

# Sample Size Determination with Examples using Sample Size Calculator (web)

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Last update: Dec 20, 2023

# Outlines

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# Software used

- Sample Size Calculator (web)

[https://wnarifin.github.io/ssc\\_web.html](https://wnarifin.github.io/ssc_web.html)



# Introduction

# Why calculate?

- An important part of a research design & plan
- Determines:
  - ✓ Number of participants
  - ✓ Funding
  - ✓ Feasibility
  - ✓ Validity

# Steps toward calculation

- (1) Define the research **objective**
- (2) Decide on the suitable **statistical analysis**
- (3) Determine the **sample size**

# Define objective

- Clearly define objective
- Outcome (dependent variables)?
- Predictors (independent variables)?
- Measurable

# Statistical analysis

- Choose suitable statistical analysis based on:
  - objective itself – compare, estimate, correlate, validate ...
  - the scale of outcome & predictors
  - sample – independent / dependent
- Examples:
  - Independent  $t$ -test for ...
  - Chi-squared test for ...
  - Linear regression for ...
  - Logistic regression for ...



# Sample size formula

- Choose suitable sample size formula / method
- Based on objective & chosen statistical analysis
- Estimation vs Hypothesis Testing
- Examples:
  - Confidence interval – estimation
  - Comparison of means – hypothesis testing

# General

# One mean

## (1) Objective:

- Estimate mean of numerical outcome in a population e.g. BMI, weight etc
- "This study aims to estimate mean BMI among UiTM students."

# One mean

(2) Statistical analysis:

- Mean, 95% Confidence Interval (CI)

(3) Sample size method:

- Single mean – Estimation

# 1. Means → Single Mean → 1 mean – Estimation

🏠 >> Sample Size Calculator

Sample Size Calculator (web)

**1 mean - Estimation**

Standard deviation ( $\sigma$ ):

Precision ( $\pm$  mean):

Confidence level  $100(1 - \alpha)$ :  %

Expected dropout rate:  %

Sample size,  $n =$

Sample size (with 10% dropout),  $n_{\text{drop}} =$

SD of BMI from other studies

Precision, in unit of measurement

Confidence level 90%, **95%**, 99%

% of participants that may dropout

# One mean

- Let say:
  - SD of BMI =  $7.5\text{kg/m}^2$
  - Precision =  $1\text{kg/m}^2$
  - 95% Confidence level
  - 20% dropout
- How many UiTM students should we sample?

# One mean

 >> Sample Size Calculator

Sample Size Calculator (web)

## 1 mean - Estimation

Standard deviation ( $\sigma$ ):

7.5 

Precision ( $\pm$  mean):

1 

Confidence level  $100(1 - \alpha)$ :

95  %

Expected dropout rate:

20  %

Calculate

Reset

Sample size,  $n =$

217 

Sample size (with 20% dropout),  $n_{\text{drop}} =$

272 

# One mean

- We have to sample 272 students to estimate mean BMI among UiTM students, with mean BMI  $\pm 1\text{kg}/\text{m}^2$ , taking into account a 20% dropout rate.



# One proportion

## (1) Objective:

- Estimate % or proportion of categorical outcome in a population e.g. obesity, diabetes etc
- "This study aims to estimate prevalence of obesity among UiTM students."

# One proportion

(2) Statistical analysis:

- Proportion, 95% Confidence Interval (CI)

(3) Sample size method:

- Single proportion – Estimation

## 2. Proportions → Single proportion → 1 proportion – Estimation

🏠 >> Sample Size Calculator

Sample Size Calculator (web)

1 proportion - Estimation

Proportion (p):

0.50

Precision ( $\pm$  proportion):

0.05

Confidence level  $100(1 - \alpha)$ :

95 %

Expected dropout rate:

10 %

Calculate

Reset

Sample size,  $n =$

Sample size (with 10% dropout),  $n_{\text{drop}} =$

Proportion of obesity from other studies

Precision, in proportion

Confidence level 90%, **95%**, 99%

% of participants that may dropout

# One proportion

- Let say:
  - Prevalence of obesity = 25% = 0.25 (in proportion).
  - Precision =  $\pm 5\%$  = 0.05 (in proportion).
  - 95% Confidence level.
  - 10% dropout (i.e. those who won't let you measure their weight and height).
- How many UiTM students should we sample?

# One proportion

 >> Sample Size Calculator

## Sample Size Calculator (web)

### 1 proportion - Estimation

Proportion (p):	<input type="text" value="0.25"/>
Precision ( $\pm$ proportion):	<input type="text" value="0.05"/>
Confidence level $100(1 - \alpha)$ :	<input type="text" value="95"/> %
Expected dropout rate:	<input type="text" value="10"/> %
<input type="button" value="Calculate"/> <input type="button" value="Reset"/>	
Sample size, n =	<input type="text" value="289"/>
Sample size (with 10% dropout), $n_{\text{drop}}$ =	<input type="text" value="322"/>

# One proportion

- We have to sample 322 students to estimate prevalence of obesity among UiTM students, with % of obesity  $\pm 5\%$ , taking into account a 10% dropout rate.

# Two means

## (1) Objective:

- Compare means of a numerical outcome in two populations e.g. BMI, weight etc
- "This study aims to compare mean BMI between 1st and final year students."

# Two means

(2) Statistical analysis:

- Independent t-test

(3) Sample size method:

- Two means – Hypothesis testing



# 1. Means → Two-mean comparison (independent) → 2 means – Hypothesis Testing

## 🏠 >> Sample Size Calculator

### Sample Size Calculator (web)

#### 2 means - Hypothesis Testing

Standard deviation ( $\sigma$ ):

Expected difference:

Significance level ( $\alpha$ ):

Two-tailed

Power ( $1 - \beta$ ):

%

Expected dropout rate:

%

Calculate

Reset

Sample size,  $n =$

Sample size (with 10% dropout),  $n_{\text{drop}} =$

SD of BMI  
from literature

Expected difference  
between two pops

Significance level  
usually 0.05

Power 80%

% of participants  
that may dropout

# Two means

- Let say:
  - Largest SD you could find from literature = 1.5
  - Expected Difference = 1 unit.
  - Significance level = 5% (0.05)
  - Leave Power = 80% – default value.
  - 30% dropout (i.e. as some weight themselves while only one foot was on the scale).
- How many students per group should we sample?

# Two means

## >> Sample Size Calculator

### Sample Size Calculator (web)

#### 2 means - Hypothesis Testing

Standard deviation ( $\sigma$ ):  

Expected difference:  


Significance level ( $\alpha$ ):   Two-tailed


Power ( $1 - \beta$ ):   %

Expected dropout rate:   %

Calculate

Reset

Sample size,  $n =$   

Sample size (with 30% dropout),  $n_{\text{drop}} =$   

# Two means

- We have to sample 52 1st year students and 52 final year students to make the comparison, expecting a difference of 1 unit BMI between the two, taking into account a 30% dropout rate.

# Two proportions

## (1) Objective:

- Compare % or proportion of categorical outcome in two populations e.g. obesity, diabetes etc
- "This study aims to compare prevalence of obesity between 1st and final year students."

# Two proportions

(2) Statistical analysis:

- Chi-squared test

(3) Sample size method:

- Two proportions – Hypothesis testing

# 2. Proportions → Two-proportion comparison (independent) → 2 proportions – Hypothesis Testing

## 🏠 >> Sample Size Calculator

### Sample Size Calculator (web)

#### 2 proportions - Hypothesis Testing

Proportion in control ( $p_0$ ):

Proportion in case ( $p_1$ ):

Significance level ( $\alpha$ ):

Two-tailed

Power ( $1 - \beta$ ):

%

Expected dropout rate:

%

Sample size,  $n =$

Sample size (with 10% dropout),  $n_{\text{drop}} =$

Prop of baseline group from literature

Prop of comparison group, hypothesized

Significance level usually 0.05

Power 80%

% of participants that may dropout

# Two proportions

- Let's say:
  - $p_0 = 35\% = 0.35$  (in proportion)  $\rightarrow$  1st year students as control.
  - $p_1 = 50\% = 0.5$  (in proportion)  $\rightarrow$  Hypothesized % for final year students.
  - Significance level = 5% (0.05)
  - Power = 80%
  - 10% dropout
- How many students per group should we sample?



# Two proportions

 >> Sample Size Calculator

Sample Size Calculator (web)

## 2 proportions - Hypothesis Testing

Proportion in control ( $p_0$ ):

Proportion in case ( $p_1$ ):

Significance level ( $\alpha$ ):  Two-tailed

Power ( $1 - \beta$ ):  %

Expected dropout rate:  %

Calculate

Reset

Sample size,  $n =$

Sample size (with 10% dropout),  $n_{\text{drop}} =$

# Two proportions

- We have to sample 189 1st year students and 189 final year students to make the comparison, expecting a difference of 15% for % of obesity between the two, taking into account a 10% dropout rate.

# Two means (paired)

## (1) Objective:

- Compare means of a numerical outcome in a population, pre-post design e.g. weight before and after intervention
- "This study aims to compare mean weight before and after a weight loss program among students."

# Two means (paired)

(2) Statistical analysis:

- Paired t-test

(3) Sample size method:

- Two means (paired) – Hypothesis testing

# 1. Means → Two-mean comparison (paired) → 2 means (paired) – Hypothesis Testing

## Sample Size Calculator (web)

### 2 means (paired) - Hypothesis Testing<sup>1,2</sup>

Standard deviation of difference ( $\sigma_d$ )\*:

If  $\sigma_d$  not available in literature, use **Standard deviation of difference** calculator below to get an approximate value of  $\sigma_d$ .

Expected difference:

Significance level ( $\alpha$ ):

Power ( $1 - \beta$ ):

Expected dropout rate:

Sample size,  $n =$

Sample size (with 10% dropout),  $n_{\text{drop}} =$

SD difference in weight from literature

Expected difference between pre-post

Significance level usually 0.05

Power 80%

% of participants that may dropout

Two-tailed

# Two means (paired)

- Let say:
  - SD of difference from literature = 2.5kg
  - Expected Difference = 5kg (post – pre weight)
  - Significance level = 5% (0.05)
  - Power = 80%
  - 20% dropout
- How many students should we sample?

# Two means (paired)

## 🏠 >> Sample Size Calculator

### Sample Size Calculator (web)

**2 means (paired) - Hypothesis Testing<sup>1,2</sup>**

Standard deviation of difference ( $\sigma_d$ )\*:

If  $\sigma_d$  not available in literature, use **Standard deviation of difference** calculator below to get an approximate value of  $\sigma_d$ .

Expected difference:

Significance level ( $\alpha$ ):  Two-tailed

Power ( $1 - \beta$ ):  %

Expected dropout rate:  %

Sample size,  $n =$

Sample size (with 20% dropout),  $n_{\text{drop}} =$

# Two means (paired)

- We have to sample 23 students to make the comparison, expecting 5kg change following the weight reduction program, taking into account a 20% dropout rate.



# Two proportions (paired)

## (1) Objective:

- Compare % or proportion of a categorical outcome in a population, pre-post design e.g. % vaccine uptake after before and after health promotion
- "This study aims to compare percentage of vaccine uptake among villagers before and after a vaccine awareness campaign."

# Two proportions (paired)

(2) Statistical analysis:

- McNemar's test

(3) Sample size method:

- Two proportions (paired) – Hypothesis testing

# 2. Proportions → Two-proportion comparison (paired) → 2 proportions (paired) – Hypothesis Testing

🏠 >> Sample Size Calculator

Sample Size Calculator (web)

Two-proportion comparison (paired) [McNemar's test]

Proportion before ( $p_{\text{before}}$ ):

Proportion after ( $p_{\text{after}}$ ):

Significance level ( $\alpha$ ):

Two-tailed

Power ( $1 - \beta$ ):

%

Expected dropout rate:

%

Calculate

Reset

Sample size,  $n =$

Sample size (with 10% dropout),  $n_{\text{drop}} =$

Prop baseline from literature

Prop hypothesized

Significance level usually 0.05

Power 80%

% of participants that may dropout

# Two proportions (paired)

- Let's say:
  - $p_{\text{before}} = 50\% = 0.5$  (in proportion)  $\rightarrow$  before campaign
  - $p_{\text{after}} = 80\% = 0.8$  (in proportion)  $\rightarrow$  after campaign
  - Significance level = 5% (0.05)
  - Power = 80% (0.8)
  - 10% dropout
- How many villagers should we sample?

# Two proportions (paired)

 >> Sample Size Calculator

Sample Size Calculator (web)

Two-proportion comparison (paired) [McNemar's test]

Proportion before ( $p_{\text{before}}$ ):	<input type="text" value="0.5"/>	
Proportion after ( $p_{\text{after}}$ ):	<input type="text" value="0.8"/>	
Significance level ( $\alpha$ ):	<input type="text" value="0.05"/>	Two-tailed
Power ( $1 - \beta$ ):	<input type="text" value="80"/>	%
Expected dropout rate:	<input type="text" value="10"/>	%

Sample size, $n =$	<input type="text" value="42"/>
Sample size (with 10% dropout), $n_{\text{drop}} =$	<input type="text" value="47"/>

# Two proportions (paired)

- We have to sample 47 villagers to make the comparison, expecting 30% change in vaccine uptake following the vaccine awareness campaign, taking into account a 10% dropout rate.

# Correlation

## (1) Objective:

- Determine correlation between two numerical outcomes e.g. between age and cholesterol level etc
- "This study aims to determine the correlation between age and cholesterol level among lecturers."

# Correlation

(2) Statistical analysis:

- Pearson's correlation coefficient

(3) Sample size method:

- Pearson's correlation – Hypothesis testing ( $H_0: r = 0$ )



# 3. Pearson's correlation → Hypothesis Testing

## 🏠 » Sample Size Calculator

### Sample Size Calculator (web)

#### Pearson's Correlation - Hypothesis Testing<sup>1</sup>

Expected correlation ( $r$ ):

Expected  $r$ , can be estimated from literature

Significance level ( $\alpha$ ):

Significance level usually 0.05

Two-tailed

Power ( $1 - \beta$ ):

%

Power 80%

Expected dropout rate:

%

% of participants that may dropout

Calculate

Reset

Sample size,  $n =$

Sample size (with 10% dropout),  $n_{\text{drop}} =$

# Correlation

- Let say:
  - Correlation  $r$  that you expect, educated guess from literature = 0.55
  - Significance level = 5% (0.05)
  - Power = 80%
  - 30% dropout
- How many lecturers should we sample?

# Correlation

## >> Sample Size Calculator

### Sample Size Calculator (web)

**Pearson's Correlation - Hypothesis Testing<sup>1</sup>**

Expected correlation ( $r$ ):	<input type="text" value="0.55"/>	
Significance level ( $\alpha$ ):	<input type="text" value="0.05"/>	Two-tailed
Power ( $1 - \beta$ ):	<input type="text" value="80"/>	%
Expected dropout rate:	<input type="text" value="30"/>	%

Sample size, $n =$	<input type="text" value="23"/>
Sample size (with 30% dropout), $n_{\text{drop}} =$	<input type="text" value="33"/>

# Correlation

- We have to sample 33 lecturers to determine the correlation between age and cholesterol level, taking into account a 30% dropout rate.

# Multivariable, numerical outcome

## (1) Objective:

- Determine associated factors of a numerical outcome, e.g. predictors of cholesterol level etc
- "This study aims to determine the associated factors of cholesterol level among lecturers."

# Multivariable, numerical outcome

## (2) Statistical analysis:

- Multiple linear regression

## (3) Sample size method:

- Rule-of-thumb – 10 subjects per independent variable\*

$$- n = k \times 10$$

\*Norman, G. R., & Streiner, D. L. (2008). Biostatistics: the bare essentials. Ontario, Canada: BC Decker Inc.

# Multivariable, numerical outcome

- Let's say:
  - 1 numerical outcome (cholesterol)
  - 3 numerical independent variables:
    - Age in years, BMI in  $\text{kg}/\text{m}^2$ , weekly physical activity in hours
  - 3 categorical independent variables:
    - Male = yes, no (binary)
    - Smoking = yes, no (binary)
    - Race = Malay, Chinese, Indian (3 levels)
  - 20% dropout
- How many lecturers should we sample?

# Multivariable, numerical outcome

Independent Variables, k	Count
Age (numerical)	1
BMI (numerical)	1
Weekly physical activity (numerical)	1
Male (binary categorical)	1
Smoking (binary categorical)	1
Race (categorical, 3 levels → 2 dummy variables)	2
Total	7

Formula:

$$n = k \times 10$$

Independent variable count,  $k = 7$

Sample size,  $n$ ?

$$n = k \times 10 = 7 \times 10 = 70$$

$n + 20\%$  dropout?

$$\begin{aligned}n_{\text{drop}} &= n / (1 - \text{dropout proportion}) \\ &= n / (1 - 0.2) \\ &= 70 / 0.8 = 87.5\end{aligned}$$

**Sample size is 88 (rounded up).**



# Multivariable, numerical outcome

- We have to sample 88 lecturers to investigate the associated factors of cholesterol level among lecturers, taking into account a 20% dropout rate.

# Multivariable, binary outcome

## (1) Objective:

- Determine associated factors of a binary outcome, e.g. predictors of hypertension (HPT)
- "This study aims to determine the associated factors of HPT among workers in ABC office."

# Multivariable, binary outcome

## (2) Statistical analysis:

- Multiple logistic regression

## (3) Sample size method:

- Rule-of-thumb – 10 events-per-parameter\*
  - Event,  $n_1$  = number with outcome e.g. CAD = yes
  - Parameters = Indep. Variables + (Intercept + Interaction Terms)
  - Sample size,  $n = n_1 + n_0$  (w/out outcome) =  $n_1 / p$  (prevalence)

\*Hosmer, D., Lemeshow, S., & Sturdivant, R. (2013). Applied logistic regression (3<sup>rd</sup> ed.). Hoboken, New Jersey: John Wiley & Sons Inc

# Multivariable, binary outcome

- Let's say:
  - 1 binary outcome (HPT)
  - Prevalence of HPT,  $p = 0.3$
  - 3 numerical independent variables:
    - Age in years, BMI in  $\text{kg}/\text{m}^2$ , weekly physical activity in hours
  - 3 categorical independent variables:
    - Male = yes, no (binary)
    - Smoking = yes, no (binary)
    - Race = Malay, Chinese, Indian (3 levels)
  - 30% dropout
- How many workers should we sample?

# Multivariable, binary outcome

Independent Variables, k	Count
Age (numerical)	1
BMI (numerical)	1
Weekly physical activity (numerical)	1
Male (binary categorical)	1
Smoking (binary categorical)	1
Race (categorical, 3 levels → 2 dummy variables)	2
Total	7

Formula:

$$n_1 = \text{parameters} \times 10$$

$$n_1 = (k + 1) \times 10$$

$$n = n_1 / p$$

1 intercept  
0 interaction

Events,  $n_1$ ?

$$n_1 = (k + 1) \times 10 = (7 + 1) \times 10 = 80$$

Sample size,  $n$ ?

$$n = n_1 / p = 80 / 0.3 = 266.7 \approx 267$$

$$n_{\text{drop}} = n / (1 - \text{dropout proportion})$$

$$= n / (1 - 0.3)$$

$$= 267 / 0.7 = 381.4 \approx 382$$

**Sample size is 382.**

# Multivariable, binary outcome

- We have to sample 334 workers to investigate the associated factors of hypertension among workers in ABC office, taking into account a 20% dropout rate.

# Validation

# Exploratory factor analysis

## (1) Objective:

- Explore internal structure validity / construct validity of a psychometric tool / questionnaire
- "This study aims to explore the internal structure validity of ABC-Q among students."



# Exploratory factor analysis

## (2) Statistical analysis:

- EFA

## (3) Sample size method:

- Rule-of-thumb – 5 respondents / item\*

\*Costello, A. B., & Osborne, J. W. (2005). Best practices in exploratory factor analysis: Four recommendations for getting the most from your analysis. *Practical Assessment Research & Evaluation*, 10(7).

# Exploratory factor analysis

- Let's say a researcher wants to validate a questionnaire with:
  - 40 items (variables)
  - 30% dropout
- How many respondents should he sample?

Formula:

$$n = k \times 5$$

$$n = k \times 5 = 40 \times 5 = 200$$

$$n_{\text{drop}} = n / (1 - \text{dropout proportion}) = 200 / (1 - 0.3) = 287.7 \approx 288$$

Sample size is 288.

# Exploratory factor analysis

- We have to sample 288 students to explore the internal structure validity of ABC-Q among the students, taking into account a 30% dropout rate.

# Confirmatory factor analysis

## (1) Objective:

- Confirm internal structure validity / construct validity of a psychometric tool / questionnaire
- "This study aims to confirm the internal structure validity of ABC-Q among students."

# Confirmatory factor analysis

## (2) Statistical analysis:

- CFA

## (3) Sample size methods:

- Rule-of-thumb – at least 200 respondents\*
- RMSEA, CFI model fit indices – Hypothesis testing

\*Kline, R. B. (2016). Principles and practice of structural equation modeling. 4th ed. New York: Guilford Publications.

# 5. Structural Equation Modeling → RMSEA

## 🏠 » Sample Size Calculator

### Sample Size Calculator (web)

**Structural Equation Modeling - Root Mean Squared Error of Aproximation (RMSEA)**

Expected RMSEA:	<input type="text" value="0.05"/>	
Number of items:	<input type="text"/>	
Number of factors:	<input type="text"/>	
Significance level ( $\alpha$ ):	<input type="text" value="0.05"/>	Two-tailed
Power ( $1 - \beta$ ):	<input type="text" value="80"/>	%
Expected dropout rate:	<input type="text" value="10"/>	%

Please input number of item.  
Please input number of factor.

Usually aim  $< 0.05$   
Can expect based on literature

Total number of items.  
Can be unequal per factor.

# Confirmatory factor analysis

- Let's say a researcher wants to validate a questionnaire with:
  - 2 factors
  - Factor 1: 8 items, Factor 2: 4 items
  - Acceptable RMSEA is set  $< 0.05$ .
  - 10% dropout
- How many respondents should he sample?

# Confirmatory factor analysis

**Structural Equation Modeling - Root Mean Squared Error of Aproximation (RMSEA)**

Expected RMSEA:	<input type="text" value="0.05"/>	
Number of items:	<input type="text" value="12"/>	8 + 4 items
Number of factors:	<input type="text" value="2"/>	2 factors
Significance level ( $\alpha$ ):	<input type="text" value="0.05"/>	Two-tailed
Power ( $1 - \beta$ ):	<input type="text" value="80"/>	%
Expected dropout rate:	<input type="text" value="10"/>	%
<input type="button" value="Reset"/>		
Degree of freedom, df =	<input type="text" value="53"/>	
Sample size, n =	<input type="text" value="235"/>	
Sample size (with 10% dropout), $n_{\text{drop}}$ =	<input type="text" value="262"/>	



# Confirmatory factor analysis

- We have to sample 262 students to confirm the internal structure validity of ABC-Q among the students, taking into account a 10% dropout rate.

# 5. Structural Equation Modeling → CFI

Requires equal number of items per factor

🏠 » **Sample Size Calculator**

## Sample Size Calculator (web)

**Structural Equation Modeling - Comparative Fit Index (CFI)**

Usually aim > 0.95  
Can expect based on literature

Total number of items.  
MUST be equal per factor.

Expected CFI:	<input type="text" value="0.95"/>
Number of items:	<input type="text"/>
Number of factors:	<input type="text"/>
Average factor loading:	<input type="text" value="0.6"/>
Average factor correlation:	<input type="text" value="0.3"/>
Significance level ( $\alpha$ ):	<input type="text" value="0.05"/> Two-tailed
Power ( $1 - \beta$ ):	<input type="text" value="80"/> %
Expected dropout rate:	<input type="text" value="10"/> %

Please input number of item.  
Please input number of factor.

# Confirmatory factor analysis

- Let's say a researcher wants to validate a questionnaire with:
  - 2 factors, 6 items each
  - Based on literature, average factor loading is around 0.7
  - Average factor-factor correlation around 0.3
  - Acceptable CFI  $> 0.95$
  - 10% dropout
- How many respondents should he sample?

# Confirmatory factor analysis

**Structural Equation Modeling - Comparative Fit Index (CFI)**

Expected CFI:	<input type="text" value="0.95"/>	
Number of items:	<input type="text" value="12"/>	6 items x 2 factors MUST be equal
Number of factors:	<input type="text" value="2"/>	2 factors
Average factor loading:	<input type="text" value="0.7"/>	
Average factor correlation:	<input type="text" value="0.3"/>	Av. factor loadings & correlation
Significance level ( $\alpha$ ):	<input type="text" value="0.05"/>	Two-tailed
Power ( $1 - \beta$ ):	<input type="text" value="80"/>	%
Expected dropout rate:	<input type="text" value="10"/>	%
<input type="button" value="Reset"/>		
Degree of freedom, $df_{\text{model}} =$	<input type="text" value="53"/>	
Degree of freedom, $df_{\text{baseline}} =$	<input type="text" value="66"/>	
Sample size, $n =$	<input type="text" value="160"/>	
Sample size (with 10% dropout), $n_{\text{drop}} =$	<input type="text" value="178"/>	

# Confirmatory factor analysis

- We have to sample 178 students to confirm the internal structure validity of ABC-Q among the students, taking into account a 10% dropout rate.

# 5. Structural Equation Modeling → CFI (advanced)

Allows unequal number of items per factor

🏠 » **Sample Size Calculator**

## Sample Size Calculator (web)

**Structural Equation Modeling - Comparative Fit Index (CFI)**

Expected CFI:

Number of items per factor (separated by comma ","):

Average factor loading:

Average factor correlation:

Significance level ( $\alpha$ ):  Two-tailed

Power ( $1 - \beta$ ):  %

Expected dropout rate:  %

Enter valid format.

Usually aim > 0.95  
Can expect based on literature

Allows setting unequal number of item per factor

# Confirmatory factor analysis

- Let's say a researcher wants to validate a questionnaire with:
  - 3 factors
  - Factor 1: 8 items, Factor 2: 4 items, Factor 3: 6 items
  - Based on literature, average factor loadings around 0.7
  - Average factor-factor correlation around 0.3
  - Acceptable CFI  $> 0.95$
  - 10% dropout
- How many respondents should he sample?

# Confirmatory factor analysis

## Structural Equation Modeling - Comparative Fit Index (CFI)

Expected CFI:

Number of items per factor  
(separated by comma ","):

Average factor loading:

Average factor correlation:

Significance level ( $\alpha$ ):  Two-tailed

Power ( $1 - \beta$ ):  %

Expected dropout rate:  %

Reset

Degree of freedom,  $df_{\text{model}} =$

Degree of freedom,  $df_{\text{baseline}} =$

Sample size,  $n =$

Sample size (with 10% dropout),  $n_{\text{drop}} =$

Factor 1: 8 items  
Factor 2: 4 items  
Factor 3: 6 items

Av. factor loadings  
& correlation



# Confirmatory factor analysis

- We have to sample 180 students to confirm the internal structure validity of ABC-Q among the students, taking into account a 10% dropout rate.

# Structural Equation Modeling

## (1) Objective:

- Test a hypothesized structural model for theory ABC
- "This study aims to test the hypothesized ABC model among students by SEM."
- "This study aims to examine/test the hypothesized structural relationships between X, Y, and Z among students."

# Structural Equation Modeling

## (2) Statistical analysis:

- SEM

## (3) Sample size methods:

- Rule-of-thumb – at least 200 respondents\*
- RMSEA model fit index – Hypothesis testing

\*Kline, R. B. (2016). Principles and practice of structural equation modeling. 4th ed. New York: Guilford Publications.

# 5. Structural Equation Modeling → RMSEA

## 🏠 >> Sample Size Calculator

### Sample Size Calculator (web)

**Structural Equation Modeling - Root Mean Squared Error of Aproximation (RMSEA)**

Expected RMSEA:

Degrees of freedom:

Significance level ( $\alpha$ ):  Two-tailed

Power ( $1 - \beta$ ):  %

Expected dropout rate:  %

Sample size,  $n =$

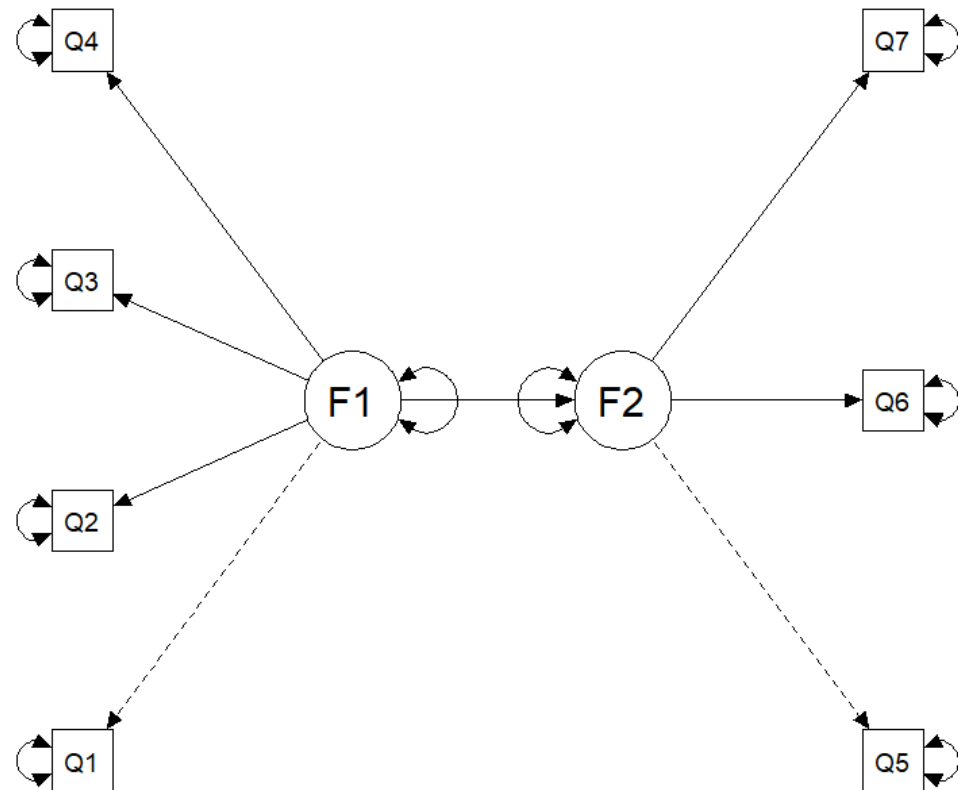
Sample size (with 10% dropout),  $n_{\text{drop}} =$

Usually aim  $< 0.05$   
Can expect based on literature

$df$ , must calculate for SEM model

# Structural Equation Modeling

- Let's say a researcher wants to validate a the provided structural model with:
  - Acceptable RMSEA is set  $< 0.05$
  - 10% dropout
- How many respondents should he sample?



# Structural Equation Modeling

Calculate *df* manually:

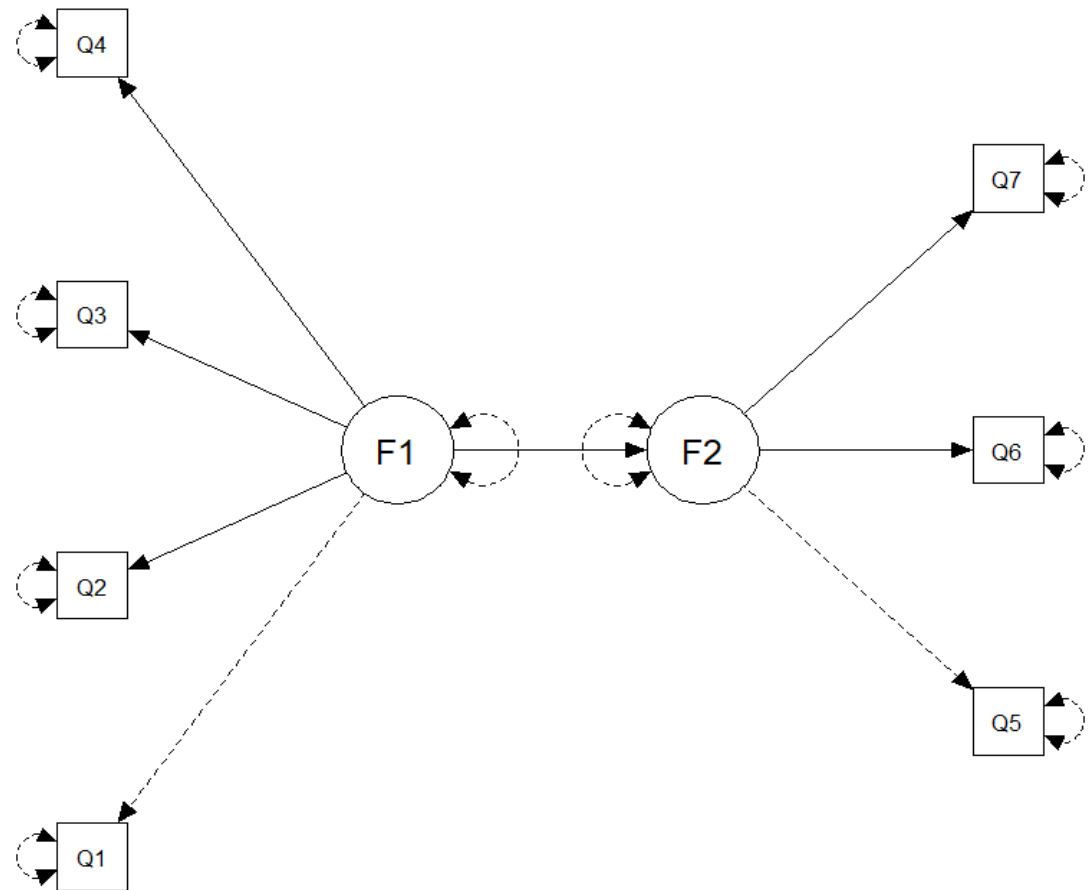
Estimated parameters	Count
Factor loadings	5
Error variances	8
Factor variances	1
Factor covariances	0
Regressions	1
Total (a)	15

$a$  = number of estimated parameters

$p$  = number of items

$b = p(p + 1)/2 = 7(7 + 1)/2 = 28$

$df = b - a = 28 - 15 = 13$



# Structural Equation Modeling

Calculate  $df$  in R:

R code

```
library(lavaan)
library(semPlot)

model1 = "
F1 =~ 1*Q1 + Q2 + Q3 + Q4
F2 =~ 1*Q5 + Q6 + Q7
F2 ~ F1
"

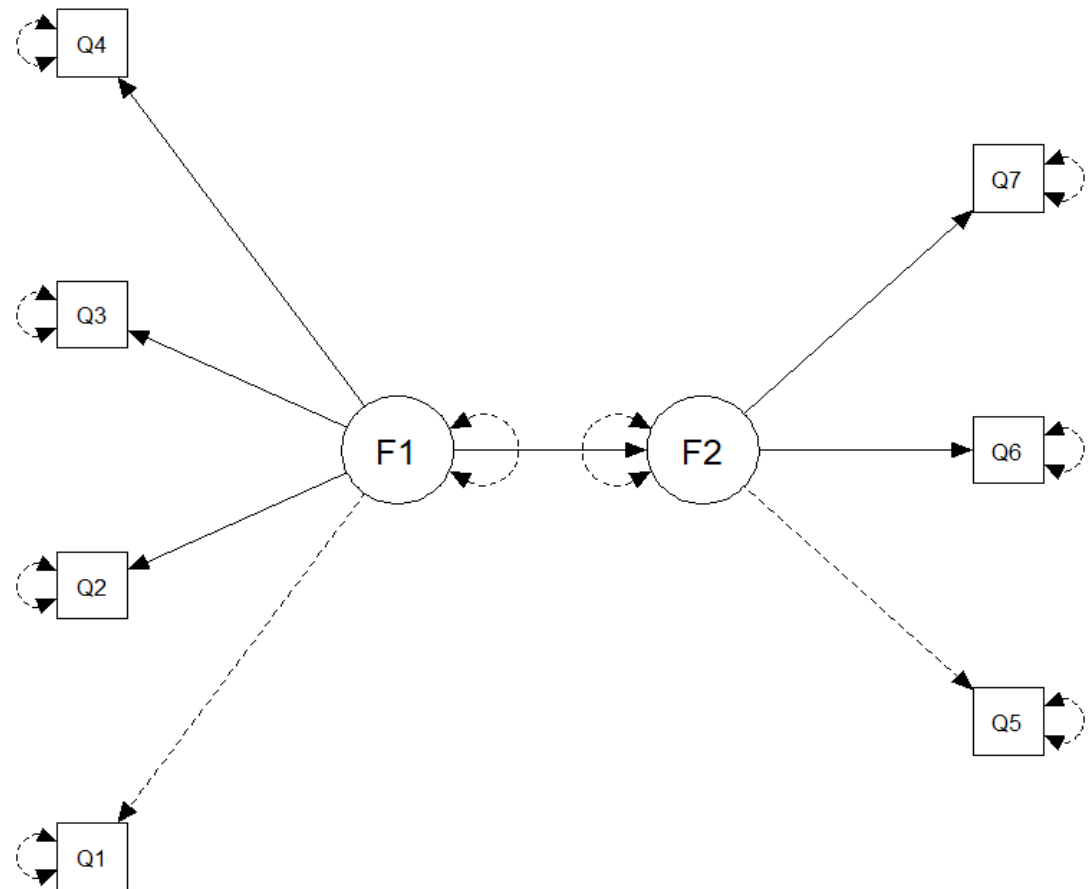
data = simulateData(model1)
fit.model1 = sem(model1, data = data)
fitmeasures(fit.model1)["df"]
```

R output

```
df
13
```

R plot

```
semPaths(lavaanify(model1), what="path",
         whatLabels="name",
         edge.color = "black",
         layout = "tree2", rotation = 2)
```



# Structural Equation Modeling

## Structural Equation Modeling - Root Mean Squared Error of Aproximation (RMSEA)

Expected RMSEA:

0.05

Calculated df

Degrees of freedom:

13

Significance level ( $\alpha$ ):

0.05

Two-tailed

Power ( $1 - \beta$ ):

80

%

Expected dropout rate:

10

%

Calculate

Reset

Sample size,  $n =$

551

Sample size (with 10% dropout),  $n_{\text{drop}} =$

613



# Structural Equation Modeling

- We have to sample 613 students to confirm the proposed structural model among the students, taking into account a 10% dropout rate.

# Cronbach's alpha

## (1) Objective:

- Determine reliability of a psychometric tool / questionnaire
- "This study aims to determine the reliability of ABC-Q among students."

# Cronbach's alpha

(2) Statistical analysis:

- Cronbach's alpha coefficient

(3) Sample size method:

- Cronbach's alpha coefficient – Hypothesis testing

# 4. Reliability → Cronbach's alpha coefficient

## 🏠 >> Sample Size Calculator

### Sample Size Calculator (web)

**Cronbach's alpha - Hypothesis Testing**

Minimum acceptable Cronbach's alpha ( $H_0$ ):

Expected Cronbach's alpha ( $H_1$ ):

Significance level ( $\alpha$ ):  Two-tailed

Power ( $1 - \beta$ ):  %

Number of items (k):

Expected dropout rate:  %

Sample size, n =

Sample size (with 10% dropout),  $n_{\text{drop}}$  =

Usually 0.65, 0.7

Expected, can be estimated from literature

# Cronbach's alpha

- Let's say a researcher wants to determine reliability of a questionnaire with:
  - Minimum acceptable Cronbach's alpha = 0.7
  - Expected Cronbach's alpha = 0.85
  - 2 factors
  - Factor 1: 8 items, Factor 2: 6 items
  - 20% dropout
- How many respondents should he sample?
  - Calculate for each factor, and take the largest.

# Cronbach's alpha

🏠 >> Sample Size Calculator

## Sample Size Calculator (web)

**Cronbach's alpha - Hypothesis Testing**

Minimum acceptable Cronbach's alpha ( $H_0$ ):	<input type="text" value="0.7"/>	
Expected Cronbach's alpha ( $H_1$ ):	<input type="text" value="0.85"/>	Expected, can be from literature
Significance level ( $\alpha$ ):	<input type="text" value="0.05"/>	Two-tailed
Power ( $1 - \beta$ ):	<input type="text" value="80"/>	%
Number of items (k):	<input type="text" value="8"/>	8 items from Factor 1
Expected dropout rate:	<input type="text" value="20"/>	%

Sample size,  $n =$

Sample size (with 20% dropout),  $n_{\text{drop}} =$

# Cronbach's alpha

🏠 >> Sample Size Calculator

## Sample Size Calculator (web)

**Cronbach's alpha - Hypothesis Testing**

Minimum acceptable Cronbach's alpha ( $H_0$ ):	<input type="text" value="0.7"/>	
Expected Cronbach's alpha ( $H_1$ ):	<input type="text" value="0.85"/>	Expected, can be from literature
Significance level ( $\alpha$ ):	<input type="text" value="0.05"/>	Two-tailed
Power ( $1 - \beta$ ):	<input type="text" value="80"/>	%
Number of items (k):	<input type="text" value="6"/>	6 items from Factor 2
Expected dropout rate:	<input type="text" value="20"/>	%

Sample size,  $n =$

Sample size (with 20% dropout),  $n_{\text{drop}} =$

# Cronbach's alpha

- We have to sample 53 students to determine the reliability of ABC-Q among the students, taking into account a 20% dropout rate.



# Intraclass correlation

## (1) Objective:

- Determine test-retest reliability / interrater reliability of a numerical measure, i.e. consistency / agreement.
- "This study aims to determine the test-retest reliability of ABC-Q among students."
- "This study aims to determine the interrater reliability of new doctors in measuring blood pressure."

# Intraclass correlation

(2) Statistical analysis:

- Intraclass correlation coefficient (ICC)

(3) Sample size method:

- Intraclass Correlation Coefficient (ICC) - Hypothesis Testing

# 4. Reliability → Intraclass correlation coefficient (ICC) → ICC - Hypothesis Testing

🏠 » Sample Size Calculator

## Sample Size Calculator (web)

**Intraclass Correlation Coefficient (ICC) - Hypothesis Testing<sup>1</sup>**

Minimum acceptable reliability (ICC) ( $\rho_0$ ):

Expected reliability (ICC) ( $\rho_1$ ):

Significance level ( $\alpha$ ):  Two-tailed

Power ( $1 - \beta$ ):  %

Number of raters/repetitions per subject (k):

Expected dropout rate:  %

Sample size, n =

Sample size (with 10% dropout),  $n_{\text{drop}}$  =

Between 0.6 to 0.74

Expected, can be estimated from literature

Test-retest = 2 / Any number of raters

# Intraclass correlation - test-retest

- Let's say for test-retest reliability:
  - Minimum acceptable ICC = 0.7
  - Expected ICC = 0.85
  - Two occasions
  - 10% dropout
- How many respondents should he sample?

# Intraclass correlation

## 🏠 » Sample Size Calculator

### Sample Size Calculator (web)

**Intraclass Correlation Coefficient (ICC) - Hypothesis Testing<sup>1</sup>**

Minimum acceptable reliability (ICC) ( $\rho_0$ ):	<input type="text" value="0.7"/>	
Expected reliability (ICC) ( $\rho_1$ ):	<input type="text" value="0.85"/>	Expected, can be from literature
Significance level ( $\alpha$ ):	<input type="text" value="0.05"/>	Two-tailed
Power ( $1 - \beta$ ):	<input type="text" value="80"/>	% 2 repetitions for test-retest
Number of raters/repetitions per subject (k):	<input type="text" value="2"/>	
Expected dropout rate:	<input type="text" value="10"/>	%

Sample size, n =

Sample size (with 10% dropout),  $n_{\text{drop}}$  =

# Intraclass correlation

- We have to sample 59 students to determine the test-retest reliability of ABC-Q among the students, taking into account a 10% dropout rate.

# Intraclass correlation - interrater

- Let's say for interrater reliability:
  - Minimum acceptable ICC = 0.8 (set higher, measuring blood pressure correctly is a serious matter)
  - Expected ICC = 0.9
  - 5 doctors
  - 10% dropout
- How many doctors should he sample?

# Intraclass correlation

## 🏠 >> Sample Size Calculator

### Sample Size Calculator (web)

**Intraclass Correlation Coefficient (ICC) - Hypothesis Testing<sup>1</sup>**

Minimum acceptable reliability (ICC) ( $\rho_0$ ):  0.8 ▾

Expected reliability (ICC) ( $\rho_1$ ):  0.9 ▾

Significance level ( $\alpha$ ):  0.05 ▾ Two-tailed

Power ( $1 - \beta$ ):  80 ▾ %

Number of raters/repetitions per subject (k):  5 ▾

Expected dropout rate:  10 ▾ %

Sample size, n =  33 ▾

Sample size (with 10% dropout),  $n_{\text{drop}}$  =  37 ▾

Annotations:

- Set higher at 0.8 (pointing to the 0.8 input)
- Expected, can be from literature (pointing to the 0.9 input)
- 5 doctors (pointing to the 5 input)



# Intraclass correlation

- We have to sample 37 patients to determine the interrater reliability of 5 doctors in measuring blood pressure, taking into account a 10% dropout rate.

# Kappa agreement

## (1) Objective:

- Determine interrater reliability of a categorical measure, i.e. consistency / agreement.
- "This study aims to determine the interrater reliability of two radiologists in rating X-ray films for the presence of pneumonia among patients with shortness of breath."
- Rater = human rater / diagnostic test

# Kappa agreement

(2) Statistical analysis:

- Kappa coefficient

(3) Sample size method:

- Kappa (2 raters) – Hypothesis testing

# 4. Reliability → Kappa coefficient → Kappa (2 raters) - Hypothesis Testing

🏠 >> Sample Size Calculator

## Sample Size Calculator (web)

**Kappa (2 raters) - Hypothesis Testing<sup>1</sup>**

Minimum acceptable kappa ( $\kappa_0$ ):

Expected kappa ( $\kappa_1$ ):

Proportion of outcome (p), e.g. p of heart disease:

Significance level ( $\alpha$ ):  Two-tailed

Power (1 -  $\beta$ ):  %

Expected dropout rate:  %

Sample size, n =

Sample size (with 10% dropout),  $n_{\text{drop}}$  =

Usually 0.6,  
Moderate agreement

Expected, can be  
estimated  
from literature

# Kappa agreement

- Let's say a researcher wants to determine the agreement between two radiologists:
  - Minimum acceptable kappa = 0.6
  - Expected kappa = 0.9
  - Presence of pneumonia among patients with shortness of breath is 30% ( $p = 0.3$ )
  - 30% dropout
- How many patients should be sampled?

# Kappa agreement

## 🏠 >> Sample Size Calculator

### Sample Size Calculator (web)

**Kappa (2 raters) - Hypothesis Testing<sup>1</sup>**

Minimum acceptable kappa ( $\kappa_0$ ):	<input type="text" value="0.6"/>	
Expected kappa ( $\kappa_1$ ):	<input type="text" value="0.9"/>	Expected, can be from literature
Proportion of outcome (p), e.g. p of heart disease:	<input type="text" value="0.3"/>	30% presence of pneumonia $p = 0.3$
Significance level ( $\alpha$ ):	<input type="text" value="0.05"/>	Two-tailed
Power (1 - $\beta$ ):	<input type="text" value="80"/>	%
Expected dropout rate:	<input type="text" value="30"/>	%
<input type="button" value="Calculate"/> <input type="button" value="Reset"/>		
Sample size, n =	<input type="text" value="66"/>	
Sample size (with 30% dropout), $n_{\text{drop}}$ =	<input type="text" value="95"/>	

# Kappa agreement

- We have to sample 95 X-ray films of patients with shortness of breath to determine the interrater reliability of two radiologists in rating X-ray films for the presence of pneumonia, taking into account a 30% dropout rate.

# Sensitivity & Specificity

## (1) Objective:

- Estimate sensitivity and specificity of a diagnostic test, e.g. RTK-Ag for COVID-19
- "This study aims to determine sensitivity and specificity of a new COVID-19 self-test kit among exposed adults."



# Sensitivity & Specificity

(2) Statistical analysis:

- Proportion, 95% Confidence Interval (CI)

(3) Sample size method:

- Sensitivity/Specificity – Estimation

# 2. Proportions → Sensitivity and specificity → Sensitivity/Specificity - Estimation

🏠 » Sample Size Calculator

Sample Size Calculator (web)

## Sensitivity/Specificity - Estimation

Expected sensitivity:	<input type="text" value="0.90"/>
Expected specificity:	<input type="text" value="0.85"/>
Prevalence of disease (p):	<input type="text" value="0.20"/>
Precision ( $\pm$ expected):	<input type="text" value="0.10"/>
Confidence level 100(1 - $\alpha$ ):	<input type="text" value="95"/> %
Expected dropout rate:	<input type="text" value="10"/> %
<input type="button" value="Calculate"/> <input type="button" value="Reset"/>	
Sample size for sensitivity, $n_{sen}$ =	<input type="text"/>
Sample size for specificity, $n_{spec}$ =	<input type="text"/>
Final sample size (largest), $n$ =	<input type="text"/>
Final sample size (with 10% dropout), $n_{drop}$ =	<input type="text"/>

Expected sensitivity, can be from literature

Expected specificity, can be from literature

Prevalence of disease

Precision, in proportion

Confidence level 90%, **95%**, 99%

% of participants that may dropout

# Sensitivity & Specificity

- Let say:
  - Expected Sensitivity = 0.9
  - Expected Specificity = 0.7
  - Prevalence of COVID-19\* = 5% = 0.05 (in proportion)
  - Precision =  $\pm 5\%$  = 0.05 (in proportion)
  - 95% Confidence level
  - 10% dropout
- How many exposed adults should we sample?

\*based on positivity rate

# Sensitivity & Specificity

🏠 » Sample Size Calculator

Sample Size Calculator (web)

**Sensitivity/Specificity - Estimation**

Expected sensitivity:	<input type="text" value="0.90"/>
Expected specificity:	<input type="text" value="0.7"/>
Prevalence of disease (p):	<input type="text" value="0.05"/>
Precision ( $\pm$ expected):	<input type="text" value="0.05"/>
Confidence level 100(1 - $\alpha$ ):	<input type="text" value="95"/> %
Expected dropout rate:	<input type="text" value="10"/> %

Sample size for sensitivity, $n_{sen}$ =	<input type="text" value="2766"/>
Sample size for specificity, $n_{spec}$ =	<input type="text" value="340"/>
Final sample size (largest), $n$ =	<input type="text" value="2766"/>
Final sample size (with 10% dropout), $n_{drop}$ =	<input type="text" value="3074"/>

Take the largest of the two

# Sensitivity & Specificity

- We have to sample 3075 patients to estimate both sensitivity and specificity of the new test kit among exposed adults, taking into account a 10% dropout rate.

# References

- Arifin, W. N. (2013). *Introduction to sample size calculation. Education in Medicine Journal, 5(2), e89-e96.*
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