

Date: Thursday, March 7, 2019

## Prior for Normal Distribution

$$x_j \sim N(\mu, \sigma)$$

1) Do we want a prior on  $\sigma$ ?

(If no, give an estimate for this using standard deviation of data)

(If yes, see left)

Standard choices:

$$2) \mu \sim N(\mu_0, \sigma_0)$$

$\mu_0$  and  $\sigma_0$  are fixed.

Values can use empirical Bayes or grid search or choose a value w/ motivation

E.B:  $\mu_0 = \bar{x}$

$$\sigma_0 = \sqrt{\frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2}$$

Sample Standard Deviation

$$b) \sigma^2 \sim \chi_{n-1}^2$$

$$\sigma \sim U(0, \text{max val})$$

Gamma distribution Exponential

# Beta-Binomial Model

Data:  $x_j \sim \text{Binomial}(n, p)$

Fixed

↓  
Modeling  
Court

↳ often used for rate problems  
(click thru, conversion, etc)

$P \sim \text{Beta}(\alpha, \beta)$

$$E[P] = \frac{\alpha}{\alpha + \beta}, \quad \text{var}(P) = \frac{\alpha\beta}{(\alpha + \beta)^2(\alpha + \beta + 1)}$$

EB approach: \* first estimate  $\frac{\bar{x}}{n}$  and  $\frac{\hat{\sigma}^2}{n^2}$   
from data  $(x_1, \dots, x_n)$

\* solve above by plugging in  
 $E[P] = \frac{\bar{x}}{n}, \quad \text{var}(P) = \frac{\hat{\sigma}^2}{n^2}$

The beta distribution is used to model data that lies between 0 and 1. Its pdf can have many shapes, depends on  $\alpha, \beta$ . when  $\alpha = \beta = 1$ , the  $\text{Beta}(\alpha, \beta)$  is a  $\text{Unif}(0, 1)$ .

## Rate Questions: Bernoulli vs Binomial

- In the context of individuals clicking on an ad
- Use Bernoulli( $p$ ) if the individual only sees this ad once (one interaction).
- Use Binomial( $n, p$ ) if there are  $n$  interactions w/ the ad.  
(Binomial always works because it's a generalization of Bernoulli).

## Binomial vs Poisson

- Both for counts
- $\text{Bin}(n, p)$  has  $n$  as maximum
- $\text{Po}(\lambda)$  has no known maximum

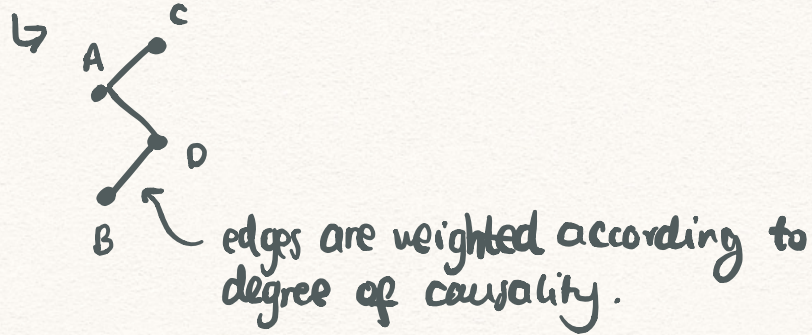
Mean counts:  $E[\text{Bin}(np)] = np$

$$E[\text{Po}(\lambda)] = \lambda$$

# Major uses/applications of Bayesian Modeling Today

- Bayesian hypothesis testing
  - ↳ Results give probability of each hypothesis being true given the data.
  - ↳ More easily interpreted than p-values
- Classification:
  - ↳ Bayesian classifier
  - ↳ Naive Bayes
  - ↳ Discriminant analysis
- Topic analysis
  - ↳ Latent Dirichlet Allocation
    - Aim: Identify topics for a corpora of documents and calculate probability for each topic.
- Bayesian Networks (AKA Directed Acyclic Graphs DAGs)
  - Idea: We have  $n$  data objects (ectors, documents, entities, etc) where some objects are thought to **cause** or **heavily influence** others.
  - Model: Probabilistic causal model where we construct a graph so that edges from node  $A \rightarrow B$  implies  $A$  caused  $B$ .

Example: Data:  $\{A, B, C, D\}$



In DAGs, there are two main objectives

- 1) From multivariate data estimate the DAG/causal graph. This relies upon techniques like Gaussian graphical models, etc.
- 2) Once we have DAG, calculate joint probabilities using conditioning techniques based on edges in the graph.

