

CSC-591: Foundations of Data Science T/Th. 12:50-2:05pm. EBI-1005.

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W16: 12/03/15

Today

- Final Review
- Final: 12/8/15, 1:00-4:00pm. (In Class)

Grading

Grading Item	Number of points	X %	Grading Score (out of 100)	Bonus Your Score x %
HW1	100	5	5	
HW2	100	5	5	
HW3	100	5	5	X 2%
HW4	100	5	5	X 2%
HW5	100	5	5	X 4%
MT1	100	15	15	
MT2	100	20	20	X 4%
Final	100	35	35	X 2%
Instructor			5 (class participation + review)	
Total			100	14

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Topics

- Probability distributions and estimation
- Sampling distribution and CLT
- Hypothesis testing
- Regression
- Attribute selection
- Sampling and cross validation
- Bayesian Inference
- Bayesian networks
- Missing data

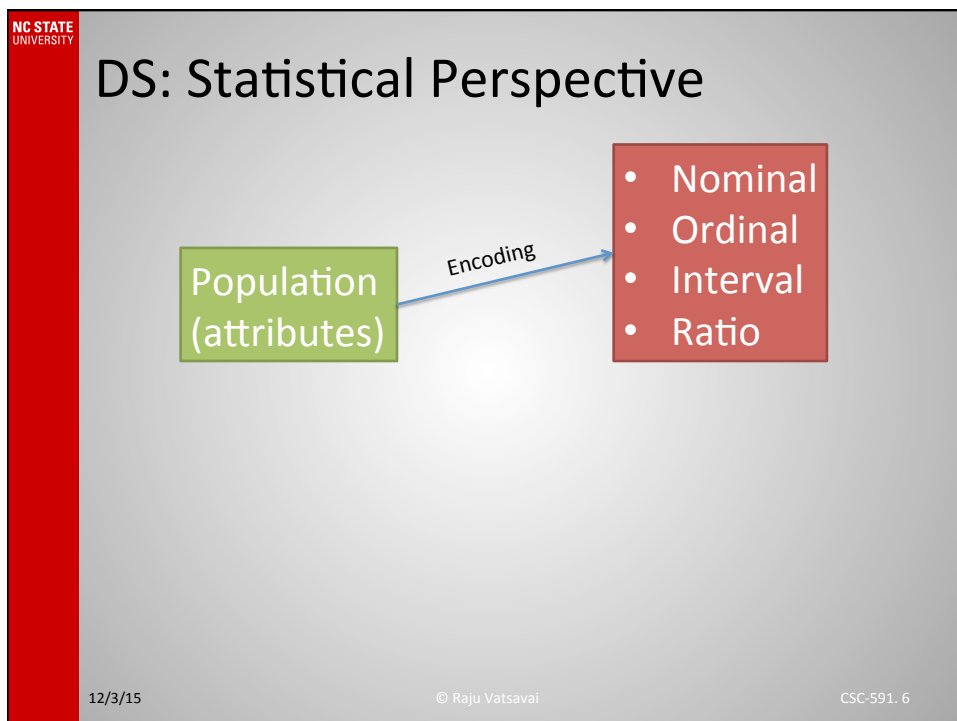
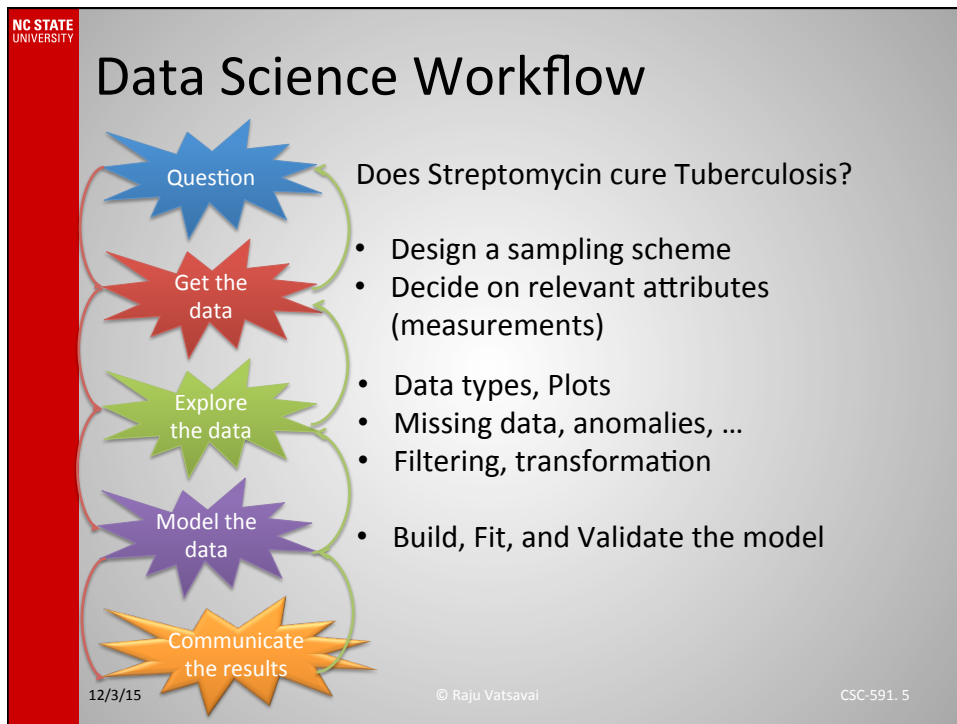
~60%

~40%

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DS: Statistical Perspective				
	Attribute Type	Description	Examples	Operations
Categorical Qualitative	Nominal	Nominal attribute values only distinguish. ($=$, \neq)	zip codes, employee ID numbers, eye color, sex: <i>{male, female}</i>	mode, entropy, contingency correlation, χ^2 test
	Ordinal	Ordinal attribute values also order objects. ($<$, $>$)	hardness of minerals, <i>{good, better, best}</i> , grades, street numbers	median, percentiles, rank correlation, run tests, sign tests
Numeric Quantitative	Interval	For interval attributes, differences between values are meaningful. ($+$, $-$)	calendar dates, temperature in Celsius or Fahrenheit	mean, standard deviation, Pearson's correlation, t and F tests
	Ratio	For ratio variables, both differences and ratios are meaningful. ($*$, $/$)	temperature in Kelvin, monetary quantities, counts, age, mass, length, current	geometric mean, harmonic mean, percent variation

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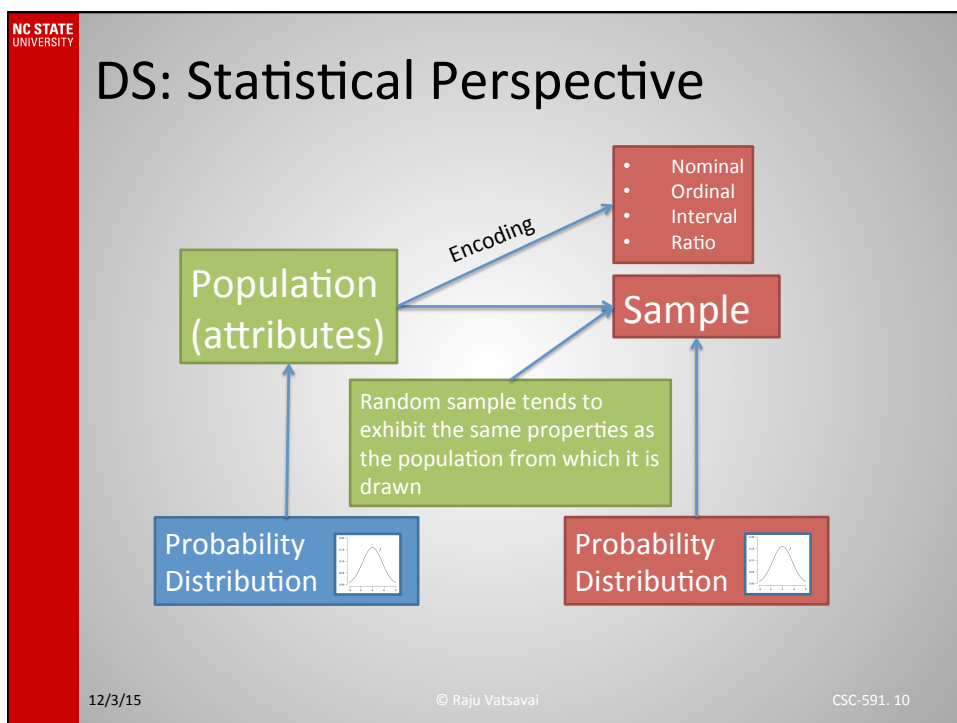
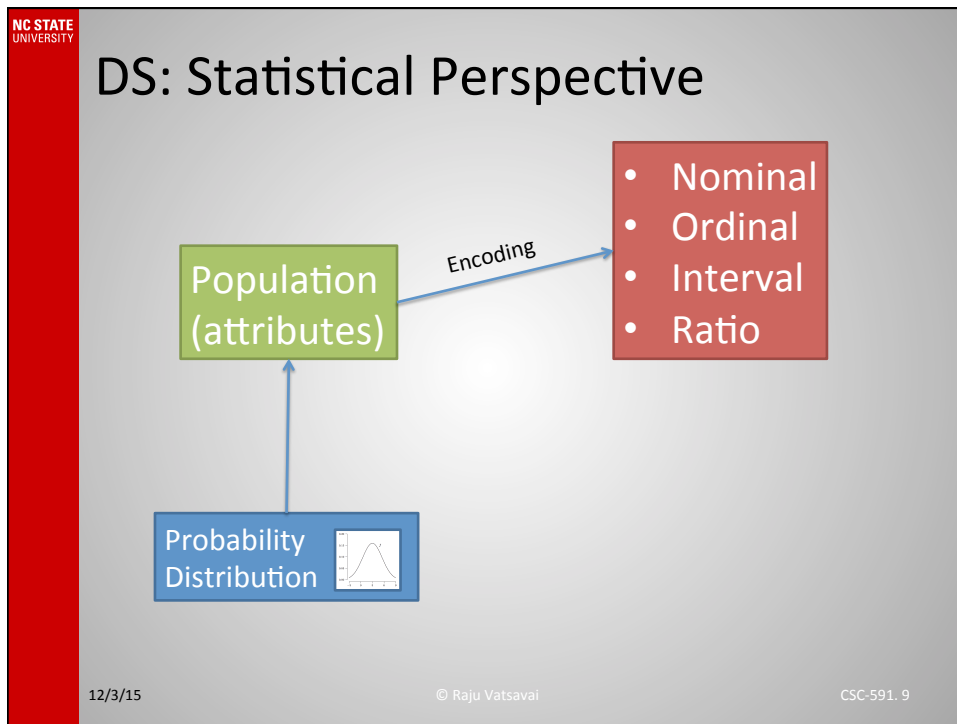
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DS: Statistical Perspective	
<p>NC STATE UNIVERSITY</p>	<ul style="list-style-type: none"> • Experiment, outcomes, sample space, events • Basic set operations • Probability and three axioms • Probability rules • Independent events • Conditional probability • Bayes theorem • Random variables: discrete and continuous • PMF and Bernoulli distribution

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DS: Statistical Perspective

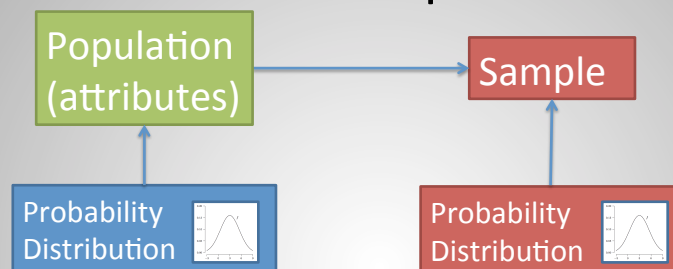
- Experiment, outcomes, sample space, events
- Basic set operations
- Probability and three axioms
- Probability rules
- Independent events
- Conditional probability
- Bayes theorem
- Random variables: discrete and continuous
- PMF and Bernoulli distribution
- Continuous uniform distribution, PDF, Normal distribution

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DS: Statistical Perspective



- Draw random samples from a population
- Make inference (conclusions) about populations from noisy data that is drawn from it
- The randomness governing data (samples) is given by densities and mass functions

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DS: Statistical Perspective

```

graph LR
    A[IID Random Samples  
(random variables)] --> B[Probability  
Distribution]
    B --> C[MLE]
  
```

The flowchart illustrates the process of Maximum Likelihood Estimation (MLE). It starts with 'IID Random Samples (random variables)', which leads to a 'Probability Distribution' (represented by a normal distribution curve). This distribution then leads to the 'MLE' (Maximum Likelihood Estimation) step.

- How good are these estimates?

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DS: Statistical Perspective

- How good are these estimates?

```

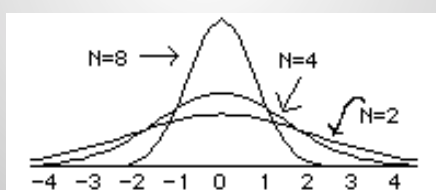
graph LR
    P((Population)) --> S1((sample))
    P --> S2((sample))
    P --> S3((sample))
    P --> S4((sample))
    P --> S5((sample))
    S1 --> SS1[Sample Statistic (e.g., Mean =  $\bar{x}$ )]
    S2 --> SS2[Sample Statistic (e.g., Mean =  $\bar{x}$ )]
    S3 --> SS3[Sample Statistic (e.g., Mean =  $\bar{x}$ )]
    S4 --> SS4[Sample Statistic (e.g., Mean =  $\bar{x}$ )]
    S5 --> SS5[Sample Statistic (e.g., Mean =  $\bar{x}$ )]
  
```

The diagram illustrates the process of sampling from a population. A central blue circle labeled 'Population' has five arrows pointing to five green circles, each labeled 'sample'. Each 'sample' circle has an arrow pointing to a corresponding 'Sample Statistic (e.g., Mean = \bar{x})' text. Below the 'sample' circles, the text 'Sample Distributions' is centered, and below the 'Sample Statistic' text, the text 'Sampling Distribution' is centered.

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DS: Statistical Perspective

- Sampling distribution is used in constructing **CI** for mean and for **significance testing**.
- Given a population with a mean of μ and a standard deviation of σ , the sampling distribution of the mean has a mean of μ and a variance of $\sigma_m^2 = \frac{\sigma^2}{N}$



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DS: Statistical Perspective

- Given a population with a finite mean μ and a finite non-zero variance σ^2 , the sampling distribution of the mean approaches a normal distribution with a mean of μ and a variance of σ^2/N as N , the sample size, increases.
 $\bar{x} \sim N(\text{mean} = \mu, \text{and SE} = s/\sqrt{n})$
- What is the significance?

Regardless of the shape of the parent population, the sampling distribution of the mean approaches a normal distribution as N increases.

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DS: Statistical Perspective

Does Streptomycin cure Tuberculosis?

```

graph LR
    Sample[Sample] --> Control[Control Group  
(standard treatment or placebo)]
    Sample --> Experimental[Experimental Group  
(new treatment or drug)]
    Control --> Comparison((Comparison  
(Hypothesis Testing)))
    Experimental --> Comparison
  
```

- Hypothesis
 - Is a statement about a population parameter
- Hypothesis testing
 - Based on a sample from population, goal is to decide on which of the two complementary hypothesis is true
- Null and alternative hypothesis
 - Two complementary hypothesis (null and alternative) – both needs to be specified

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DS: Statistical Perspective

Does Streptomycin cure Tuberculosis?

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graph LR
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    Control --> Comparison((Comparison  
(Hypothesis Testing)))
    Experimental --> Comparison
  
```

- General Procedure

Step 1	State the hypotheses and identify the claim.
Step 2	Find the critical value(s) from the appropriate table
Step 3	Compute the test value.
Step 4	Make the decision to reject or not reject the null hypothesis.
Step 5	Summarize the results.

```

graph TD
    Pop[Population attributes] -- Encoding --> Enc[Encoding: Nominal, Ordinal, Interval, Ratio]
    Pop --> Sample[Sample]
    Sample --> CG[CG]
    Sample --> EG[EG]
    CG --> Comp((Comparison Hypothesis Testing))
    EG --> Comp
    Note[Random sample tends to exhibit the same properties as the population from which it is drawn]
    ProbDist1[Probability Distribution] --> Pop
    ProbDist2[Probability Distribution] --> MLE[MLE]
    SDCLT[Sampling distribution and CLT] --> MLE
  
```

Population (attributes)

Encoding

- Nominal
- Ordinal
- Interval
- Ratio

Sample

CG

EG

Comparison (Hypothesis Testing)

Random sample tends to exhibit the same properties as the population from which it is drawn

Probability Distribution

Probability Distribution

Sampling distribution and CLT

MLE

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DS: Statistical Perspective

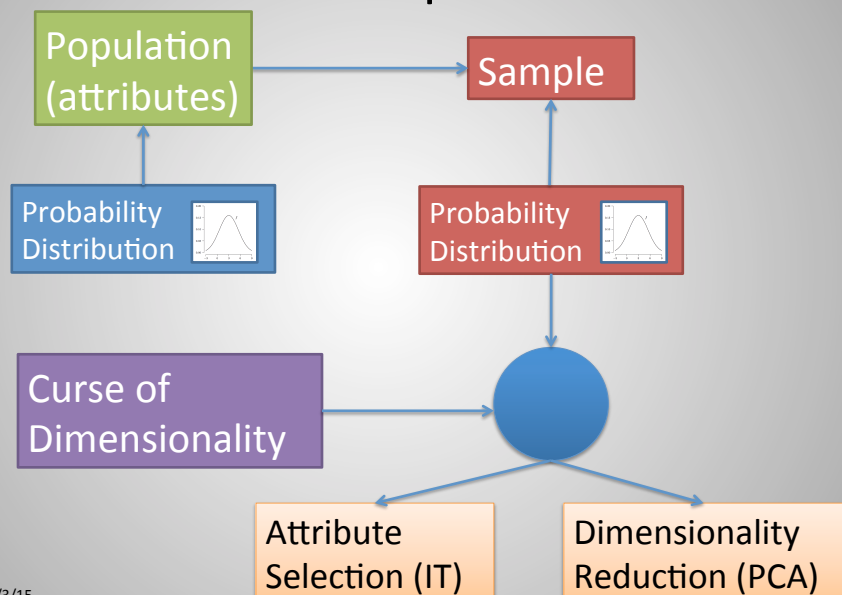
- Linear Regression
 - Least Squares
- Correlation
- Regression Parameters
- Properties
- Significance of “r”
- Total Variation (explained + unexplained)
- Coefficient of Determination
- Standard Error of estimate, Prediction Interval
- Multiple Linear Regression
- Multiple Correlation Coefficient (R)
- Testing for significance of R
- Regression as classification: Logistic Regression

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DS: Statistical Perspective



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DS: Statistical Perspective

```

graph TD
    PD1[Probability Distribution] --> Pop[Population (attributes)]
    Pop --> Sample[Sample]
    Sample --> PD2[Probability Distribution]
    PD2 --> MLE[MLE]
    PD2 --> Bayesian[Bayesian]
    Sample --> Bayesian
  
```

- Bayesian analysis is a statistical procedure which endeavors to estimate parameters of an underlying distribution based on the observed distribution

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- Posterior distribution is the most important quantity in Bayesian inference.

$$f(\theta | x) = \frac{f(x | \theta) f(\theta)}{\int f(x | \theta) f(\theta) d\theta}$$

- Let $X=x$ denote the observed realization of a uni- or multivariate r.v. X with density function $f(x | \theta)$. Specifying a prior distribution $f(\theta)$ allows us to compute the density function $f(\theta | x)$ of the posterior distribution using Bayes' theorem.

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- Bayesian inference allows the probabilistic specification of prior beliefs through a prior distribution.
- It is often useful and justified to restrict the **range of possible prior distributions** to a specific family with one or two parameters, say. The choice of this family can be based on the type of **likelihood function** encountered.

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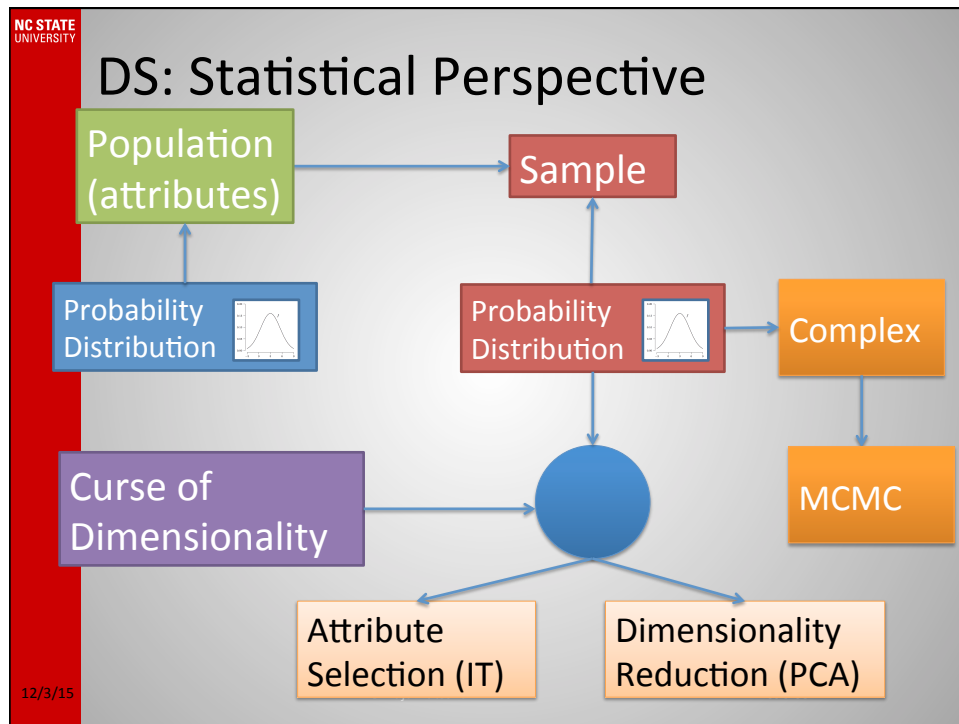
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- A pragmatic approach to choosing a prior distribution is to select a member of a specific family of distributions such that the posterior distribution belongs to the same family. This is called a **conjugate prior distribution**.

Summary of conjugate prior distributions for different likelihood functions

Likelihood	Conjugate prior distribution	Posterior distribution
$X \pi \sim \text{Bin}(n, \pi)$	$\pi \sim \text{Be}(\alpha, \beta)$	$\pi x \sim \text{Be}(\alpha + x, \beta + n - x)$
$X \pi \sim \text{Geom}(\pi)$	$\pi \sim \text{Be}(\alpha, \beta)$	$\pi x \sim \text{Be}(\alpha + 1, \beta + x - 1)$
$X \lambda \sim \text{Po}(e \cdot \lambda)$	$\lambda \sim \text{G}(\alpha, \beta)$	$\lambda x \sim \text{G}(\alpha + x, \beta + e)$
$X \lambda \sim \text{Exp}(\lambda)$	$\lambda \sim \text{G}(\alpha, \beta)$	$\lambda x \sim \text{G}(\alpha + 1, \beta + x)$
$X \mu \sim \text{N}(\mu, \sigma^2 \text{ known})$	$\mu \sim \text{N}(\nu, \tau^2)$	$\mu x \sim \text{N}\left(\left(\frac{1}{\sigma^2} + \frac{1}{\tau^2}\right)^{-1} \cdot \left(\frac{x}{\sigma^2} + \frac{\nu}{\tau^2}\right), \left(\frac{1}{\sigma^2} + \frac{1}{\tau^2}\right)^{-1}\right)$
$X \sigma^2 \sim \text{N}(\mu \text{ known}, \sigma^2)$	$\sigma^2 \sim \text{IG}(\alpha, \beta)$	$\sigma^2 x \sim \text{IG}\left(\alpha + \frac{1}{2}, \beta + \frac{1}{2}(x - \mu)^2\right)$



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DS: Statistical Perspective

- Model Selection
- Mixture Models
- Expectation Maximization
- Missing Data

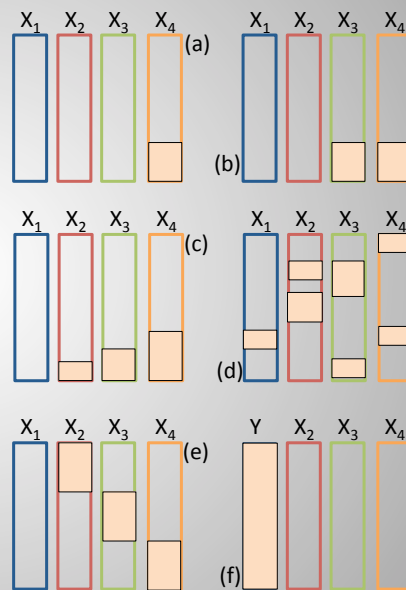
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Missing Data Patterns

- (a) Univariate pattern
- (b) Unit nonresponse pattern
- (c) Monotone pattern
- (d) General pattern
- (e) Planned missing pattern
- (f) Latent variable pattern



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Missing Data Theory

- Rubin, et. al. introduced a classification system for missing data problems
 - Introduced three so-called **missing data mechanisms** that describe how the probability of a missing value relates to the data, if at all.
 - **MAR**: data are missing at random
 - **MCAR**: missing completely at random
 - **MNAR**: missing not at random

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For Final, focus on these topics

- Probability distributions and estimation
- Sampling distribution and CLT
- Hypothesis testing ~60%
- Regression
- Attribute selection
- Sampling and cross validation
- Bayesian Inference
- Bayesian networks ~40%
- Missing data

Exam Structure

- One big with several short questions – to test fundamental understanding (~20%)
- 4-5 Numerical questions (~60%)
- 1-2 Tricky question (~20%)
 - (not to trick you; but solution could be more easy if you think about it little bit). May be numerical, but you can obtain solution without going through routine computations
- Bonus question (2%)
 - Little bit hard
- Roughly 2.5 Hours

Acknowledgements

- Thank you
- Please complete review to get 1% grade point