

# Resources

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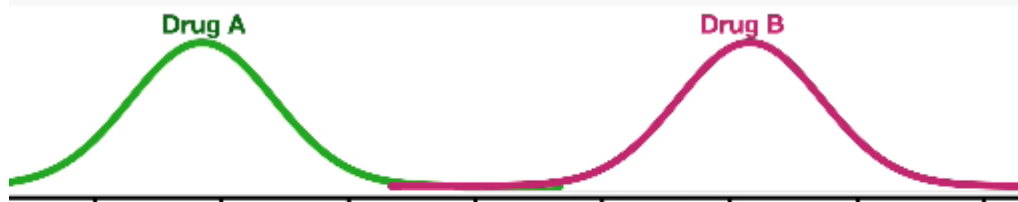
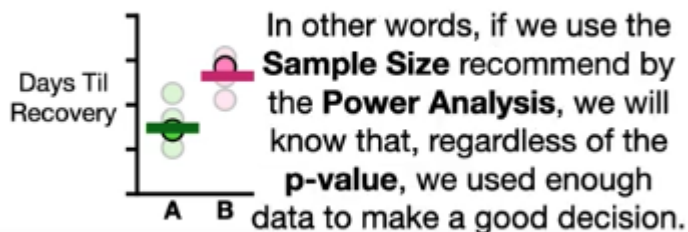
- [statquest: Power Analysis, Clearly Explained!!!](#)

## Power Analysis

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Let's say we have two drugs A and B with 5 samples each. The t-test for means gives us  $p\text{-value} = 0.06$ . But the y-axis (recovery days) for drug A looks smaller and we want to add more points to get lower p-value. This is p-hacking, we should not do this. Instead, we should do power analysis to find the large enough sample size. The power gives, regardless of p-value, we have enough information to make a good decision.

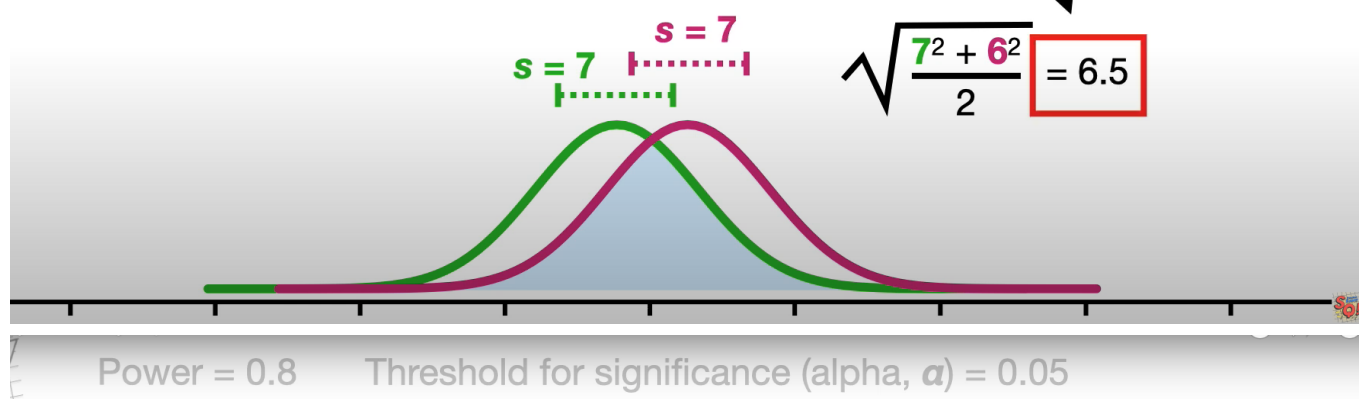
Power is affected by: a) Overlap between distributions b) Sample size.



For example, if I want to have **Power = 0.8**, meaning, I want to have at least an **80%** chance of **correctly** rejecting the **Null Hypothesis...**

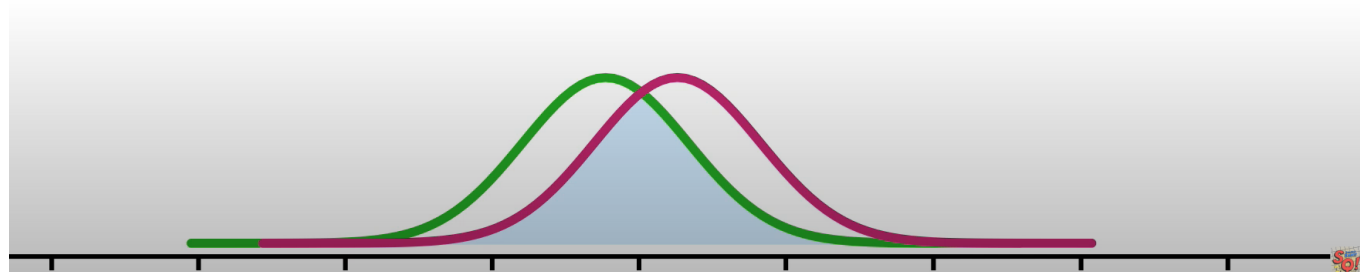
$$\text{Effect Size (d)} = \frac{\text{The estimated difference in the means}}{\text{Pooled estimated standard deviations}} = \frac{10}{6.5}$$

...and we get 6.5.



$$\text{Effect Size (d)} = \frac{\text{The estimated difference in the means}}{\text{Pooled estimated standard deviations}}$$

However, in general, the mean and standard deviations can be estimated with prior data, a literature search, or, in a worst case scenario, an educated guess.



If we have more-overlap, we need more samples to have more power. Even when the distributions overlap, the sample means become separated if we have large enough sample size.

## Statistical Power

This is the probability of rejecting null hypothesis when alternative hypothesis is true.

Power = 1 - Type II Error

Higher the statistical power, lower is the probability of making type II error.

If statistic power is low: Large risk of committing type II error. (FN)

If statistic power is large: Small risk of type II error.

NOTE: Type I error: Reject  $H_0$  when  $H_0$  is true. FP. p-value is optimistically low.

Type II error: Fail to reject  $H_0$  when  $H_1$  is true. False NEGATIVE.

## Sample size

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Usually we use power analysis to get the sample size required for a study.

```
# estimate sample size via power analysis
from statsmodels.stats.power import TTestIndPower
# parameters for power analysis
effect_size_d = 0.8
alpha = 0.05
power = 0.8

# perform power analysis
analysis = TTestIndPower()
sample_size = analysis.solve_power(effect_size_d, power=power,
                                   nobs1=None, ratio=1.0, alpha=alpha)
print('Sample Size: %d' % int(sample_size)) # Sample Size: 25
```

An effect size refers to the size or magnitude of an effect or result as it would be expected to occur in a population.

The effect size is estimated from samples of data.

Usually effect size is taken as Pearson Correlation  $r$  or cohen's  $d$  measurement.