#### **Continuous Distributions**

#### Lecture 9

Reading: Sections 9.8 – 9.11, "Normal Table: Read It, Use It" (Optional: 9.7, 9.12, "Normal Probability Plots" pp. 280-2)

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#### Discrete versus Continuous

- Discrete random variable: can find  $P(X = x_1), ..., P(X = x_m)$
- Continuous random variable: probability that X equals a specific number is 0
  - E.g.: If measure weight perfectly at birth, P(W = 3302g) = ?
- Typically round weight to nearest gram
  - Technically a finite # of possible values
  - But treat as continuous: a good approximation
- Discrete vs. continuous:
  - # cars through a toll booth in a month?
  - # UoT snow days in a year?

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### Probability Density Function: f(x)

- Density function requirements:
  - $-f(x) \ge 0$  for all x- Total area under f(x) equals 1
- f(x) is height, not a probability
- Area under a density P(2700 < W < 3700) = 0.6827 function is probability P(W = 3302) = 0 of a range of values P(3301.5 < W < 3302.5) = 0.0008

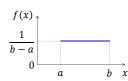
	Birth Weight Distribution mu = 3200, sigma = 500				
.0008-					
.2 .0006					
-9000. Density					
.0002					
0+	222222				
100 500 500 300 300 x00 x00					
Weight in grams					

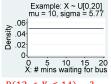
## Uniform Distribution: $X \sim U[a, b]$

- <u>Uniform</u>:  $f(x) = \frac{1}{b-a}$  for a < x < b and 0 otherwise
  - Two parameters: a and b
  - Bounded support

$$-E[X] = \mu = \frac{a+b}{2}$$

$$-V[X] = \sigma^2 = \frac{(b-a)^2}{12}$$

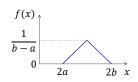




 $P(12 < X \le 14) = ?$ 

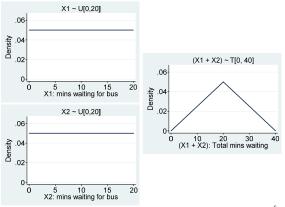
#### Triangle Distribution: $X \sim T[2a, 2b]$

 Triangle distribution created as sum of two independent and identically distributed Uniform random variables



- $-X_1$  and  $X_2$  independent
- $\ X_1{\sim}U[a,b]$  and  $\ X_2{\sim}U[a,b]$  (identically distributed)
- $-~(X_1+X_2){\sim}T[2\alpha,2b]$
- E.g. If wait time is Uniformly distributed from 0 to 20 minutes, U[0,20], then total wait time for two buses  $(X_1+X_2)$  follows the Triangle distribution: T[0,40]

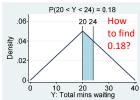
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## Finding Areas: $A = \frac{1}{2}base * height$





$$y - y_1 = m(x - x_1)$$
$$y - 0 = -\frac{0.05}{20}(x - 40)$$

$$y = 0.1 - 0.0025x$$

$$A = 0.5 * (40 - 24) * (0.1 - 0.0025 * 24) = 0.32$$

# Summary: Uniform $X \sim U[a, b]$

- Symmetric, rectangleshaped, even density throughout
- Parameters: a and b
   Bounded support: [a, b]
- Find probabilities with A = bh and symmetry

• 
$$E[X] = \mu = \frac{a+b}{2}$$

• 
$$V[X] = \sigma^2 = \frac{(b-a)^2}{12}$$

# Summary: Triangle $X \sim T[2a, 2b]$

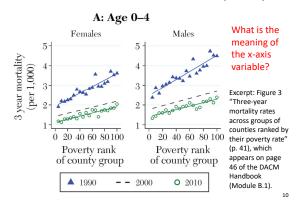
- Symmetric, triangleshaped, more density around mean
- Parameters: 2a and 2b
  Bounded sup.: [2a, 2b]
- Find probabilities with  $A = \frac{1}{2}bh$  and symmetry
- E[X] =
- V[X] =

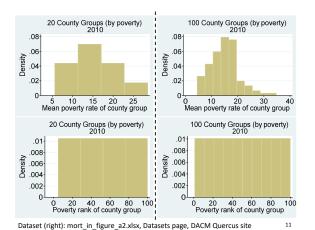
#### Percentiles & Uniform Distribution

- · Percentiles are often interesting
  - Standardized tests: for example, a score of 590 on the math SAT is at the 73<sup>rd</sup> percentile (good)
  - Birth weights: for example, a 10-month old girl who weighs 6.8 kg is at the 3<sup>rd</sup> percentile (small)
- No matter how a variable's distribution is shaped – positively skewed, negatively skewed, bi-modal, Normal (Bell) – percentiles will be Uniformly distributed from 0 to 100

The WHO Child Growth Standards, Weight-for-age, <a href="http://www.who.int/childgrowth/standards/weight\_for\_age/en/">http://www.who.int/childgrowth/standards/weight\_for\_age/en/</a>

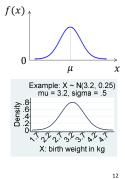
#### Recall: Currie & Schwandt (2016)





## Normal Distribution: $X \sim N[\mu, \sigma^2]$

- $f(x) = \frac{1}{\sigma\sqrt{2\pi}}e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2}$ for  $-\infty < x < \infty$ 
  - Parameters:  $\mu$  and  $\sigma^2$
  - Unbounded support
  - For area under f(x), no formula (calculus cannot solve): use tables/software



#### Recall: Empirical Rule

- In sample from Normal population, then about:
  - 68.3% of obs lie w/in 1 s.d. of mean (i.e. btwn  $\bar{X} - s$  and  $\bar{X} + s$ )
  - 95.4% of obs lie w/in 2 s.d. of mean (i.e. btwn  $\bar{X} - 2s$  and  $\bar{X} + 2s$ )
  - 99.7% of obs lie w/in 3 s.d. of mean (i.e. btwn  $\bar{X} - 3s$  and  $\bar{X} + 3s$ )
- · For the Normal model, exactly (with rounding):
  - 68.3% of obs lie w/in 1 s.d. of mean (i.e. btwn  $\mu - \sigma$  and  $\mu + \sigma$ )
  - 95.4% of obs lie w/in 2 s.d. of mean (i.e. btwn  $\mu - 2\sigma$  and  $\mu + 2\sigma$ )
  - 99.7% of obs lie w/in 3 s.d. of mean (i.e. btwn  $\mu - 3\sigma$  and  $\mu + 3\sigma$ )

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### Standard Normal: $Z \sim N(0, 1)$

f(x)

- If  $X \sim N(\mu, \sigma^2)$  then  $Z \sim N(0,1)$  where

  - transformations do NOT change shape & Laws of Expectation & Variance:

$$E\left[-\frac{\mu}{\sigma} + \frac{1}{\sigma}X\right] = -\frac{\mu}{\sigma} + \frac{1}{\sigma}E[X] = -\frac{\mu}{\sigma} + \frac{\mu}{\sigma} = 0$$

$$V\left[-\frac{\mu}{\sigma} + \frac{1}{\sigma}X\right] = \frac{1}{\sigma^2}V[X] = \frac{1}{\sigma^2}\sigma^2 = 1$$

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### **Using Standard Normal Table**

$$P(x_1 < X < x_2) = P\left(\frac{x_1 - \mu}{\sigma} < Z < \frac{x_2 - \mu}{\sigma}\right)$$

For example, if birth weights are Normally distributed with a mean of 3,200 grams and a standard deviation of 500 grams, what is the probability that a randomly selected newborn weighs less than 2,600 grams?

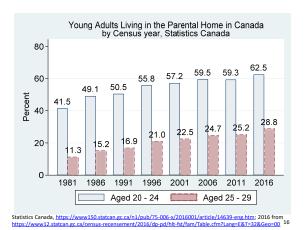
$$P(X < 2,600) = P\left(Z < \frac{2,600 - 3,200}{500}\right)$$

$$P(Z < -1.2) = 0.5 - 0.3849$$
  
= 0.1151

"Normal Table: Read it, Use it": Use one-page table for HW



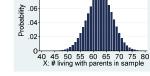
Table rounds to 4 decimal places: software gives more P(0 < Z < 1.2) = 0.38493



## Sampling (Sampling Error) Example

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 In random sample of 100 young adults aged 20-24 in 2016, what is the chance that 56 or fewer live with parents (recalling that Census says 62.5%)?



 Define X as count of those living with parents  $P(X \le 56) = P(X < 57) = ?$ 

n = 100, p = 0.625 E[X] = 62.500, SD[X] = 4.841

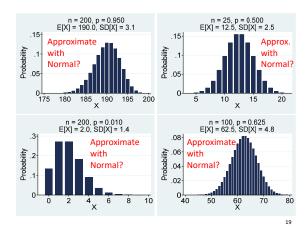
– How is X distributed?

Reasonable to compute with handheld calculator?

$$P(X=56) = \frac{100!}{56!(100-56)!}0.625^{56}(1-0.625)^{100-56} = 0.0331_{17}$$

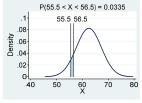
### Normal Approximation to Binomial

- $X \sim B(n, p)$  is sometimes well approximated by  $X \sim N(\mu, \sigma^2)$  w/  $\mu = np$  and  $\sigma^2 = np(1-p)$ 
  - Two common rules of thumb (you may use either) to check if it's a reasonable approximation:
    - 1)  $np \ge 10$  and  $n(1-p) \ge 10$
    - 2)  $np 3\sqrt{np(1-p)} \ge 0$  and  $np + 3\sqrt{np(1-p)} \le n$
- As *n* rises, Normal approximation improves
  - There is nothing magic about marginally passing or marginally failing either rule of thumb



### **Continuity Correction**

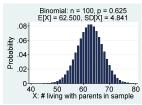
- · Binomial is discrete  $-P(X=56) \neq 0 \text{ but}$ P(56 < X < 57) = 0
- Normal is continuous -P(X = 56) = 0 but  $P(56 < X < 57) \neq 0$
- · "Continuity correction" addresses this disparity
  - Page 286 textbook (box)

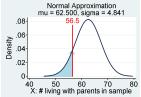


 $P(X=56)\approx 0.0335$ 

To approximate  $P(X \le 56)$ , use Normal to find P(X < 56.5)

*Note:* Correction (slightly) improves the approximation. With a large nit makes little difference: researchers usually ignore it.





 $P(X \le 56) \approx P(X < 56.5)$ 

$$P(X < 56.5)$$
=  $P\left(Z < \frac{56.5 - 62.5}{4.841}\right)$ 
=  $P(Z < -1.24)$ 
=  $0.5 - 0.3925 = 0.1075$ 

Is having so few (56) living at home in a sample of 100 surprising?

#### Normal + Normal = Normal?

- <u>Linear combination</u>:  $a_0 + a_1 X_1 + \dots + a_J X_J$  where  $a_0, a_1, \dots, a_J$  are constants
- A linear combination of independent Normal random variables yields a Normal random variable

In 2016, chance a random sample of 50 people aged 25-29 has *more* living w/ parents than sample of 50 aged 20-24?

$$P(X_{25-29} > X_{20-24}) = ?$$

$$P((X_{25-29}-X_{20-24})>0)=?$$

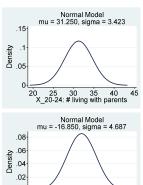
$$X_{20-24} \sim B(n = 50, p = 0.625)$$

$$X_{25-29} \sim B(n = 50, p = 0.288)$$

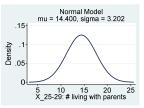
Use Normal approx. for each?

If so, is difference Normal?

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-20 -10 (X\_25-29 - X\_20-24)



$$P((X_{25-29}-X_{20-24})>0)=?$$

$$P(Z>\frac{0--16.850}{4.687})$$

$$=P(Z>3.60)=0.0002$$
Would it be surprising if  $X_{25-29}$ 

were larger than  $X_{20-24}$ ?

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