observations

Let's flip the coin (N) 5 times. We observe (z) 3 Heads.

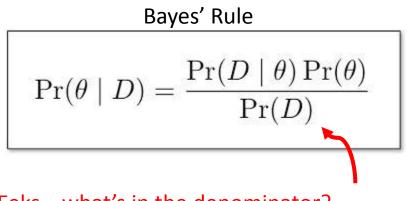


### Impact of Data

### Belief is influenced by Observations. But, note that:

### Belief ≠ observation





Eeks... what's in the denominator?

### Numerator is easy

- $p(\theta)$  was uniform. So, nothing to calculate.
- How to calculate  $p(D|\theta)$ ?



Jacob Bernoulli 1655 – 1705.

$$\theta^{z}(1-\theta)^{n-z}$$

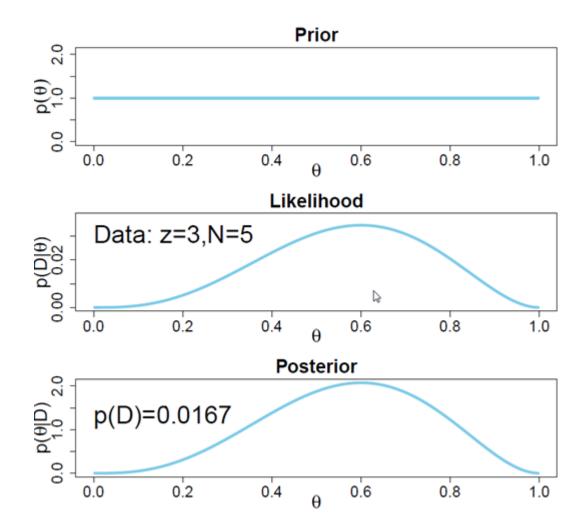
If D observed is HHHTT and  $\theta$  is 0.5, We have:  $p(D|\theta) = (0.5)^3(1-0.5)^{5-3}$ 

Remember, two things: 1)we are interested in the distribution 2) Order of H,T does not matter.

## Quiz

- Calculate  $p(D|\theta)$  for the observation TTHHH and  $\theta = 0.3$ .
  - $(0.3)^3(0.7)^2 = 0.013$

### **Bayesian Update**



### Painful Denominator

• Recall, for discrete distributions:

$$P(\theta|D) = \frac{P(D|\theta) P(\theta)}{\sum_{i} P(D|\theta_i) P(\theta_i)}$$

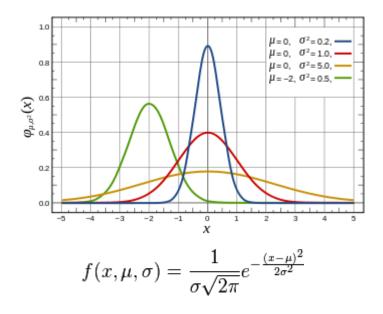
• And, for continuous distributions:

$$P(\theta|D) = \frac{P(D|\theta) P(\theta)}{\int P(D|\theta) P(\theta) d\theta}$$

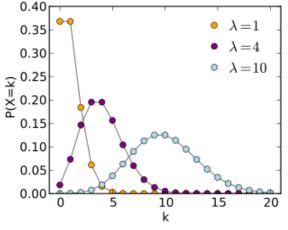
### A Simpler Way

#### Form, Functions and Distributions

Normal (or Gaussian)

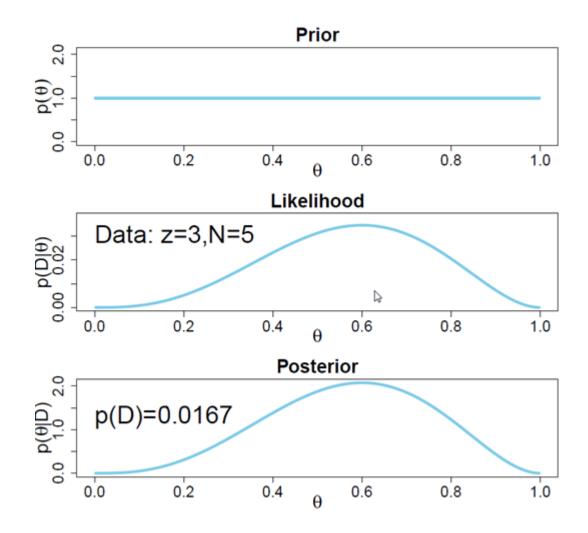


Poisson

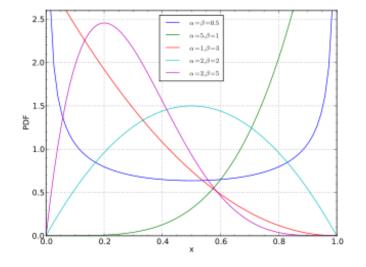


$$f(k; \lambda) = \Pr(X = k) = \frac{\lambda^k e^{-\lambda}}{k!},$$

### What form will suit us?



### Beta Distribution



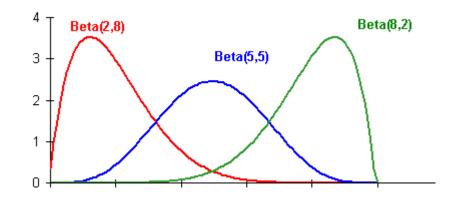
$$f(x;\alpha,\beta) = \frac{\Gamma(\alpha+\beta)}{\Gamma(\alpha)\Gamma(\beta)} x^{\alpha-1} (1-x)^{\beta-1}$$



Vadivelu, a famous Tamil comedian. This is one of his great expressions terrified and confused.

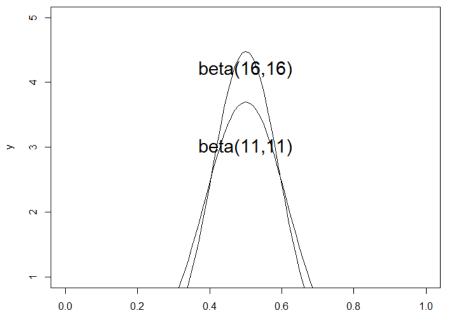
#### Beta Distribution

- Takes two parameters Beta(a,b)
- For now, assume *a* = #*H* + 1 and *b* = #*T* + 1.
- After 10 flips, we may end up in one of these three:



#### Prior and Posterior

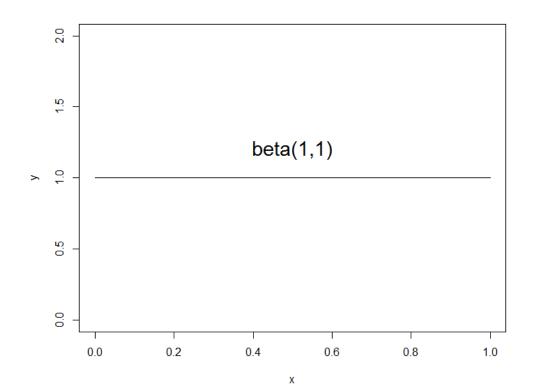
Let's say we have a Strong Prior – Beta (11,11). What should happen if we see 10 more observations with 5H and 5T?



х

#### Prior and Posterior...

What if we have not seen any data?



## Conjugate Prior

#### So, we see that:

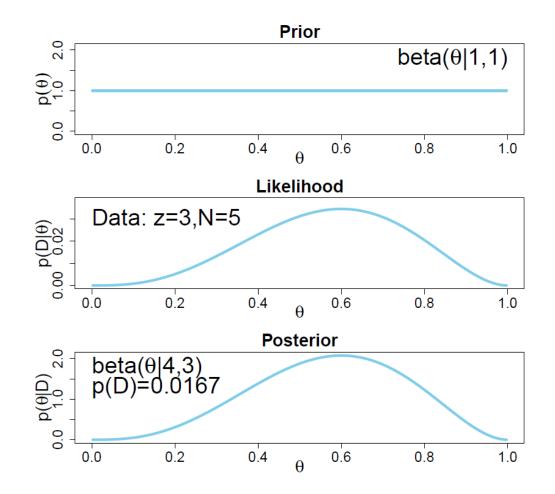


# Such Priors <u>that have same form</u> are called Conjugate Priors

# Summary

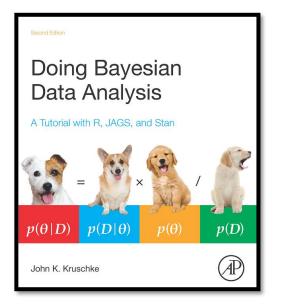
Prior
Likelihood
Posterior
Bayes' Rule
Bernoulli Distribution

Beta Distribution



#### References

<u>Book on Doing Bayesian</u> <u>Data Analysis</u> – John K. Kruschke.



<u>YouTube Video on</u> <u>Statistics 101: The</u> <u>Binomial Distribution</u> – Brandon Foltz.

#### BINQMIAL DISTRIBUTIONS: SALES PERFORMANCE

Brandon C. Foltz, M.Ed. education / training & development / business / tech / math / opinion http://www.botoltz.com/blog Twitter: @BCFoltz YouTube: BCFoltz

## Maximum Likelihood Estimation

- An Observation
  - HHTT TTHH TTTT TTTT
- What values of P(H) and P(T) will maximize the probability of the above observation?
- P(...Observation...) =  $x^4(1-x)^{12}$
- For what value of x will this function maximize?

## Probabilistic Retrieval

- Information Need: Taj Mahal
- Let a query q be "Taj"
- Let the results be:
  - d1: Taj
  - d2: Taj Mahal
  - d3: Taj Tea
- Two judges were asked to provide relevance judgments:

Document	Judge 1	Judge 2
Тај	R	Ν
Taj Mahal	R	R
Тај Теа	Ν	Ν

## Probability of Relevance

- Documents <u>can</u> have probability of being relevant and of being non-relevant at the same time.
- Example:
  - Documents in our collection :

Document	P(R=0 d,q)	P(R=1 d,q)
Тај	0.5	0.5
Taj Mahal	?	?
Тај Теа	?	?

 $R = 0 \rightarrow Non-Relevant$ 

 $R = 1 \rightarrow Relevant$ 

## Probability of Relevance

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- Example:
  - Documents in our collection :

Document	P(R=0 d,q)	P(R=1 d,q)
Тај	0.5	0.5
Taj Mahal	0	1
Тај Теа	1	0

 $R = 0 \rightarrow Non-Relevant$ 

 $R = 1 \rightarrow Relevant$ 

## Probability Ranking Principle

Rank documents by the probability of relevance, P(R=1|q,d) R∈{0,1}

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Document	P(R=0 d,q)	P(R=1 d,q)
Тај	0.5	0.5
Taj Mahal	0	1
Тај Теа	1	0

 $R = 0 \rightarrow \text{Non-Relevant}$ R = 1  $\rightarrow$  Relevant

#### Search Result:

- 1. Taj Mahal
- 2. Taj
- 3. Taj Tea

#### **Bayes Optimal Decision Rule**

#### d is relevant if P(R=1|d,q) > P(R=0|d,q)

Document	P(R=0 d,q)	P(R=1 d,q)
Тај	0.5	0.5
Taj Mahal	0	1
Тај Теа	1	0

Search Result: 1. Taj Mahal

# **Predicting Relevance**

Document	P(R=0 d,q)	P(R=1 d,q)
Тај	0.5	0.5
Taj Mahal	0	1
Тај Теа	1	0

This is user given relevance.

Can we estimate/predict relevance based on term occurrence ?

## Predicting Relevance

- You may use labeled set from judges (or mined from clicklogs)
- You may assume query and document as set of words.

Query	Document	Relevance	
q1 = (x1,x2,)	d1 = (xi, xj,)	1	This is user given relevance.
q1	d2	1	C relevance.
q1	d3	0	Can we
q2	d1	0	estimate/predict
q2	d2	0	relevance ?
q2	d3	1	

# Binary Independence Model (BIM)

- Each document is a binary vector of terms.
- Occurrence of terms is mutually independent.

$$P(R = 1|d,q) = \frac{P(d|R = 1,q)P(R = 1|q)}{P(d|q)}$$

**Bayes Rule** 

#### Quiz

$$P(R = 1|d,q) = \frac{P(d|R = 1,q)P(R = 1|q)}{P(d|q)}$$

**Bayes Rule** 

- P(d=Taj|R=1,q) = ?
- P(R=1|q) = ?
- P(d|q) = ?

Document	P(R=0 d,q)	P(R=1 d,q)
Тај	0.5	0.5
Taj Mahal	0	1
Тај Теа	1	0

## Quiz

$$P(R = 1|d,q) = \frac{P(d|R = 1,q)P(R = 1|q)}{P(d|q)}$$

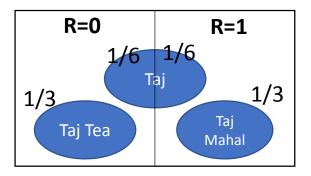
**Bayes Rule** 

• P(R=1|q) = 1/2

$$P(R=1|d=Taj,q) = (1/3)(1/2)/(1/3) = \frac{1}{2}$$

• P(d=Taj|q) = 1/3

Document	P(R=0 d,q)	P(R=1 d,q)
Тај	0.5	0.5
Taj Mahal	0	1
Тај Теа	1	0



#### **Predicting Relevance**

Odds of Relevance is easier to calculate

 $O(R|\vec{x},\vec{q}) = \frac{P(R=1|\vec{x},\vec{q})}{P(R=0|\vec{x},\vec{q})} = \frac{\frac{P(R=1|\vec{q})P(\vec{x}|R=1,\vec{q})}{P(\vec{x}|\vec{q})}}{\frac{P(R=0|\vec{q})P(\vec{x}|R=0,\vec{q})}{P(\vec{x}|\vec{q})}} = \frac{P(R=1|\vec{q})}{P(R=0|\vec{q})} \cdot \frac{P(\vec{x}|R=1,\vec{q})}{P(\vec{x}|R=0,\vec{q})}$ 

 In BIM, we assume that the term occurrence is mutually independent

$$\frac{P(\vec{x}|R=1,\vec{q})}{P(\vec{x}|R=0,\vec{q})} = \prod_{t=1}^{M} \frac{P(x_t|R=1,\vec{q})}{P(x_t|R=0,\vec{q})} = \prod_{t:x_t=1} \frac{P(x_t=1|R=1,\vec{q})}{P(x_t=1|R=0,\vec{q})} \cdot \prod_{t:x_t=0} \frac{P(x_t=0|R=1,\vec{q})}{P(x_t=0|R=0,\vec{q})}$$
$$= \prod_{t:x_t=q_t=1} \frac{p_t}{u_t} \cdot \prod_{t:x_t=0,q_t=1} \frac{1-p_t}{1-u_t}$$

#### Retrieval Status Value

Not document specific. Constant for a query.

$$\prod_{t:x_t=q_t=1} \frac{p_t(1-u_t)}{u_t(1-p_t)} \cdot \prod_{t:q_t=1} \frac{1-p_t}{1-u_t}$$

RSV = 
$$\prod_{t:x_t=q_t=1} \frac{p_t(1-u_t)}{u_t(1-p_t)}$$

**RSV** is used for ranking documents.

# Thank You