

Power

Statistical Inference

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Power

- · Power is the probability of rejecting the null hypothesis when it is false
- Ergo, power (as it's name would suggest) is a good thing; you want more power
- A type II error (a bad thing, as its name would suggest) is failing to reject the null hypothesis when it's false; the probability of a type II error is usually called β
- $\cdot \ \ \text{Note Power} = 1 \beta$

Notes

- · Consider our previous example involving RDI
- $H_0: \mu=30$ versus $H_a: \mu>30$
- · Then power is

$$Pigg(rac{ar{X}-30}{s/\sqrt{n}}>t_{1-lpha,n-1}\mid \mu=oldsymbol{\mu_a}$$
 also/H1 ist true

- · Note that this is a function that depends on the specific value of $\mu_a!$
- Notice as μ_a approaches 30 the power approaches α

Calculating power for Gaussian data

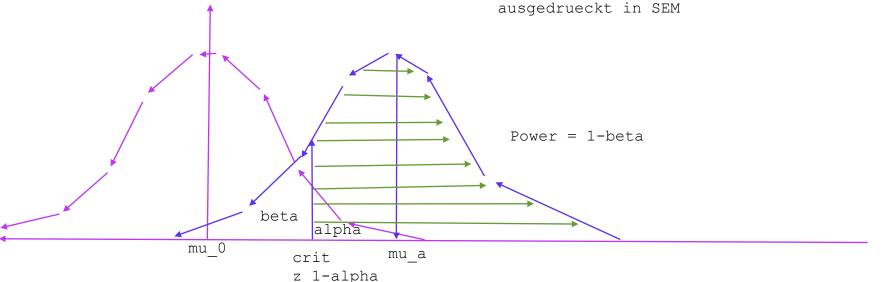
Assume that n is large and that we know σ Das ist keine Z-Statistik, weil wir 30 abziehen, und 30 ist der Mittelwert unter H0, nicht unter H1 (wo ja mu_a gilt)

$$1-eta=Pigg(rac{ar{X}-30}{\sigma/\sqrt{n}}>z_{1-lpha}\mid \mu=\mu_aigg)$$
 hier mit norm. Vertl statt T-Test >> z, sigma
$$=Pigg(rac{ar{X}-\mu_a+\mu_a-30}{\sigma/\sqrt{n}}>z_{1-lpha}\mid \mu=\mu_aigg) \qquad ext{Trick: mu_a abziehen und dazuzaehlen.}$$

$$=P\!\left(\frac{\bar{X}-\mu_a}{\sigma/\sqrt{n}}>z_{1-\alpha}-\frac{\mu_a-30}{\sigma/\sqrt{n}}\mid\mu=\mu_a\right) \begin{array}{l} \text{Jetzt ist es eine Z-Statistik, weil wir mu_a abziehen, unter H1.} \end{array}$$

$$=Pigg(Z>z_{1-lpha}-rac{\mu_a-30}{\sigma/\sqrt{n}}\mid \mu=\mu_aigg)$$

Das ist Differenz zw mu_a und mu_0 ausgedrueckt in SEM



Example continued

- · Suppose that we wanted to detect a increase in mean RDI of at least 2 events / hour (above 30).
- Assume normality and that the sample in question will have a standard deviation of 4;
- What would be the power if we took a sample size of 16?

```
 \begin{array}{lll} - Z_{1-\alpha} = 1.645 & & \underbrace{(\text{X}_- - \text{mu}_-0)}_{\text{sigma/sqrt}\,(\text{n})} & \sim \text{N(0,1) unter H0.} \\ - \frac{\mu_a - 30}{\sigma/\sqrt{n}} = 2/(4/\sqrt{16}) = 2 & & \text{sigma/sqrt}\,(\text{n}) \\ - P(Z > 1.645 - 2) = P(Z > -0.355) = 64\% & & \underbrace{\text{Unter H1 aendert sich mu und sigma:}}_{\text{mu}_-a - \text{mu}_-0}, & 1) \\ - \frac{\text{mu}_-a - \text{mu}_-0}_{\text{sigma/sqrt}\,(0)}, & 1) \\ & & & \\ \text{pnorm(-0.355, lower.tail = FALSE)} & & \\ \text{Daher geht auch:} \end{array}
```

```
pnorm(1.645, mu=2, sd=1)
```

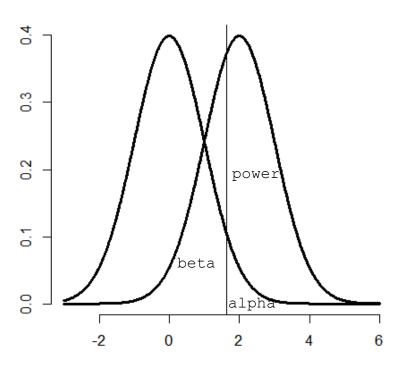
[1] 0.6387

```
Unter H0: (Xmean - mu_0) / (sigma/sqrt(n)) \sim N(0,1)
Unter H1? immer noch \sim N, mit mu = mu_a - mu_0 und var 1.
```

Note

- Consider $H_0: \mu = \mu_0$ and $H_a: \mu > \mu_0$ with $\mu = \mu_a$ under H_a .
- ' Under H_0 the statistic $Z=rac{\sqrt{n}(ar{X}-\mu_0)}{\sigma}$ is N(0,1)
- ' Under $H_a~Z$ is $Nigg(rac{\sqrt{n}(\mu_a-\mu_0)}{\sigma}\,,1igg)$
- We reject if $Z > Z_{1-\alpha}$

```
sigma <- 10; mu_0 = 0; mu_a = 2; n <- 100; alpha = .05
plot(c(-3, 6),c(0, dnorm(0)), type = "n", frame = false, xlab = "Z value", ylab = "")
xvals <- seq(-3, 6, length = 1000)
lines(xvals, dnorm(xvals), type = "l", lwd = 3)
lines(xvals, dnorm(xvals, mean = sqrt(n) * (mu_a - mu_0) / sigma), lwd = 3)
abline(v = qnorm(1 - alpha))</pre>
```



Question

• When testing $H_a: \mu > \mu_0$, notice if power is $1 - \beta$, then

$$1-\beta=P\left[Z>z_{1-\alpha}-\frac{\mu_a-\mu_0}{\sigma/\sqrt{n}}\,|\,\mu=\mu_a\right)=P(Z>z_\beta)$$

 • This yields the equation
$$z_{1-\alpha}-\frac{\sqrt{n}(\mu_a-\mu_0)}{\sigma}=z_\beta$$

• Unknowns: μ_a , σ , n, β

• Knowns: μ_0 , α

· Specify any 3 of the unknowns and you can solve for the remainder

Notes

- The calculation for $H_a: \mu < \mu_0$ is similar
- For $H_a: \mu \neq \mu_0$ calculate the one sided power using $\alpha/2$ (this is only approximately right, it excludes the probability of getting a large TS in the opposite direction of the truth)
- Power goes up as α gets larger
- Power of a one sided test is greater than the power of the associated two sided test
- · Power goes up as μ_1 gets further away from μ_0
- Power goes up as n goes up
- Power doesn't need μ_a , σ and n, instead only $\frac{\sqrt{n}(\mu_a-\mu_0)}{\sigma}$
 - The quantity $\frac{\mu_a \mu_0}{\sigma}$ is called the effect size, the difference in the means in standard deviation units.
 - Being unit free, it has some hope of interpretability across settings

T-test power

- \cdot Consider calculating power for a Gossett's T test for our example
- · The power is

$$Pigg(rac{ar{X}-\mu_0}{S/\sqrt{n}}>t_{1-lpha,n-1}\mid \mu=\mu_aigg)$$

- · Calcuting this requires the non-central t distribution.
- · power.t.test does this very well
 - Omit one of the arguments and it solves for it

Example

[1] 0.604

```
power.t.test(n = 16, delta = 2, sd=4, type = "one.sample", alt = "one.sided")$power
```

[1] 0.604

```
power.t.test(n = 16, delta = 100, sd=200, type = "one.sample", alt = "one.sided")$power
```

[1] 0.604

>> Resultat immer gleich, da Power abhaengt von (mu0-mu1)/sigma = delta/sigma [hier = sd]

Example

Man kann der Fkt power.t.test immer drei Variablen geben, und sie spukt die vierte aus:

```
power.t.test(power = .8, delta = 2 / 4, sd=1, type = "one.sample", alt = "one.sided")$n
```

[1] 26.14

power.t.test(power = .8, delta = 2, sd=4, type = "one.sample", alt = "one.sided")\$n

[1] 26.14

power.t.test(power = .8, delta = 100, sd=200, type = "one.sample", alt = "one.sided")\$n

[1] 26.14

Achtung: viele 'dials' -> viele Moeglichkeiten, sich zu irren (oder selbst zu beluegen). Daher die Powerberechnung so simple wie moeglich machen!