# Power and sample sizes <br> math612.0 A1: From numbers through algebra to calculus and linear algebra 

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## The power of a test

Suppose we have a method to test a null hypothesis against an alternative hypothesis. The test would be "controlled" at some level $\alpha$, i.e. $P$ reject $\left.H_{0}\right] \leq \alpha$ whenever $H_{0}$ is true.

On the other hand, when $H_{0}$ is false one wants $P\left[\right.$ reject $\left.H_{0}\right]$ to be as high as possible.

If the parameter to be tested is $\theta$ and $\theta_{0}$ is a value within $H_{0}$ and $\theta_{a}$ is in $H_{a}$ then we want $P_{\theta_{0}}$ [reject $\left.H_{0}\right] \leq \alpha$ and $P_{\theta_{a}}$ [reject $\left.H_{0}\right]$ as large as possible.

For a general $\theta$ we write

$$
\beta(\theta)=P_{\theta}\left[r e j e c t ~ H_{0}\right]
$$

for the power of the test

## The power of tests for proportions



## The Power of the one sided $z$ test for the mean

The one sided z-test for the mean $(\mu)$ is based on a random sample where $X_{1} \ldots X_{n} \sim n\left(\mu, \sigma^{2}\right)$ are independent and $\sigma^{2}$ is known.

The power of the test for an arbitrary $\mu$ can be computed as:

$$
\beta(\mu)=1-\Phi\left(\frac{\mu_{0}-\mu}{\frac{\sigma}{\sqrt{n}}}+z_{1-\alpha}\right)
$$

## Power and sample size for the one-sided $z$-test for a single normal mean

Suppose we want to test $H_{0}: \mu=\mu_{0}$ vs $H_{a}: \mu>\mu_{0}$. We will reject $H_{0}$ if the observed value

$$
z=\frac{\bar{x}-\mu_{0}}{\sigma / \sqrt{n}}
$$

is such that $z>z_{1-\alpha}$.

## The non central t - distribution

Recall that if $Z \sim n(0,1)$ and $U \sim \chi^{2}{ }_{v}$ are independent then

$$
\frac{Z}{\sqrt{\frac{u}{v}}} \sim t_{v}
$$

and it follows for a random sample $X_{1} \ldots X_{n} \sim n\left(\mu, \sigma^{2}\right)$ independent; that

$$
\frac{\bar{X}-\mu}{\frac{s}{\sqrt{n}}}=\frac{\frac{\bar{X}-\mu}{\frac{\sigma}{\sqrt{n}}}}{\sqrt{\frac{\sum\left(X_{i}-\bar{X}\right)^{2}}{\frac{\sigma^{2}}{n-1}}}} \sim t_{n-1}
$$

## The power of t-test for a normal mean (warning: errors)

## Power and sample size for the one sided t-test for a mean

Suppose we want to calculate the power of a one sided t-test for a single mean (one sample), this can easily be done in $R$ with the power.t.test command.

## The power of the 2-sided t-test

A power analysis on a two-sided t-test can be done in R using the power.t.test command.

## The power of the 2-sample one and two-sided t-tests

The power of a two sample, one-sided t-test can be computed as follows:

$$
\beta_{\left(\mu_{1} \mu_{2}\right)}=P_{\mu_{1} \mu_{2}}\left[\frac{Z+\Delta}{\sqrt{U /(n+m-2)}}>t_{1-\alpha, n+m-2}^{*}\right]
$$

and the power of a two sample, two-sided t-test is give by:

$$
\beta_{\left(\mu_{1} \mu_{2}\right)}=P_{\mu_{1} \mu_{2}}\left[\frac{Z+\Delta}{\sqrt{U /(n+m-2)}}>t_{1-\alpha, n+m-2}^{*}\right]+P_{\mu_{1} \mu_{2}}\left[\frac{Z+\Delta}{\sqrt{U /(n+m-}}\right.
$$

where $\Delta=\frac{\left(\mu_{1}-\mu_{2}\right)}{\sigma \sqrt{\frac{1}{n}+\frac{1}{m}}}$ and $U$ is the SSE.

## Sample sizes for two-sample one and two-sided t-tests

The sample size should always satisfy the desired power.

## A case study in power

Want to compute power in analysis of covariance:

$$
y_{i j}=\mu_{i}+\beta x_{i j}+\epsilon_{i j}, \quad i=1,2, j=1, \ldots J
$$

where $\epsilon_{i j} \sim n\left(0, \sigma^{2}\right)$ are i.i.d.?
This can be done by simulation and can easily be expanded to other cases.

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