

# Quantitative Research Methods

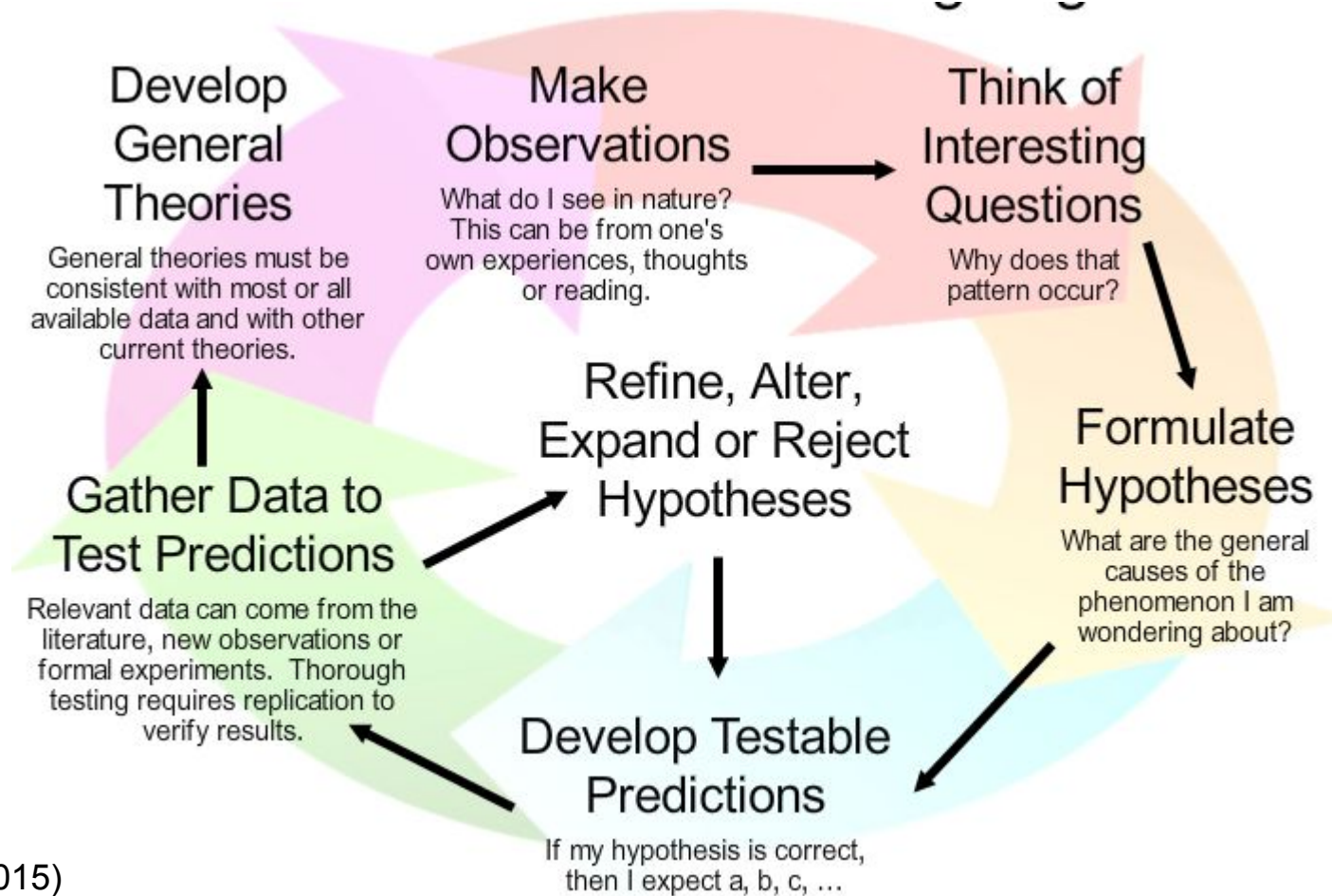
Probability and Statistics for Data Science  
CSE594 - Spring 2016

# Quantitative Research Methods

Probability and Statistics for Data **Science**

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# The Scientific Method



(Garland, 2015)

# Hypothesis Testing

Hypothesis -- something one asserts to be true.

Classical Approach:

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$H_1$ : *the alternative* -- usually that one’s “hypothesis” is true

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**Goal:** Use probability to determine if we can “reject the null” ( $H_0$ ) in favor of  $H_1$ .  
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$H_0$ : the coin is not biased (i.e. flipping  $n$  times results in a Binomial( $n$ , 0.5))



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More formally: Let  $X$  be a random variable and let  $R$  be the range of  $X$ .  $R_{\text{reject}} \subset R$  is the *rejection region*. If  $X \in R_{\text{reject}}$  then we reject the null.

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in the example, if  $n = 1000$ , then then  $R_{\text{reject}} = [0, 469] \cup [531, 1000]$

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# Hypothesis Testing

Example: Communities with higher population have different amounts of violent crimes (per capita) than those with lower population.

Assignment 1, Programming Problem “C) 9.”



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$$t = \frac{\bar{x}_1 - \bar{x}_2}{\sqrt{s_{x_1} + s_{x_2}} \cdot \sqrt{\frac{1}{n}}}$$

t statistic for 2 iid (independent, identically distributed) samples

# Hypothesis Testing

Important logical question:

Does failure to reject the null mean the null is true?

