#### Mathematics and Problem Solving Lecture 11.2

Hypothesis Testing

#### Starting with a question

- Often you'll come to data with a question:
  - "Do players prefer character A or character B?"

#### Hypotheses

- To answer this with statistics, you need to turn it into **two** hypotheses:
  - Null Hypothesis
  - Alternate Hypothesis

## Null Hypothesis

- The null hypothesis says that
  - whatever results you got were the result of chance
  - i.e. nothing interesting is happening
- e.g. "Players like both characters more or less the same"

## Alternate Hypothesis

- Your **alternate hypothesis** is what you're trying to prove
  - e.g. "Players either prefer Character A or they prefer character B"
  - This is the hypothesis that we're going to test

- Now, imagine we find the mean score for characters A and B to be 4.5 and 4.7
- Can we confirm our hypothesis?
  - No, because we only sampled our population, the difference might be chance
- Inferential statistics tell you how likely it is that the effect you observe is the result of chance.



- Instead, imagine we have the distributions of the results.
  - Now can we confirm our hypothesis?



#### **Statistics**

- Key values
  - Test statistic (varies)
  - p
  - α
  - Effect size (if possible)

## **Statistical Significance**

- p = probability the effect observed was chance
  - p < α
    - statistically significant
  - p > α
    - not statistically significant

# Alpha

- Alpha ( $\alpha$ ) is usually set at 0.05
- Reduced for multiple testing

#### **Effect Size**

- How big an effect is
  - e.g. Cohen's d
    - *d* = 0.2 is small
    - *d* = 0.5 is medium
    - *d* = 0.8 is large
- Important for interpreting results

## Type I error

- Falsely rejecting the null hypothesis
  - we claim an effect when there is no effect
- If  $\alpha$  = 0.05, we have a 5% chance of making a Type I error.

## Type II error

- Falsely accepting the null hypothesis
  - there is an effect, but you don't detect it
- The probability of making a type II error is called beta (β)
  - This is related to your statistical **power**. (=  $1-\beta$ )
  - A good **power** is 0.8 ( $\beta$  = 0.2)





4.7

# In closing

- Inferential statistics = lots of numbers about your data
- It's frighteningly easy to make mistakes
  - If that is in science that people rely on, that's a big problem