

Lab 05: Power Calculations

Step by Step guidance

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Overview of Lab 05

Power

In the following slides AND IN THE HANDOUT, you can find step by step guidance on using G*Power to calculate your Power - and therefore the number of participants you would be required to test.

Remember, it's not life and death if you don't meet sample size this year, but it would be next year.

You will have potential access to the RPS next year.

You will be asked to take part in all available Mini-Dissertation studies this year in weeks 11-15

I got the, I got the power!

The formal definition of a p value is the probability of observing a result at least as extreme as the one observed, assuming the null hypothesis is true (e.g. Cohen, 1994).

Cohen, J (1988). Statistical power analysis for the behavioral sciences (2nd ed.). Lawrence Erlbaum Associates. I wouldn't bother.

What?

This means, and assuming we have a crystal ball to know the 'truth' of whether there is an effect of our manipulation..

- a small p value indicates the results are surprising if the null hypothesis IS true
- a large p value indicates the results are not very surprising if the null IS true.

G*Power

You can download G*Power on this page. Under the heading "download" click on the appropriate version for whether you have a Windows or Mac computer.

<https://www.psychologie.hhu.de/arbeitsgruppen/allgemeine-psychologie-und-arbeitspsychologie/gpower>

2x2 factorial ANOVA Power analysis

In a 2x2 design, there are three effects to compare:

- the main effect of IV1,
- the main effect of IV2,
- and the interaction between IV1 and IV2.

Look after the little effect first

There are three comparisons to make here and it is unlikely you would expect the same effect size for all three comparisons, meaning you must ensure all three effects are sufficiently powered by a single experimental design. This just means that your sample size will need to be sufficient to detect the weakest effect size of the three.

You might expect the effect size (or difference) for IV1 to be a lot larger than the effect size (or difference) for IV2, for example.

This means you would have to ensure the smallest effect is sufficiently powered, in the above example, the IV2 effect size would drive the sample size calculation.

if the smallest effect is covered, by virtue of the fact power exists along a curve - then the larger effects would have sufficient power!

G*Power guide

If you are running a design with 2 between participant IVs, use Slide 1

If you are running a mixed design, and want to explore the between Main effect - slide 2

If you are running a mixed design, and want to explore the within Main effect - slide 3

If you are running a mixed design, and want to explore the interaction effect - slide 4

If you are running a design with 2 between participant IVs, use Slide 5

G*Power guide - Key info

The following slides give you the specifics of what you need to include depending on the effect you are exploring. The only thing you need is the effect size of interest.

This is usually gleaned from your target paper (or other literature).

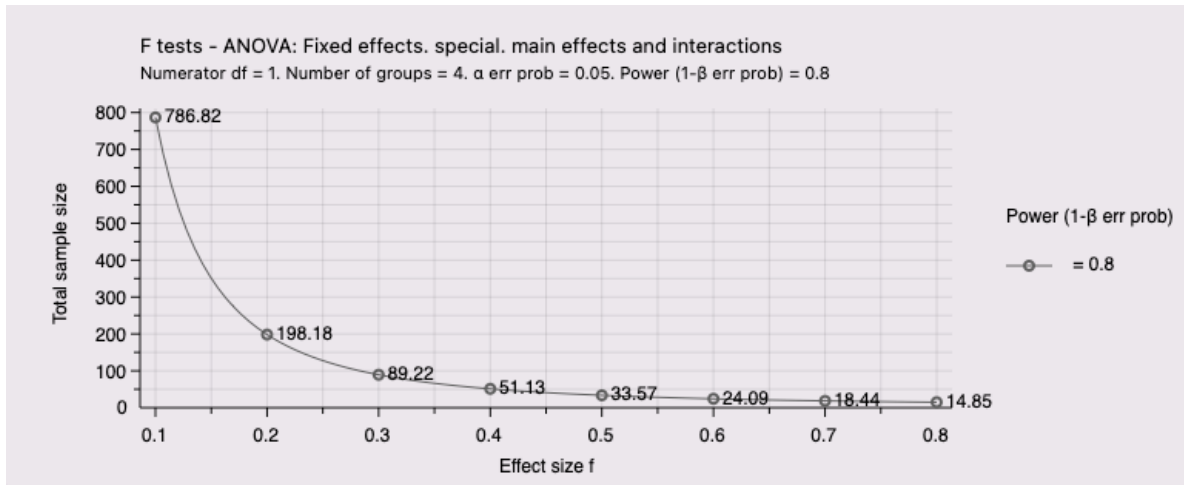
You will need to examine each of your effect sizes in your groups and cover the weakest with the the sample size, so the effect that requires the largest sample size will give the group sample size.

If you cannot find an effect, you may use a 0.3 f effect size, but this is to be used only when no alternative is available.

1 - Fully Between Designs (any effect of interest)

Test family		Statistical test																					
F tests		ANOVA: Fixed effects, special, main effects and interactions																					
Type of power analysis																							
A priori: Compute required sample size - given α , power, and effect size																							
Input parameters		Output parameters																					
Determine <table border="1"> <tr> <td>Effect size f</td> <td>0.3</td> </tr> <tr> <td>α err prob</td> <td>0.05</td> </tr> <tr> <td>Power (1-β err prob)</td> <td>0.8</td> </tr> <tr> <td>Numerator df</td> <td>1</td> </tr> <tr> <td>Number of groups</td> <td>4</td> </tr> </table>		Effect size f	0.3	α err prob	0.05	Power (1- β err prob)	0.8	Numerator df	1	Number of groups	4	<table border="1"> <tr> <td>Noncentrality parameter λ</td> <td>8.1000000</td> </tr> <tr> <td>Critical F</td> <td>3.9518824</td> </tr> <tr> <td>Denominator df</td> <td>86</td> </tr> <tr> <td>Total sample size</td> <td>90</td> </tr> <tr> <td>Actual power</td> <td>0.8034953</td> </tr> </table>		Noncentrality parameter λ	8.1000000	Critical F	3.9518824	Denominator df	86	Total sample size	90	Actual power	0.8034953
Effect size f	0.3																						
α err prob	0.05																						
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Denominator df	86																						
Total sample size	90																						
Actual power	0.8034953																						

Power plot



2 - Mixed design (looking at a between main effect)

Test family: F tests

Statistical test: ANOVA: Repeated measures, between factors

Type of power analysis: A priori: Compute required sample size - given α , power, and effect size

Input parameters:

Determine

Effect size f: 0.3

α err prob: 0.05

Power ($1-\beta$ err prob): 0.8

Number of groups: 2

Number of measurements: 2

Corr among rep measures: 0.5

Output parameters:

Noncentrality parameter λ : 8.1600000

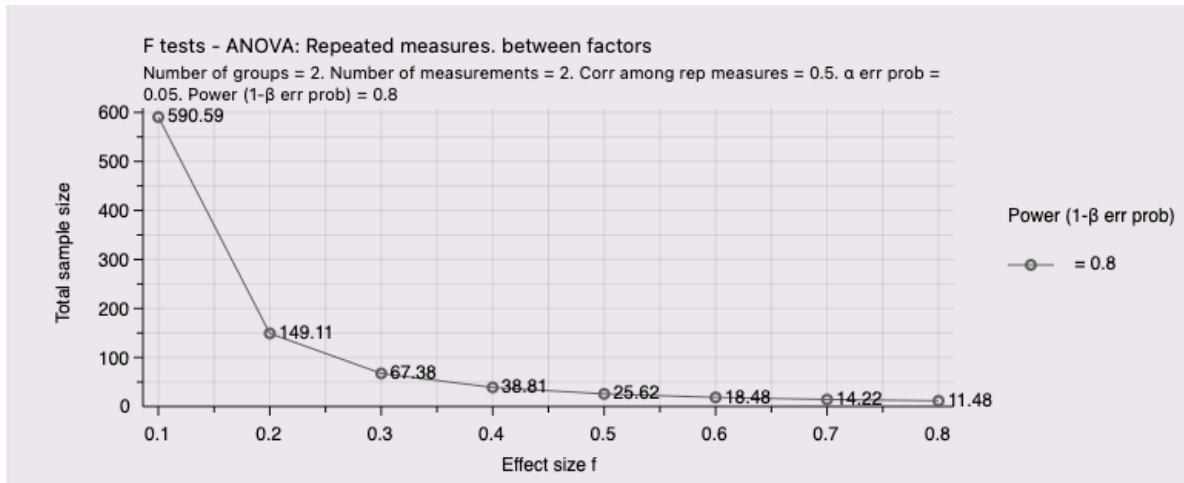
Critical F: 3.9862695

Numerator df: 1.0000000

Denominator df: 66.0000000

Total sample size: 68

Actual power: 0.8036734

graph**3 - Mixed design (looking at within main effect)**

Test family: F tests

Statistical test: ANOVA: Repeated measures, within factors

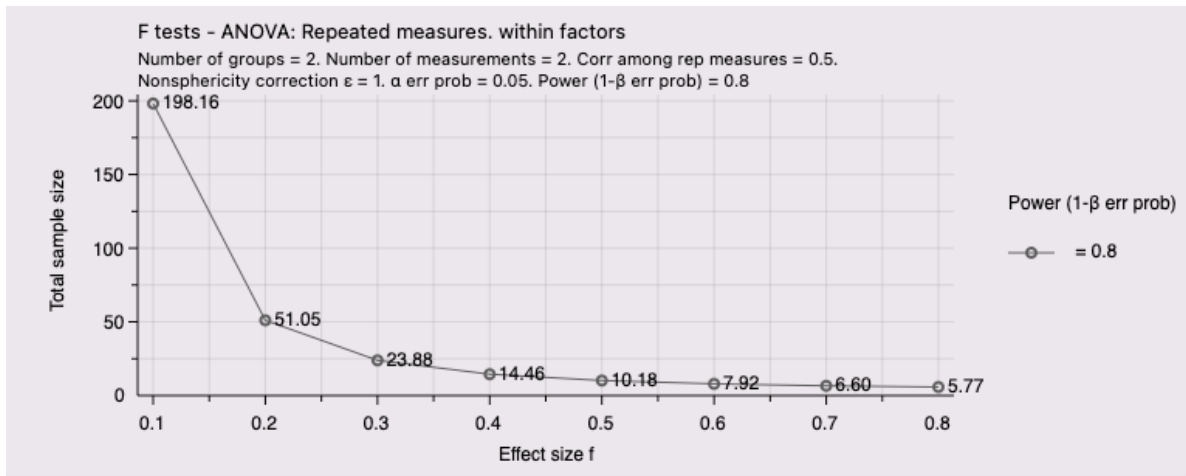
Type of power analysis: A priori: Compute required sample size - given α , power, and effect size

Input parameters:

- Determine
- Effect size f: 0.3
- α err prob: 0.05
- Power ($1-\beta$ err prob): 0.8
- Number of groups: 2
- Number of measurements: 2
- Corr among rep measures: 0.5
- Nonsphericity correction ϵ : 1

Output parameters:

- Noncentrality parameter λ : 8.6400000
- Critical F: 4.3009495
- Numerator df: 1.0000000
- Denominator df: 22.0000000
- Total sample size: 24
- Actual power: 0.8020788



4 - Mixed design (looking at interaction effect)

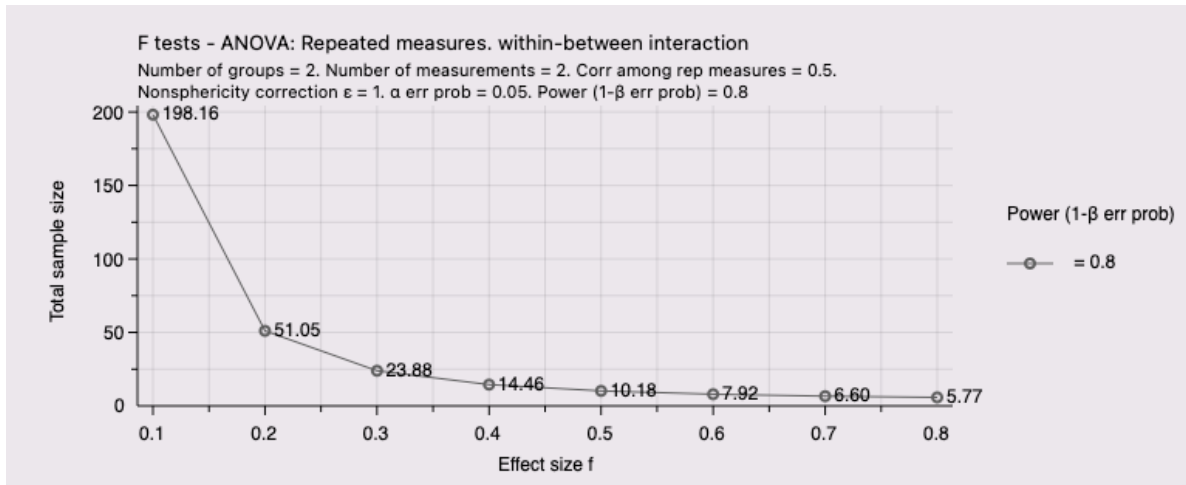
Test family		Statistical test	
F tests		ANOVA: Repeated measures, within-between interaction	

Type of power analysis

A priori: Compute required sample size - given α , power, and effect size

Input parameters		Output parameters	
Determine	Effect size f	0.3	Noncentrality parameter λ
	α err prob	0.05	Critical F
	Power ($1-\beta$ err prob)	0.8	Numerator df
	Number of groups	2	Denominator df
	Number of measurements	2	Total sample size
	Corr among rep measures	0.5	Actual power
	Nonsphericity correction ϵ	1	

graph



5 - Repeated Measures ANOVA (any effect)

Test family: F tests

Statistical test: ANOVA: Repeated measures, within factors

Type of power analysis: A priori: Compute required sample size - given α , power, and effect size

Input parameters:

Determine

Effect size f: 0.3

α err prob: 0.05

Power ($1-\beta$ err prob): 0.8

Number of groups: 1

Number of measurements: 4

Corr among rep measures: 0.5

Nonsphericity correction ϵ : 1

Output parameters:

Noncentrality parameter λ : 12.2400000

Critical F: 2.7980606

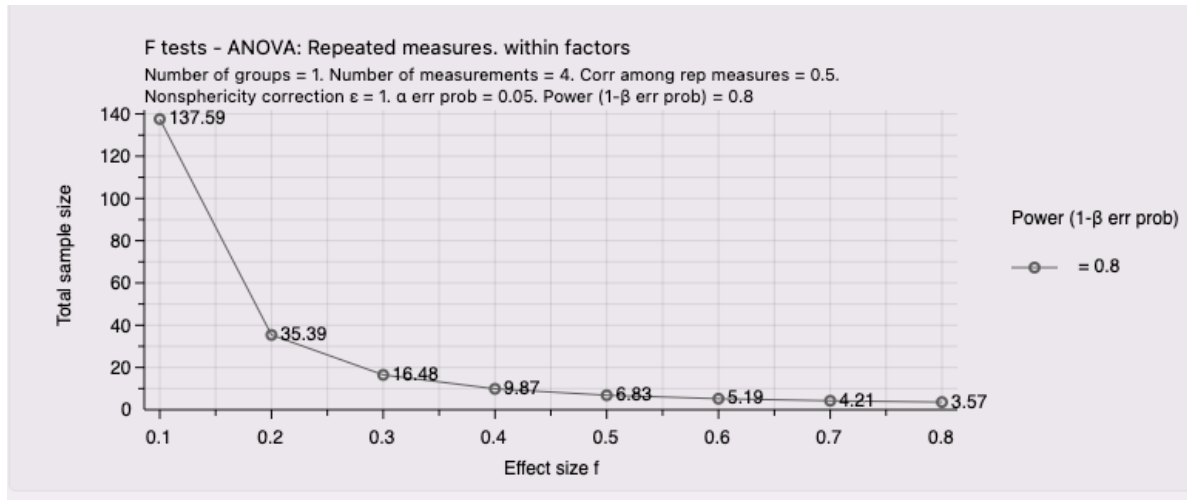
Numerator df: 3.0000000

Denominator df: 48.0000000

Total sample size: 17

Actual power: 0.8146644

Graph



Writing up

“In order to detect an effect size of Cohen’s $d = 0.xx$ with 80% power ($\alpha = .05$, two-tailed), G*Power analysis determines a sample size of XX participants in a *insert type 2x2 ANOVA*”. The smallest effect size of interest was set to $d = 0.xx$ based on the meta-analysis by Dougal et al. (2023).”