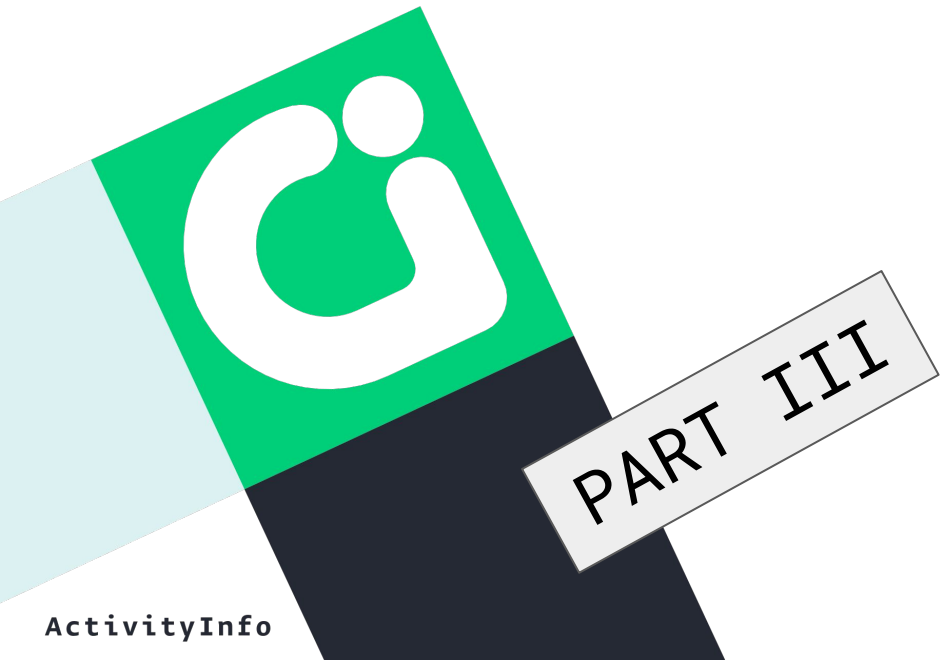


# Measuring Impact Quantitatively

May 10th, 2022

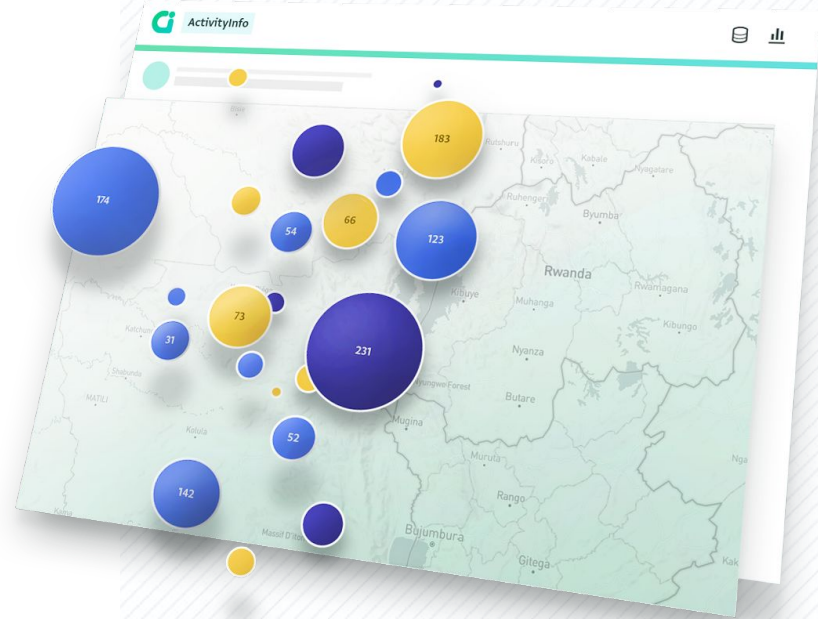
A graphic in the bottom-left corner featuring a green square with a white stylized 'i' logo, a dark blue diagonal bar, and a light blue triangle. A grey rectangular label with the text 'PART III' is tilted and placed over the dark blue bar.

PART III

# Presented by the ActivityInfo Team

## Monitoring & Evaluation Software

- Track activities, outcomes
- Beneficiary management
- Surveys
- Work offline / online



# Outline

1. Quick review
2. Statistical significance
3. Effect size
4. Questions!



Quick review

# Quantitative Impact Evaluation

Causal impact

Outcome Y  
*without* the  
program

$$\Delta = (Y \mid P = 1) - (Y \mid P = 0)$$

Outcome Y *with*  
the program

# Key points from Part I

- Why conduct a quantitative impact evaluation?
- When would you not conduct a quantitative impact evaluation?
- Types of measurements
- Sources of measurement error
- Reliability: Cronbach's alpha
- Cognitive interviewing, a tool for improving questionnaires

## Key point from Part II

- Fundamental problem of causal inference
- Identify four strategies for “counterfeit counterfactuals”
- Identify risks of before-and-after comparisons



Statistical significance



## Null hypothesis

The hypothesis that our program has had **ZERO** impact.



# Null hypothesis

Causal impact

Outcome Y  
*without* the  
program

$$0 = (Y \mid P = 1) - (Y \mid P = 0)$$

Outcome Y *with*  
the program

# Significance testing

Can we disprove the null hypothesis?

# Significance testing

IF the null hypothesis was true, what are the chances of sampling the data we sampled?



# Hypothesis testing: simple example

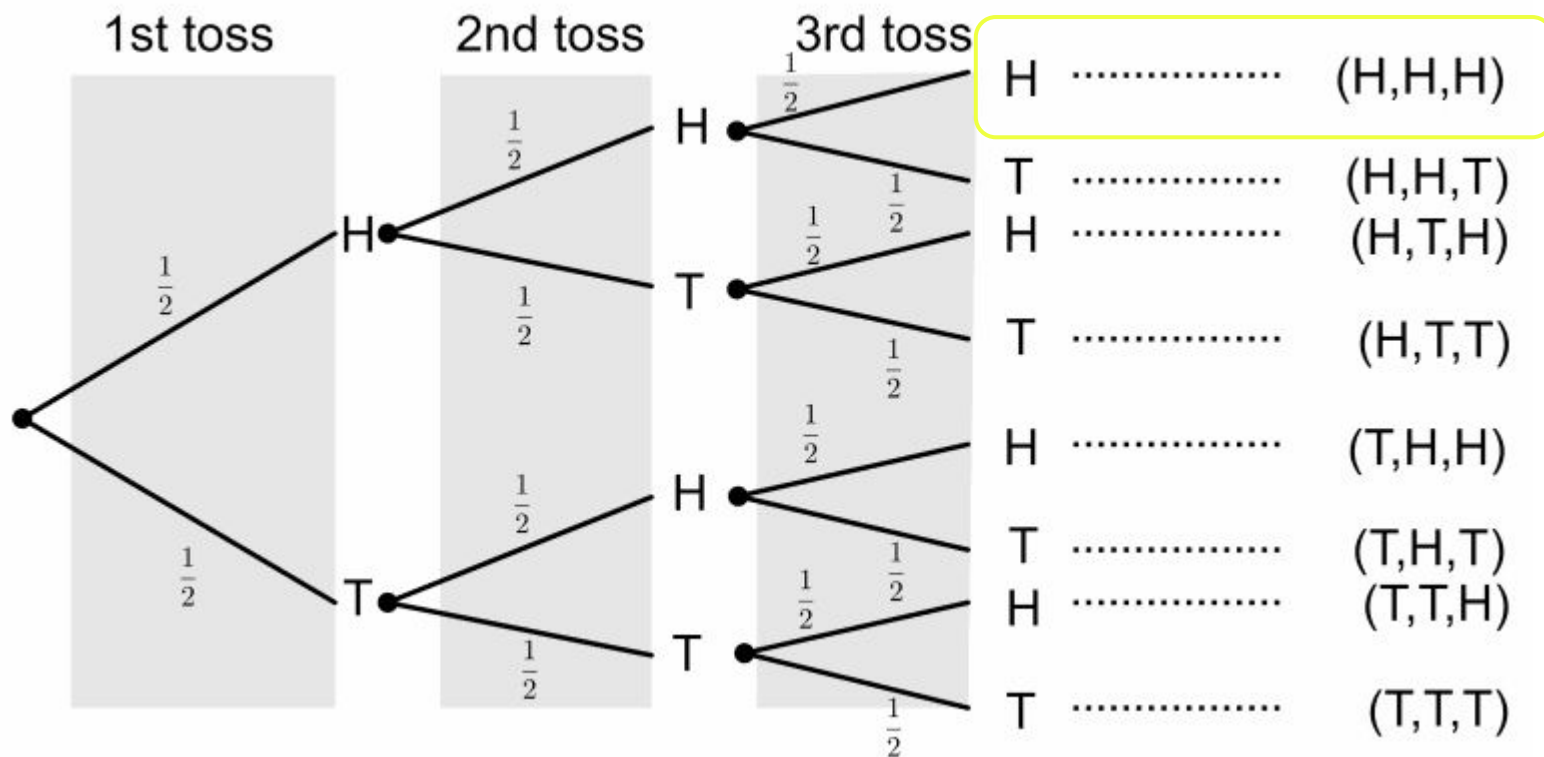
**Hypothesis:** the coin is fair --  
the odds of flipping a coin are 50/50

**Experiment:** 3 heads, 0 tails

**What are the chances if the coin is fair?**







# Hypothesis testing: simple example



**Hypothesis:** the coin is fair --  
the odds of flipping a coin are 50/50

**Experiment:** 3 heads, 0 tails

**What are the chances if the coin is fair?**  
12.5%,  $p = 0.125$

**Can we reject the hypothesis?**

No, this is not *so* unusual, even if the coin was fair.

# Hypothesis testing: simple example



**Hypothesis:** the coin is fair --  
the odds of flipping a coin are 50/50

**Experiment:** 8 heads, 2 tails

**What are the chances if the coin is fair?**  
4% probability,  $p = 0.04$

**Can we reject the hypothesis?**

**Borderline.** This is somewhat unusual,  
but it would still happen ~ 1 out of 20  
times.



# Hypothesis testing: simple example



**Hypothesis:** the coin is fair --  
the odds of flipping a coin are 50/50

**Experiment:** 19 heads, 1 tails

**What are the chances if the coin is fair?**  
0.001907% probability,  $p < 0.0001$

**Can we reject the hypothesis?**

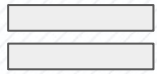
**Yes!** If the coin was fair the chances of  
this happening are freakishly low.

# Hypothesis testing: impact

$$(Y \mid P = 1) = 60\$/\text{month}$$

$$(Y \mid P = 0) = 50\$/\text{month}$$

+10\$



**Randomly to  
receive training**

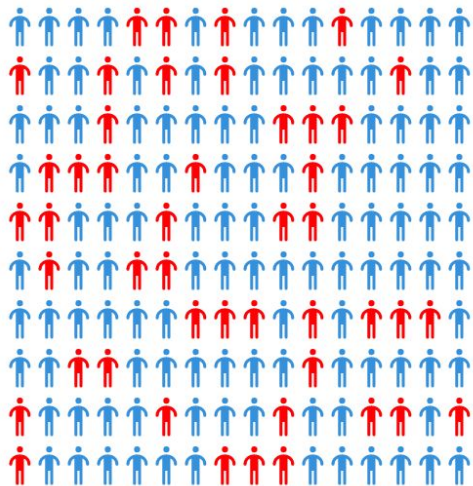


**Control group who  
received no training**





# Hypothesis testing



**(Null) Hypothesis:** our training program has zero impact on incomes

**Experiment:** Sample 50 people from each group, difference is +10 USD.

**What are the chances if our training program has no impact?**

# Calculating probability of mean difference

Student's t-test calculates the probability of getting this much difference in averages, if the null hypothesis were true.



# Data collection > Analysis



INCOME SURVEY [CANCEL RECORD ENTRY](#)

FIELD 3 OF 3

**Income over the last 30 days\*** REQUIRED

How much did you earn over the last 30 days?

48

[Previous](#) [Submit record](#)

quant-webinar - RStudio

File Edit Code View Plots Session Build Debug Profile Tools Help

Go to file/function Addins

Untitled1\*

```
1 library(activityinfo)
2
3 survey <- queryTable(form = "cftn7a7l30lb9ox2")
4
5 training <- subset(survey, group == "Training")
6 control <- subset(survey, group == "Control")
7
8 t.test(training$income, control$income,
9       alternative = "two.sided",
10      var.equal = FALSE)
11
```

11:1 (Top Level)

Console Terminal Jobs

R 4.1.2 · ~/dev/quant-webinar/

data: training\$income and control\$income

t = 4.845, df = 97.115, p-value = 4.792e-06

alternative hypothesis: true difference in means is not equal to 0

95 percent confidence interval:

6.104355 14.575645

sample estimates:


mean of x mean of y

59.40 49.06

>



# t-Tests with Excel

 T-Test ☆ 📁 ☁

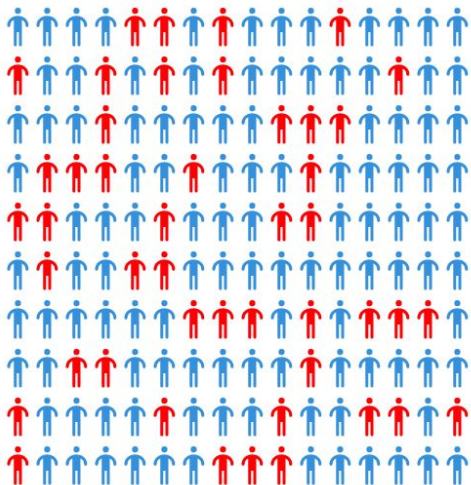
File Edit View Insert Format Data Tools Extensions Help [Last edit was seconds ago](#)

↶ ↷ 🖨 📄 100% ▾ \$ % .0 .00 123 ▾ Arial ▾ 10 ▾ **B** *I* S A 🔍 📊 📈 ▾

E3 ▾  $\text{fx}$  =T.TEST(C2:C51,C52:C101,2,3)

	A	B	C	D	E	F	G	H
1	Name	Group	Income					
2	Dr. Lexie Boyle	Control	50					
3	Julia Jenkins	Control	53		0.0004792%			
4	Miss Sadye Moen	Control	54					
5	Bernard Blanda	Control	59					
6	Susanna Klein	Control	55					
7	Armand Stanton DVM	Control	45					

# Hypothesis testing



**(Null) Hypothesis:** our training program has zero impact on incomes

**Experiment:** Sample 50 people from each group, difference is +10 USD.

**What are the chances if our training program has no impact?** 0.0004792%

**Can we reject the hypothesis?**

**Yes!** If our program had no impact, there would be practically no way of getting these results by chance.

# Hypothesis standards

$p < 0.05$  \*      “statistically significant”

$p < 0.01$  \*\*      “highly statistically significant”

$p < 0.001$  \*\*\*



# Hypothesis testing - pitfalls

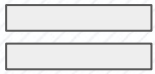
“Not zero” is a very low standard is a very low standard for impact!

# Hypothesis testing: small effects

$$(Y \mid P = 1) = 51\$/\text{month}$$

$$(Y \mid P = 0) = 50\$/\text{month}$$

+1\$



**Randomly to  
receive training**



**Control group who  
received no training**



# Miniscule differences can still be “significant” with enough data!

Impact of program: +1 USD

Sample size 50,  $p = 0.519$

Sample size 100,  $p = 0.236$

Sample size 500,  $p = 0.014^*$

Sample size 1000,  $p = 0.000^{***}$

Effect size

## Effect size

Statistics which help understand the **size** of our impact, not just whether the impact is non-zero.



# Examples of effect size statistics

- Difference in means (averages)
- Cohen's d
- Odds ratios
- Eta-squared



# Revisiting our small size example

Impact of program: +1 USD

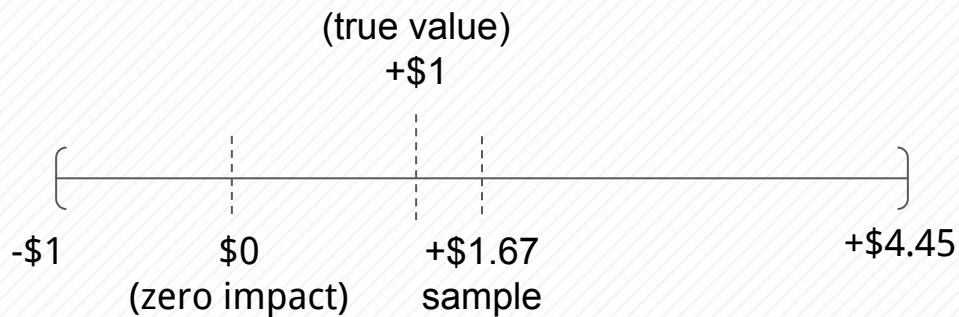
Sample size 50,  $p = 0.519$ , [-2.42 - 4.75]

Sample size 100,  $p = 0.236$ , [-1.11 - 4.45]

Sample size 500,  $p = 0.014$ , [0.34 - 2.96]

Sample size 1000,  $p = 0.000$ , [1.12 - 2.95]

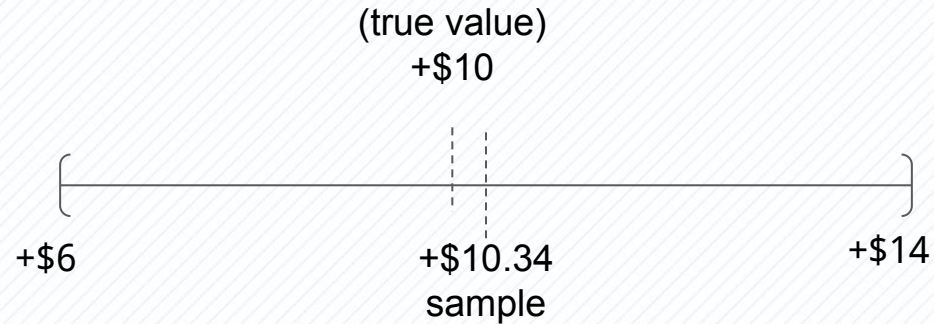
# Confidence intervals



Sample size 100,  $p = 0.236$ ,  $[-1.11 - 4.45]$



# Confidence intervals



Sample size 50,  $p < 0.001$ , [6.10 - 14.57]

# Cohen's d

- *Standardized* mean difference
- Useful for values without easily interpretable units

$$d = \frac{\bar{x}_1 - \bar{x}_2}{s}.$$

# Interpreting Cohen's $d$

Roughly speaking, how much variation is explained

0.5494623 1.3885407

<i>Effect size</i>	<i>d</i>
Very small	0.01
Small	0.20
Medium	0.50
Large	0.80
Very large	1.20
Huge	2.0

## Note on complex samples

If your sample uses clusters or stratification, you **cannot** use a simple t-Test. See the R survey package:

<https://cran.r-project.org/web/packages/survey/index.html>

```
design <- svydesign(id=~dnum+snum, data=training)  
svyttest(enroll~income, dclus2)
```



Communicating about effect size



# Try ActivityInfo!

The most complete  
database system for M&E  
available.

[Try a template](#)

Demo Webinar:

May 31st, 14:00 CEST

