

Power Analysis Simulation



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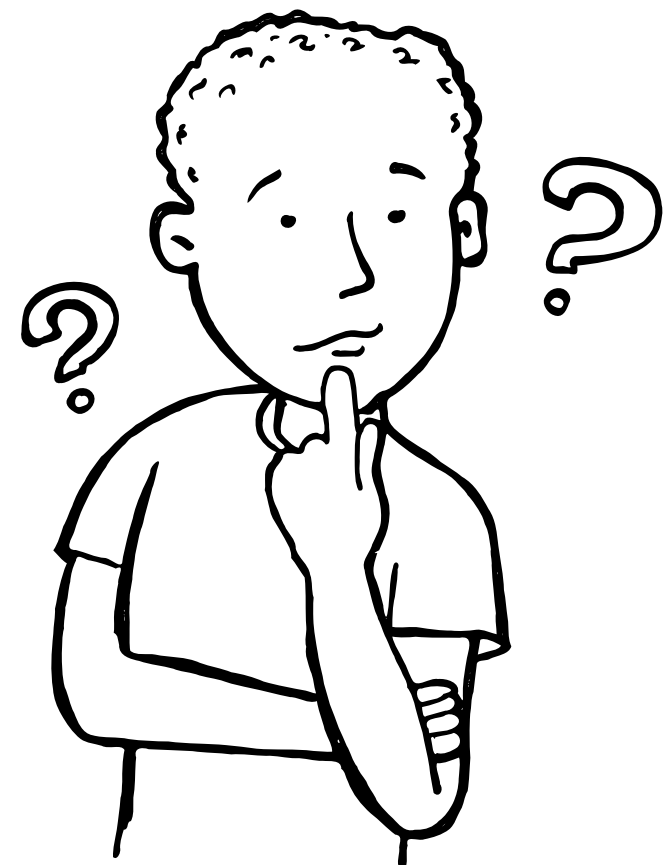
Shiny App Prototype

Test of Hypothesis

Null Hypothesis (H_0)

Alternative Hypothesis (H_1)

- generated from the data



POWER

(of the implied statistical test)

$P(H_0 \text{ is rejected} \mid H_0 \text{ is false})$

Type II Error 

Accepting H_0 when H_0 is false (β)

1 COLLECT THE DATA

**2 CALCULATE THE TEST
STATISTICS AND IT'S P-VALUE**

**Decision to Accept or Reject
the Null Hypothesis**



How to estimate Power?

1

Specify the model

- the statistical model
- parameters
- the effect size

2

Simulate Data

- generate observations from the model

3

Calculate the P-value

4

Repeat the Process

POWER: proportion of times where the test actually gave a significant result

Simulating the Power of Statistical Tests: A Collection of R Examples

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Abstract

This paper illustrates how to calculate the power of a statistical test by computer simulation. It provides R code for power simulations of several classical inference procedures including one- and two-sample t tests, chi-squared tests, regression, and analysis of variance.

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Regression

BinomialTest

ANOVA

Test of Correlation

Two-Sample t-Test

One-Sample t-Test

2-sample t-test for independent samples

This paper illustrates how to calculate the power of one- and two-sample t tests, chi

Two-sample t-test tests the hypothesis $H_0 : \mu_x - \mu_y = \delta = 0$, where μ_x and μ_y are the means of each sample. The present application generates 2000 times a pair of samples based on the parameters that are entered below and at each iteration performs a t-test and calculates the accompanying p-value. Power is the proportion of times where the test actually gave a significant result ($p < 0.05$).

Please enter the parameters for your samples

First sample

sample size

sample mean

standard deviation

Second sample

sample size

sample mean

standard deviation

Calculate power

Shiny App!

```
ui <- fluidPage(  
  titlePanel("Test of Correlation"),  
  withMathJax(),  
  "Power (also called true positive rate, hit rate, sensitivity, or recall) is defined as  $1 - \beta$ . It is the probability of a statistical test to detect an effect of a given size. Therefore, designing an experiment to have a good chance to find an effect means making sure its power is high enough. High power is a necessary condition for valid inference. A high power value can be considered when the proportion of significant results is about 0.8 or higher.",  
  hr(),  
  fluidRow(  
    column(2, numericInput("n", "Number of observations", 60, min = 10, max = 500, step = 1)),  
    column(2, offset = 1, numericInput("s", "Standard deviation:", 15, min = 1, max = 100, step = .1)),  
    column(2, offset = 1, numericInput("r", "Effect size:", .3, min = 0, max = 1, step = .01)),  
    column(2, offset = 1, numericInput("rho0", "Null Hypothesis (H0):", .6, min = 0, max = 1, step = .01)),  
    column(2, offset = 5, actionButton("goButton", "Calculate power")),  
    column(10, offset = 0, verbatimTextOutput("results")),  
    plotOutput("plotResults")  
  )  
)  
  
corr <- function(n, s, r, rho0){  
  pval <- replicate(2000, {  
    sxy <- r*s*s  
    x <- MASS::mvrnorm(n, mu = c(100, 100),  
      Sigma = matrix(c(s^2, sxy, sxy, s^2), 2,  
        z <- 1/2 * (log((1 + cor(x[, 1], x[, 2]))) / (1 - cor(x[, 1],  
          log((1 + rho0) / (1 - rho0))) * sqrt(n - 3)  
        2*pnorm(-abs(z))  
      })  
    })  
  }  
  
server <- function(input, output){  
  rv <- eventReactive(input$goButton,  
    corr(input$n, input$s, input$r, input$rho0)  
  
  output$results <- renderPrint({  
    data2 <- rv() < 0.05  
    input$goButton  
    isolate(  
      cat("The sample size of", input$n, "is sufficient to detect a deviation of",  
        input$r, "from the null hypothesis (H0 = ", input$rho0, ") with a power of",  
        rv(), ".")  
    )  
  })  
  
  output$plotResults <- renderPlot({  
    input$goButton  
    data <- rv()  
  })  
}
```

Test of Correlation

Power (also called true positive rate, hit rate, sensitivity, or recall) is defined as $1 - \beta$. It is the probability of a statistical test to detect an effect of a given size. Therefore, designing an experiment to have a good chance to find an effect means making sure its power is high enough. High power is a necessary condition for valid inference. A high power value can be considered when the proportion of significant results is about 0.8 or higher.

Number of observations

60

Standard deviation:

15

Effect size:

0,3

Null Hypothesis (H0):

0,6

Calculate power

The sample size of 60 is sufficient to detect a deviation of 0.3 from the null hypothesis (H0 = 0.6) with a power of 0.831

P-VALUES. Power: 0.831

