## ALY 6050[21598]

## Introduction to Enterprise Analytics



M1: Review of Probability Distributions

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#### Prerequisites

• Excel

• R programming experience

• Completion of ALY6010 & ALY6015

#### Administrative Notes

- Academic Integrity
  - -Cheating
  - Fabrication
  - -Plagiarism
  - Unauthorized Collaboration

**—** ...



#### Administrative Notes

- Discussion:
  - -One primary response(250 words) by **Thursday**
  - -Two secondary responses (100 words) by **Sunday**
  - -Last discussion is due by Friday of week 6
- Weekly Projects: due on Sunday
  - -Last project is due by Friday of week 6
- TA: Sashank Yakkali yakkali.s@northeastern.edu



#### Northeastern University

#### Grade Breakdown

Assignment	Grade	Weight in Course Grade
6 Weekly Discussions	240 points (40 points each)	24%
5 Weekly Projects	500 points (100 points each)	50%
Final Assessment	100 points	10%
Week 6 Final Project	160 points	16%
Total:	1000 points	100%

#### Topics Covered in this Course

- Review of Probability Distributions
- Simulation
- Forecasting & Regression
- Decision Modeling
- Optimization I
- Optimization II



#### Learning Objectives

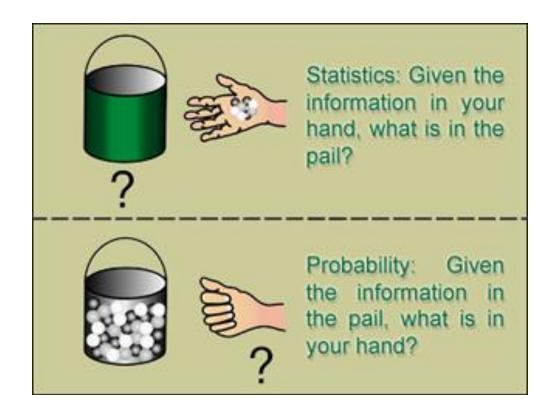
- Discrete distributions:
  - Bernoulli, binomial, hypergeometric, and
     Poisson
- Continuous distributions:
  - -uniform, triangular, normal, exponential, beta, gamma, log-normal, Weibull
- Goodness-of-fit test



## Lecture Part I Recap



#### Probability vs. Statistics



#### Descriptive Statistics

- Measures of central tendency
  - -mean, median, mode
- Measures of dispersion or variation
  - -range, variance, standard deviation
- Measures of position
  - -percentile, quartile
- Measures of frequency
  - -count, percent



#### Random Variable(RV)

- A variable is an attribute of an object of study
  - Categorical
  - -Numerical
- A random variable is associated with uncertainty







#### Discrete vs. Continuous Variable

- Discrete
  - Only certain values
  - -Countable & Finite





- -Number of people in a race, coin toss
- Continuous
  - —Any value on an interval
  - -Measurable & Infinite





-Time to run a race, amount of snow in winter



#### Expectation & Variance

$$E[X] = \mu_{x} = \sum_{i=1}^{n} x_{i} p_{i} = \bar{X}$$

$$Var[X] = \sigma_x^2 = \sum_{i=1}^{n} (x_i - \mu_x)^2 p_i$$



#### Variance

$$Var[X] = \sigma_x^2 = \sum_{i=1}^{\infty} (x_i - \mu_x)^2 p_i$$

$$= \sum_{i}^{n} x_i^2 p_i - 2 \mu_x x_i p_i + \mu_x^2 p_i$$

$$= E[X^2] - 2\mu_x \sum_{i=1}^{\infty} x_i p_i + \mu_x^2$$

$$= E[X^2] - \mu_x^2$$



#### Covariance

$$Cov[X,Y] \stackrel{\text{def}}{=} E[(X - \overline{X}) (Y - \overline{Y})]$$

$$= E[XY] - E[X]E[Y]$$

$$= E[XY] - \mu_x \mu_y$$

$$Cov[X,X] = Var(X)$$

#### **Correlation**

$$\begin{aligned}
& \boldsymbol{Cov}[\boldsymbol{U}, \boldsymbol{V}] = \boldsymbol{Cov} \left[ \frac{\boldsymbol{X} - \boldsymbol{\mu}_{x}}{\sigma_{x}}, \frac{\boldsymbol{Y} - \boldsymbol{\mu}_{y}}{\sigma_{y}} \right] \\
&= E \left[ \frac{\boldsymbol{X} - \boldsymbol{\mu}_{x}}{\sigma_{x}} \frac{\boldsymbol{Y} - \boldsymbol{\mu}_{y}}{\sigma_{y}} \right] - E \left[ \frac{\boldsymbol{X} - \boldsymbol{\mu}_{x}}{\sigma_{x}} \right] E \left[ \frac{\boldsymbol{Y} - \boldsymbol{\mu}_{y}}{\sigma_{y}} \right] \\
&= \frac{E[\boldsymbol{XY}] - \boldsymbol{\mu}_{x} \boldsymbol{\mu}_{y}}{\sigma_{x} \sigma_{y}} - \frac{E[\boldsymbol{X}] E[\boldsymbol{Y}] - \boldsymbol{\mu}_{x} \boldsymbol{\mu}_{y}}{\sigma_{x} \sigma_{y}} \\
&= \frac{E[\boldsymbol{XY}] - E[\boldsymbol{X}] E[\boldsymbol{Y}]}{\sigma_{x} \sigma_{y}} = \frac{\boldsymbol{Cov}[\boldsymbol{X}, \boldsymbol{Y}]}{\sigma_{x} \sigma_{y}} = \boldsymbol{\rho}_{xy}
\end{aligned}$$



#### Markov's Inequality

$$P(X \ge \varepsilon) \le \frac{E[X]}{\varepsilon}$$

$$E[X] = \int_0^\infty x f(x) dx \ge \int_{\varepsilon}^\infty x f(x) dx$$

$$\geq \int_{\varepsilon}^{\infty} \varepsilon f(x) dx = \varepsilon \int_{\varepsilon}^{\infty} f(x) dx$$

$$= \varepsilon P(X \ge \varepsilon)$$



#### Chebyshev's Inequality

$$P(|X - \mu| \ge k\sigma) \le \frac{1}{k^2}$$

$$P(|X - \mu| \ge k\sigma) = P((X - \mu)^2 \ge k^2 \sigma^2)$$

$$\leq \frac{E[(X-\mu)^2]}{k^2\sigma^2} = \frac{\sigma^2}{k^2\sigma^2} = \frac{1}{k^2}$$



## Chebyshev's Inequality II

$$P(|X - \mu| \ge \varepsilon) \le \frac{\sigma^2}{\varepsilon^2}$$

$$P(|X - \mu| \ge \varepsilon) = P((X - \mu)^2 \ge \varepsilon^2)$$

$$\leq \frac{E[(X-\mu)^2]}{\varepsilon^2} = \frac{\sigma^2}{\varepsilon^2}$$

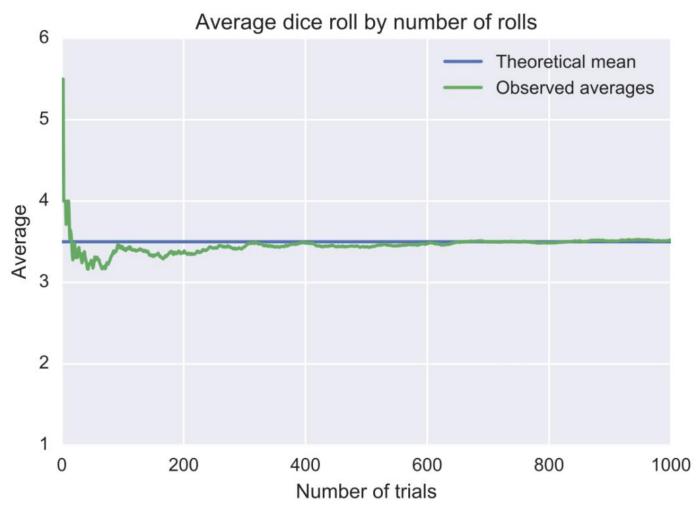


#### Law of Large Numbers (LLN)

 Sample mean converges to population mean as sample size n → infinity

• No guarantee when sample size is small

#### Illustration for LLN



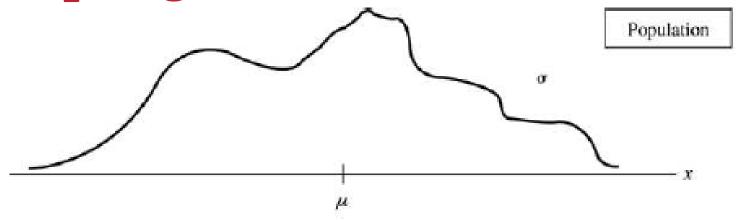
#### Central Limit Theorem (CLT)

- Regardless the underly population
- The distribution of sample means is approximately **normal** 
  - -mean:  $\mu_{\bar{x}} = \mu$
  - -standard deviation:  $\sigma_{\bar{\chi}} = \frac{\sigma}{\sqrt{n}}$
- Confidence Interval  $100(1 \alpha)\%$ :

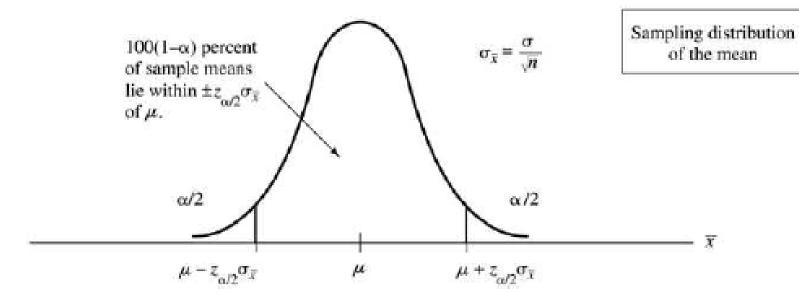
$$\bar{x} \pm z_{\frac{\alpha}{2}} \cdot \frac{\sigma}{\sqrt{n}}$$
 > qnorm(0.025, lower.tail = FALSE) 
> qnorm(0.05, lower.tail = FALSE) 
> qnorm(0.05, lower.tail = FALSE) 
[1] 1.644854



#### Sampling Distribution of the Mean



$$z_{0.025} = 1.96$$
 and  $z_{0.05} = 1.645$ .





## **Probability Distributions**

Distribution	Abbreviation	Distribution	Abbrevlation
Beta	beta	Logistic	logis
Binomial	binom	Multinomial	multinom
Cauchy	cauchy	Negative binomial	nbinom
Chi-squared (noncentral)	chisq	Normal	norm
Exponential	exp	Poisson	pois
F	f	Wilcoxon signed rank	signrank
Gamma	gamma	Т	t
Geometric	geom	Uniform	unif
Hypergeometric	hyper	Weibull	weibull
Lognormal	lnorm	Wilcoxon rank sum	wilcox

#### R Probability Functions

- [d]ensity
- [p]robability
- [q]uantile
- [r]andom

## R Probability Functions - Normal

- **d**norm()
- **p**norm()
- **q**norm()
- **r**norm()

#### Bayes' Theorem

H: Hypothesis

D: Data or Event

#### Likelihood

How probable is the event given the hypothesis is true

#### Prior

How probable was the hypothesis before observing any event?

$$P(H|D) = \frac{P(D|H) * P(H)}{P(D)}$$

#### **Posterior**

How probable is the hypothesis given the observed event?

#### Marginal

How probable is the new event under all possible hypothesis?

$$P(D) = \sum P(D|H_i)P(H_i)$$



## Food for Thought

• Q1. A special party invites family with at least one son. Bob's family has two children and gets invited. What's the probability that both children are boys?

• Q2. Bob's family has two children. If you see one of his children is boy, what's the probability that both children are boys?

# Lecture Part II Discrete Probability Distributions



#### Discrete Probability Distributions

- Bernoulli Distribution
- Binomial Distribution
- Multinomial Distribution
- Hypergeometric Distribution
- Geometric Distribution
- Poisson Distribution



#### Bernoulli Distribution

- Two possible outcomes (mutually exclusive)
  - -p: success
  - -q: failure (q = 1 p)
- One trial

$$X \sim Bernoulli(p), 0 \le p \le 1$$

$$p(x) = p^{x}(1-p)^{1-x}, x = 0 \text{ or } 1$$



#### Ex for Bernoulli

- 1000 tickets are sold at \$1 each
- Winner gets \$750
- What's the expected value of the gain if you purchase one ticket?

•	Win	Lose
Gain X	\$749	-\$1
Probability P(X)	$\frac{1}{1000}$	999 1000

$$E(X) = \$749 * \frac{1}{1000} - \$1 * \frac{999}{1000} = -\$0.25$$

#### **Binomial Distribution**

- Two possible outcomes (mutually exclusive)
- n trials
  - Independent
  - -Identical

$$X \sim Binomial(n, p), 0 \le p \le 1$$

$$p(x) = C_n^x p^x (1-p)^{n-x}$$



#### Ex for Binomial

- One out of five Americans has visited a doctor in any given month
- Randomly selected 10 people
- Find the probability that **exact 3** people have visited a doctor last month.

$$p = \frac{1}{5}, q = \frac{4}{5}, n = 10, x = 3$$



#### Ex for Binomial

$$p(x) = C_n^x p^x (1-p)^{n-x}$$

$$p(3) = C_{10}^3 \left(\frac{1}{5}\right)^3 \left(\frac{4}{5}\right)^7 \approx 0.201$$

```
> dbinom(3,10,0.2)
[1] 0.2013266
```



#### Multinomial Distribution

- More than two possible outcomes
- n trials
  - Independent
  - -Identical

 $X \sim Multinomial(n, p)$ 

$$p(x) = \frac{n!}{x_1! \cdot x_2! \cdots x_k!} p_1^{x_1} p_2^{x_2} \cdots p_k^{x_k}$$

where 
$$x_1 + x_2 + \dots + x_k = n$$
  
and  $p_1 + p_2 + \dots + p_k = 1$ 



#### Ex for Multinomial

- 65% use herbicides for commercial purposes
- 27% for agricultural purposes
- 8% for home and garden purposes
- Find the probability that 3 used them for commercial purposes, 1 for agriculture, and 1 for home or garden purposes.

$$n = 5$$
, where  $x_1 = 3$ ,  $x_2 = 1$ ,  $x_3 = 1$   
 $p_1 = 0.65$ ,  $p_2 = 0.27$ ,  $p_3 = 0.08$ 

#### Ex for Multinomial

$$p(x) = \frac{n!}{x_1! \cdot x_2! \cdots x_k!} p_1^{x_1} p_2^{x_2} \cdots p_k^{x_k}$$
$$p(x) = \frac{5!}{3!1!1!} (0.65)^3 (0.27)^1 (0.08)^1$$
$$\approx 0.119$$

```
> dmultinom(x=c(3,1,1), size=5, prob=c(0.65,0.27,0.08))
[1] 0.118638
```



# Hypergeometric Distribution

- Similar to binomial
- Sampling without replacement
- Applications: quality control

$$X \sim Hyper(m, n, k)$$

$$p(x) = \frac{C_m^x C_n^{k-x}}{C_{m+n}^k}$$



# Ex for Hypergeometric

- A lot of 12 compressor tanks with 3 defectives
- Three tanks are checked for leaks.
- If 1 or more of the 3 is defective  $\rightarrow$  rejected
- Find the probability rejecting the lot
- m = 3 defective tanks
- n = 9 good tanks
- k = 3 number of tank checked
- x = 0 no defective in the 3 tanks checked



# Ex for Hypergeometric

$$p(0) = \frac{C_m^x C_n^{k-x}}{C_{m+n}^k} = \frac{C_3^0 C_9^3}{C_{3+9}^3} \approx 0.382$$

$$p(rejection) = 1 - p(0) = 0.618$$

```
> dhyper(x=0, m=3, n=9, k=3)
[1] 0.3818182
> 1-dhyper(x=0, m=3, n=9, k=3)
[1] 0.6181818
```



#### Geometric Distribution

- Two possible outcomes (mutually exclusive)
- n trials until a success has been met
  - Independent
  - -Identical

$$X \sim Geometric(p), 0 \le p \le 1$$

$$p(n) = p(1-p)^{n-1}$$



#### Ex for Geometric



• Find the probability of getting the first 2 on the 3<sup>rd</sup> roll of a die.

$$p(not \ 2 \ \& \ not \ 2 \ \& \ 2) = \frac{5}{6} \frac{5}{6} \frac{1}{6} = \frac{25}{216}$$

$$p(3) = p(1-p)^{n-1} = \frac{1}{6} \left(\frac{5}{6}\right)^{3-1} = \frac{25}{216}$$

> dgeom(x=2, prob=1/6)
[1] 0.1157407

> 25/216 Γ1] 0.1157407



#### Poisson Distribution

- Modeling the frequency of rare events.
- The occurrences are random and independent
- The average number of occurrences over an interval is known

$$X \sim Poisson(\lambda), \quad \lambda > 0$$

$$p(x) = e^{-\lambda} \frac{\lambda^x}{x!}, \qquad x = 0,1, \dots$$



# Expectation of Poisson Distribution

$$E[X] = \lambda$$

$$p(x) = e^{-\lambda} \frac{\lambda^x}{x!}, \qquad x = 0,1,...$$

$$x = 0,1,...$$

$$E[X] = \sum_{i=1}^{n} x_i p_i$$

#### Ex for Poisson

- 200 typos in a 400-page manuscript.
- Find the probability of a given page having exactly zero errors.

$$\lambda = \frac{200}{400} = 0.5$$

$$p(0) = e^{-\lambda} \frac{\lambda^x}{x!} = 2.7183^{-0.5} = 0.61$$



# Lecture Part III Continuous Probability Distributions



#### Continuous Distributions

- Normal Distribution
- Log-normal Distribution
- Exponential Distribution
- Beta Distribution
- Gamma Distribution
- Weibull distribution

#### Normal Distribution

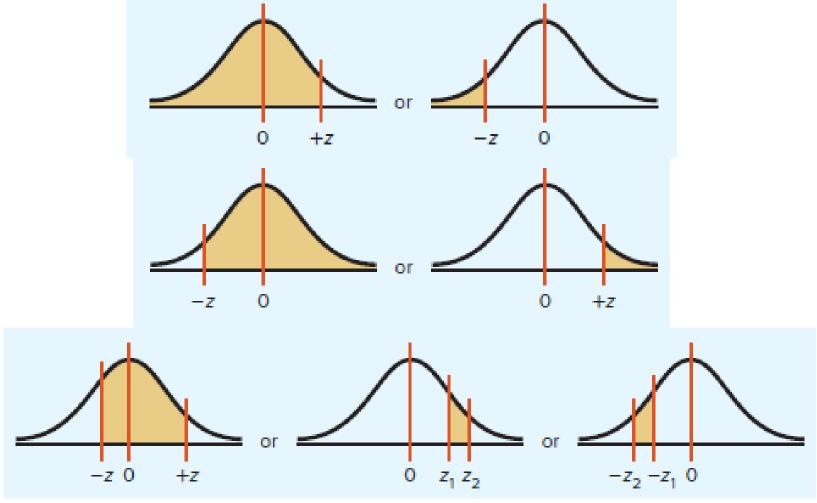
- Gaussian distribution or Bell-shaped curve
- The mean, median, and mode are **equal** and at center
- The area under the curve is 1
- The curve never touches the x-axis

$$X \sim Normal(\mu, \sigma^2)$$

$$f(x) = \frac{1}{\sqrt{2\pi}\sigma} \exp\left[-\frac{(x-\mu)^2}{2\sigma^2}\right]$$

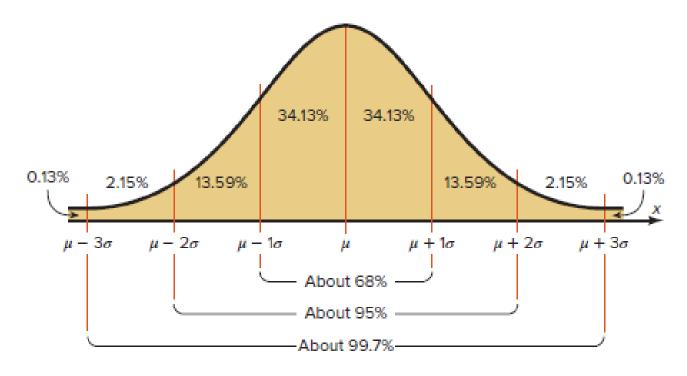


# Calculate the Probability





#### Standard Normal Distribution



```
> pnorm(1)-pnorm(-1)
[1] 0.6826895
> pnorm(2)-pnorm(-2)
[1] 0.9544997
> pnorm(3)-pnorm(-3)
[1] 0.9973002
```

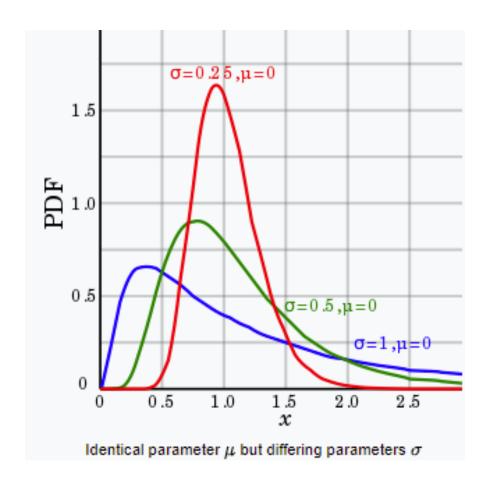


#### Lognormal Distribution

- A variable X is lognormally distributed if Y = ln(X) is normally distributed
- Applications:
  - time to complete a task

$$X \sim Lognormal(\mu, \sigma^2)$$
  
 $ln(X) \sim N(\mu, \sigma^2)$ 

# Visualizing Log-normal



## **Exponential Distribution**

- One parameter
  - $-rate \lambda$

• Applications: modeling decaying phenomena

$$X \sim Exponential(\lambda)$$

$$f(x) = \lambda e^{-\lambda x}, x \ge 0$$



# Ex for Exponential

- The mean time to failure of a critical component of an engine is 8000 hours
  - $-\lambda = 1/8000$
- Find the probability of failing before 5000 hours.

$$F(5000) = 0.465$$

```
> pexp(5000, rate=1/8000)
[1] 0.4647386
```



#### Beta Distribution

- Two parameters:
  - -a shape parameter  $\alpha > 0$
  - -a shape parameter  $\beta > 0$
- Applications:
  - Task cost and schedule modeling

$$\frac{x^{\alpha-1}(1-x)^{\beta-1}}{\mathrm{B}(\alpha,\beta)}$$

$$X \sim Beta(\alpha, \beta)$$

where 
$$B(\alpha, \beta) = \frac{\Gamma(\alpha)\Gamma(\beta)}{\Gamma(\alpha + \beta)}$$



# Visualizing Beta Distribution



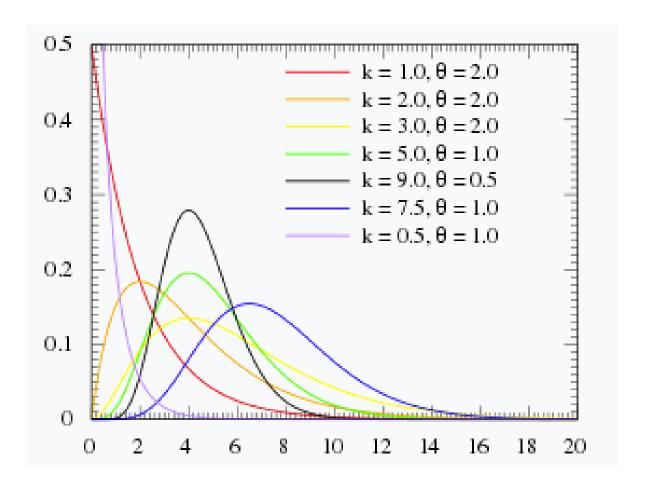
#### Gamma Distribution

- Two parameters:
  - -a shape parameter  $\alpha > 0$
  - -a rate parameter  $\beta > 0$
- Applications:
  - Size of loan defaults
  - Aggregate insurance claims
  - The amount of rainfall accumulated in a reservoir

$$X \sim Gamma(\alpha, \beta)$$

$$f(x) = rac{eta^{lpha}}{\Gamma(lpha)} x^{lpha-1} e^{-eta x}$$

## Visualizing Gamma Distribution



$$k = \alpha, \theta = 1/\beta$$



#### Weibull Distribution

- Three parameters:
  - -a location parameter  $\gamma(-\infty < \gamma < +\infty)$
  - -a scale parameter  $\alpha > 0$
  - -a shape parameter  $\beta > 0$
- Applications: reliability analysis
  - time to failure of mechanical and electrical parts

$$f(x) = \frac{\beta}{\alpha} \left(\frac{x - \gamma}{\alpha}\right)^{\beta - 1} \exp\left[-\left(\frac{x - \gamma}{\alpha}\right)^{\beta}\right], x \ge \gamma$$

#### Weibull Distribution

• Let  $\gamma = 0$ 

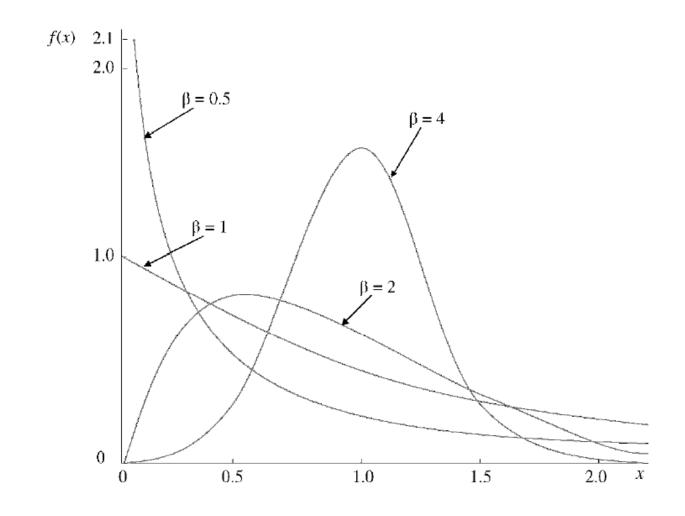
$$f(x) = \frac{\beta}{\alpha} \left(\frac{x}{\alpha}\right)^{\beta - 1} \exp\left[-\left(\frac{x}{\alpha}\right)^{\beta}\right], x \ge 0$$

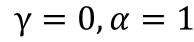
• Let  $\gamma = 0$  and  $\beta = 1$ 

$$f(x) = \frac{1}{\alpha} \exp\left[-\frac{x}{\alpha}\right], x \ge 0$$



#### PDF for Weibull







#### Ex for Weibull

- The time to failure for a cathode ray tube
   (CRT) is modeled using Weibull distribution
  - -location parameter  $\gamma = 0$
  - -scale parameter  $\alpha = 200$  hours
  - -shape parameter  $\beta = \frac{1}{3}$



• What's the probability of a tube operating for at least 800 hours?



#### Ex for Weibull

$$F(x) = 1 - \exp\left[-\left(\frac{x - \gamma}{\alpha}\right)^{\beta}\right], x \ge \gamma$$

$$F(x) = 1 - \exp\left[-\left(\frac{x}{200}\right)^{\frac{1}{3}}\right], x \ge 0$$

$$P(x > 800) = 1 - P(x \le 800)$$

$$= 1 - \left\{1 - \exp\left[-\left(\frac{800}{200}\right)^{\frac{1}{3}}\right]\right\}$$

$$= \exp\left[-\left(4\right)^{\frac{1}{3}}\right]$$

$$\approx 0.204$$

LVX VERITAS VIRTVS

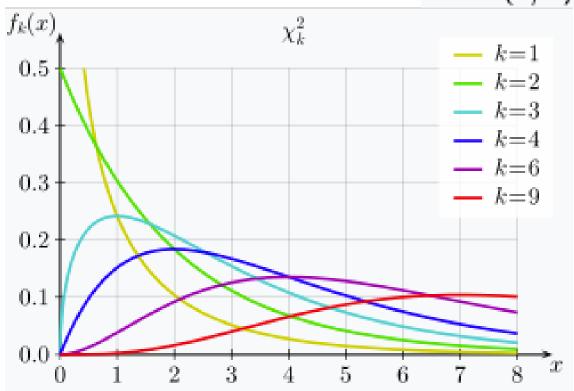
# Lecture Part IV Goodness-of-fit Test



# Chi-Square Distribution

- Based on degrees of freedom
- Positively skewed

$$rac{1}{2^{k/2}\Gamma(k/2)}\;x^{k/2-1}e^{-x/2}$$





# Chi-Square Goodness-of-Fit Test

- Degrees of freedom = no. of categories 1
- O = observed frequency
- E = expected frequency

$$\chi^2 = \sum \frac{(O-E)^2}{E}$$

Larceny thefts	Property crimes	Drug use	Driving under the influence
38	50	28	44



#### **Arrests for Crimes**

#### Step 1. Hypotheses

H0: No difference in the number of arrests for each type of crime.

Ha: There is difference.

#### Step 2. Key parameters

degree of freedom = 4-1=3

$$\alpha = 0.05$$

critical value = 7.815

```
> qchisq(0.95, df=4-1)
[1] 7.814728
```



#### **Arrests for Crimes**

#### Step 3. Calculate test value (E=n/k=160/4=40)

	Larceny thefts	Property crimes	Drug use	Driving under the influence
Observed	38	50	28	44
Expected	40	40	40	40

$$\chi^2 = \sum \frac{(0-E)^2}{E} = \frac{(38-40)^2}{40} + \frac{(50-40)^2}{40} + \frac{(28-40)^2}{40} + \frac{(44-40)^2}{40}$$
$$= 0.1 + 2.5 + 3.6 + 0.4$$
$$= 6.6$$

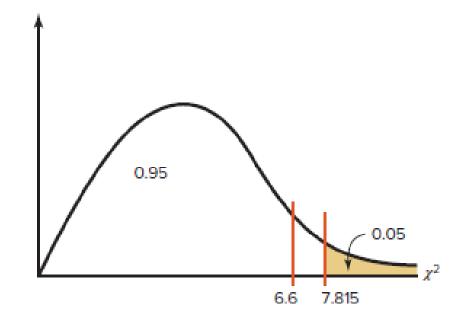


#### **Arrests for Crimes**

#### Step 4. Make decision

Since 6.6 < 7.815

Do not reject H<sub>0</sub>



#### Step 5. Summary

There is not enough evidence to reject the claim. Fail to reject.

#### Other Goodness-of-Fit Tests

- Anderson-Darling Test
  - -ad.test {goftest}

- Kolmogorov-Smirnov Test
  - -ks.test {stats}

# Example: AD Test

```
> x <- rnorm(10, mean=2, sd=1)
> ad.test(x, "pnorm", mean=2, sd=1)

Anderson-Darling test of goodness-of-fit
    Null hypothesis: Normal distribution
    with parameters mean = 2, sd = 1
    Parameters assumed to be fixed

data: x
An = 1.3598, p-value = 0.2136
```

# Example: KS Test

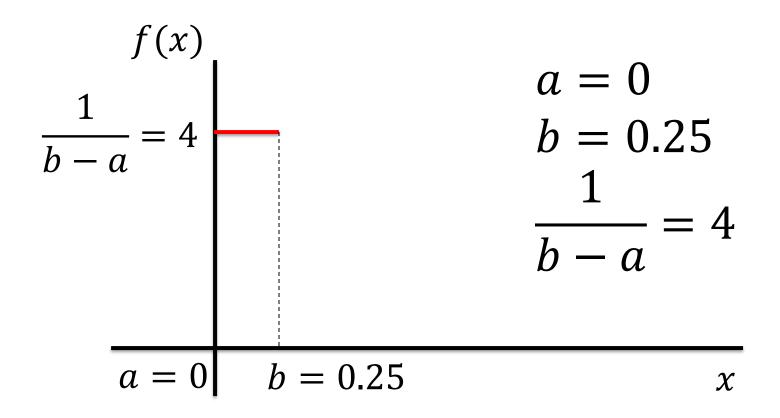
```
> x <- rnorm(50)
> y <- runif(30)
> ks.test(x, y)

Two-sample Kolmogorov-Smirnov test

data: x and y
D = 0.48, p-value = 0.0002033
alternative hypothesis: two-sided
```

# Questions?





$$Probability = Area = 4 * 0.25 = 1$$

