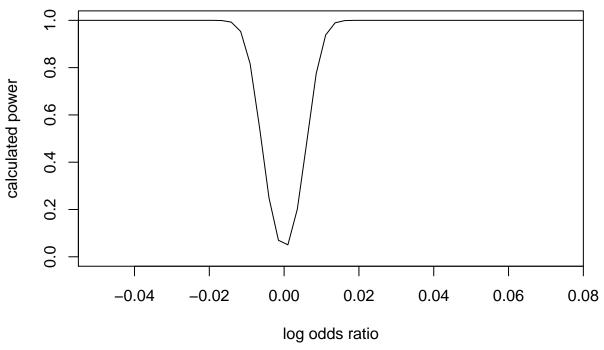
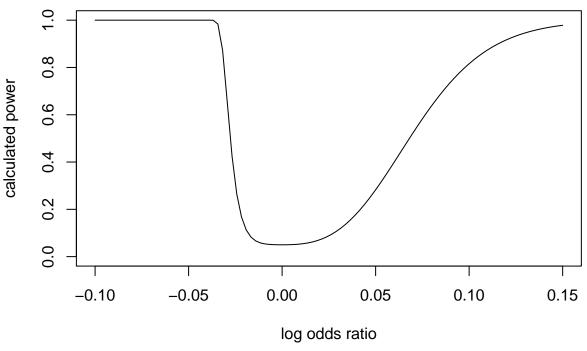
powerlog: power calculations for logistic regressions

Yezi "Zora" Yang

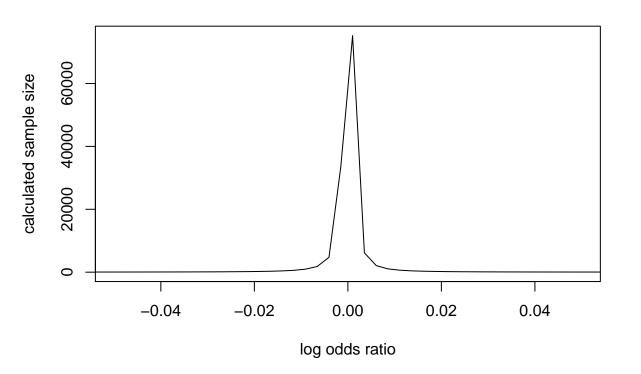
```
# install.packages("../powerlog_0.1.0.tar.gz", repos = NULL, type = "source")
# library(powerlog)
source("R/power.R")
source("R/sampsize.R")
source("R/effectsize.R")
Example 1: S-B Table 1
calc_pwr(b0 = -1, b1 = 0.33, n = 500, dist = "normal", mu = 0, sigma = 1)
## [1] 0.8901227
calc_{samp_{size}(b0 = -1, b1 = 0.33, targetpwr = 0.9, alpha = 0.05,
               dist = "normal", mu = 0, sigma = 1)
## [1] 517.2049
approx_pwr(b0 = -1, b1 = 0.33, b2 = 0.9, rho = 0.5, n = 500,
           dist = "normal", mu = 0, sigma = 1)
## [1] 0.7473011
approx_samp_size(b0 = -1, b1 = 0.33, b2 = 0.9, rho = 0.5, targetpwr = 0.9,
                 alpha = 0.05, dist = "normal", mu = 0, sigma = 1)
## [1] 761.8756
# TODO: replace with data URL
nhanes = read.csv("Data/nhaneslarge.csv")
Example 2: NHANES The data set
names (nhanes)
## [1] "BPXSAR"
                    "BPXDAR"
                                "BPXDI1"
                                             "BPXDI2"
                                                         "race_ethc"
## [6] "gender"
                    "DR1TFOLA" "RIAGENDR" "BMXBMI"
                                                         "RIDAGEYR"
nhanes <- nhanes[complete.cases(nhanes[,c("BMXBMI","RIDAGEYR","BPXDAR")]),]</pre>
nhanes$HighDBP = ifelse(nhanes$BPXDAR>70, 1, 0)
nhanes$BMXBMI = as.numeric(nhanes$BMXBMI)
nhanes$RIDAGEYR = as.numeric(nhanes$RIDAGEYR)
mod = glm(HighDBP ~ RIDAGEYR, data = nhanes)
mod.adj = glm(HighDBP ~ RIDAGEYR + BMXBMI, data = nhanes)
prop_nhanes = sum(nhanes$HighDBP==1)/length(nhanes$HighDBP)
Power curves
granu = 100
lowerb = -0.1
upperb= 0.15
b1_hypos = seq(lowerb, upperb, length.out = granu)
xmat_nhanes_univ = cbind(rep(1, nrow(nhanes)), nhanes$RIDAGEYR)
b0_hats = sapply(X = b1_hypos, FUN = findb0emp, prop=prop_nhanes,
```





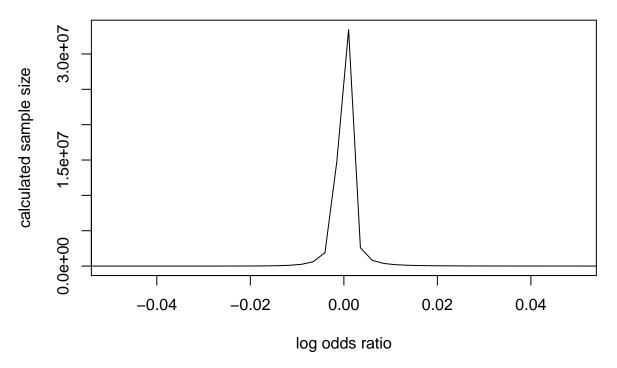
```
mod.adj = glm(HighDBP ~ RIDAGEYR + BMXBMI, data = nhanes); summary(mod.adj)
##
## Call:
## glm(formula = HighDBP ~ RIDAGEYR + BMXBMI, data = nhanes)
##
## Deviance Residuals:
##
       Min
                 1Q
                      Median
                                   3Q
                                           Max
##
  -0.8099 -0.4444 -0.2384
                               0.4991
                                        0.7953
##
## Coefficients:
##
                 Estimate Std. Error t value Pr(>|t|)
## (Intercept) -0.0314634 0.0666723 -0.472
                                                0.637
                0.0032497
                           0.0007862
                                       4.134 3.88e-05 ***
## RIDAGEYR
## BMXBMI
                          0.0024380
                                       5.388 8.90e-08 ***
                0.0131364
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## (Dispersion parameter for gaussian family taken to be 0.2337317)
##
       Null deviance: 245.66 on 990 degrees of freedom
## Residual deviance: 230.93 on 988 degrees of freedom
## AIC: 1376.8
## Number of Fisher Scoring iterations: 2
b0_MLE = -0.0314634; b1_MLE = 0.0032497; b2_MLE = 0.0131364
fitted_p = expit(xmat_nhanes_biv%*%c(b0_MLE, b1_MLE, b2_MLE))
p_vals = vector()
```

```
for (i in 1:1000){
  fitted_Y = rbinom(nrow(nhanes), 1, prob = fitted_p)
 p_vals[i] = summary(glm(fitted_Y ~ RIDAGEYR + BMXBMI, data = nhanes))$coeff[2,4]
sum(p_vals<0.05)/length(p_vals)</pre>
## [1] 0.181
Sample size curve univariate
target_power = .80
# HighDBP ~ RIDAGEYR
ss_univ = vector(length = granu)
for (i in 1:length(powers_univ)){
  ss_i = calc_samp_size_emp(b0 = b0_hats[i], b1 = b1_hypos[i], xmat = xmat_nhanes_univ,
                   targetpwr = target_power, reg = "uni")
  ss_univ[i] = ss_i
}
plot(x = b1_hypos, y = ss_univ, type = "l", xlim = c(-0.05, 0.05),
     xlab = "log odds ratio", ylab = "calculated sample size",
     main = pasteO("Sample Size vs. Log Odds Ratio When Target Power = ", target_power))
```



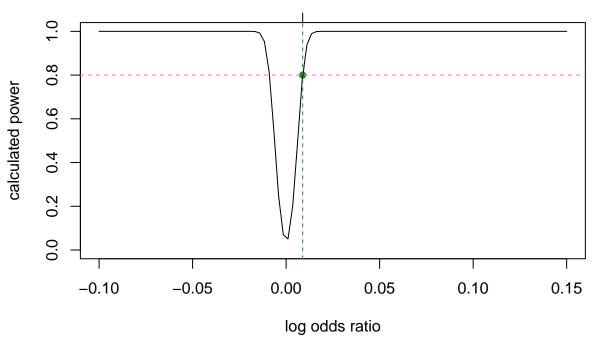
Sample size curve bivariate

```
# HighDBP ~ RIDAGEYR + BMXBMI
ss_biv = vector(length = granu)
for (i in 1:length(ss_biv)){
  b0_hat = findb0emp(b1 = b1_hypos[i], b2 = b2_hypos[i], prop = prop_nhanes, xmat = xmat_nhanes_biv, rest = calc_samp_size_emp(b0 = b0_hat, b1 = b1_hypos[i], b2 = b2_hypos[i], xmat = xmat_nhanes_biv, tall
```

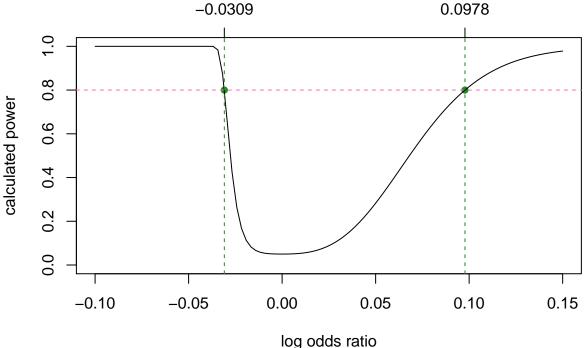


log odds ratio

Power vs. log odds ratio When Sample Size (N) = 991 0.00885



```
# HighDBP ~ RIDAGEYR + BMI
b1hat2u = calc_effect_size_emp(xmat=xmat_nhanes_biv, prop=prop_nhanes,
                              b2=b2_hypo, targetpwr=target_power,
                              alpha=0.05, reg = "multi", lower=0, upper=0.15)
b1hat21 = calc_effect_size_emp(xmat=xmat_nhanes_biv, prop=prop_nhanes,
                              b2=b2_hypo, targetpwr=target_power,
                              alpha=0.05, reg = "multi", lower=-.10, upper=0)
par(mar = c(5, 4, 5, 2) + 0.1)
plot(x = b1hat2u, y = target_power, type = "p", col = "forestgreen", pch = 16,
     xlab = "log odds ratio", ylab = "calculated power",
     ylim = c(0, 1), xlim = c(lowerb, max(upperb, b1hat2u*1.1)),
     main = paste0("Power vs. Log Odds Ratio When Sample Size (N) = ", nrow(nhanes)))
points(x = b1hat21,y = target_power, type = "p", col = "forestgreen", pch = 16 )
lines(x = b1_hypos, y = powers_biv)
abline(h = target_power, col = "hotpink", lty = 2)
axis(side = 3, at = b1hat2u, labels = signif(b1hat2u, 3))
axis(side = 3, at = b1hat21, labels = signif(b1hat21, 3))
abline(v = b1hat2u, col = "forestgreen", lty = 2)
abline(v = b1hat21, col = "forestgreen", lty = 2)
```



Genetic data example:

0.00

##

29.40

45.50

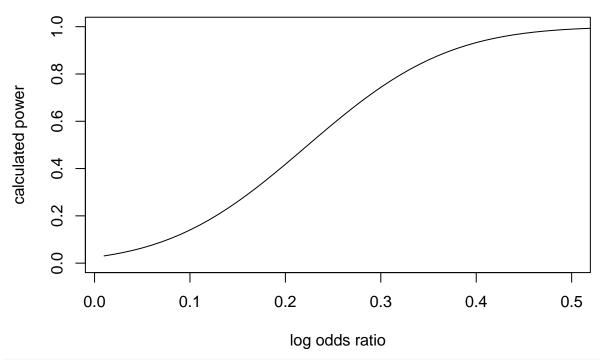
52.57

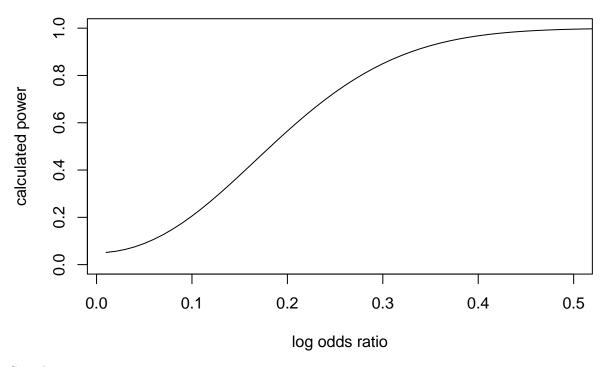
```
fms = read.delim("http://stat-gen.org/book.e1/data/FMS_data.txt", header=T, sep="\t")
#FAMuSS - functional SNPs associated with muscle size and strength
# trait is NDRM.CH - percentage change in muscle strength in non-dom arm
# before and after exercise training
# SNP is r577x, more at https://www.snpedia.com/index.php/Rs1815739
fms$Geno = as.numeric(fms$actn3_r577x)-1
fms$Male = as.numeric(fms$Gender=="Male") # a covariate to adjust for, if you need one.
table(fms$Male)
##
##
    0
## 607 426
table(fms$Geno)
##
             2
##
    0
         1
## 216 318 201
fms$Trait = fms$NDRM.CH
summary(fms$Trait)
##
      Min. 1st Qu. Median
                              Mean 3rd Qu.
                                              Max.
                                                       NA's
```

66.70 250.00

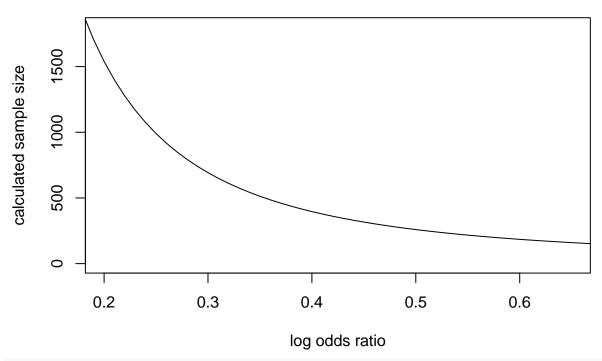
386

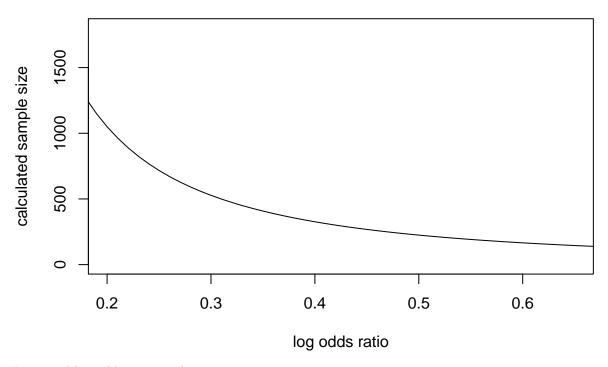
```
fms$binaryTrait = as.numeric(fms$Trait>=60)
fms2 = na.omit(fms[c("binaryTrait", "Geno")])
summary(glm(binaryTrait ~ Geno, data=fms2))$coef
                 Estimate Std. Error t value
                                                  Pr(>|t|)
## (Intercept) 0.30401277 0.03198963 9.503480 4.770272e-20
               0.05200051 0.02595494 2.003492 4.557299e-02
fms3 = na.omit(fms[c("binaryTrait", "Geno", "Male")])
summary(glm(binaryTrait ~ Geno + Male, data=fms3))$coef
                  Estimate Std. Error t value
                                                    Pr(>|t|)
## (Intercept) 0.43968357 0.03355002 13.105313 1.093962e-34
## Geno
                0.05327588 0.02436242 2.186806 2.914211e-02
## Male
               -0.33698760 0.03717703 -9.064403 1.762057e-18
Same as before, parameters and etc.
mod.b = glm(binaryTrait ~ Geno, data=fms2)
prop.fms2 = sum(fms2$binaryTrait==1)/length(fms2$binaryTrait); prop.fms2
## [1] 0.3548922
mod.b.adj = glm(binaryTrait ~ Geno + Male, data=fms3)
mean(fms3$Geno, na.rm=T); mean(fms3$Male)
## [1] 0.9784411
## [1] 0.4063018
sd(fms3$Geno, na.rm=T); sd(fms3$Male)
## [1] 0.7501049
## [1] 0.4915499
prop.fms3 = sum(fms3$binaryTrait==1)/length(fms3$binaryTrait); prop.fms3
## [1] 0.3548922
Setup
#library(powerlog)
granu.b = 100
lowerb.b = 0.01
upperb.b = 1
b1.hypos.b = seq(lowerb.b, upperb.b, length.out = granu.b)
xmat.fms.univ = cbind(rep(1, nrow(fms2)), fms2$Geno)
b0.hats.b = sapply(X = b1.hypos.b, FUN = findb0emp, prop=prop.fms2,
                   xmat = xmat.fms.univ, lower = -1, upper = 1)
b2.hypo.b = -0.33
xmat.fms.biv = cbind(rep(1, nrow(fms3)), fms3$Geno, fms3$Male)
Power curves
# BinaryTrait ~ Genotype
powers.univ.b = vector(length = granu.b)
for (i in 1:length(powers.univ.b)){
 pwr.i = calc_pwr_emp(b0 = b0.hats.b[i], b1 = b1.hypos.b[i], xmat = xmat.fms.univ,
                   alpha =.05, reg = "uni")
```



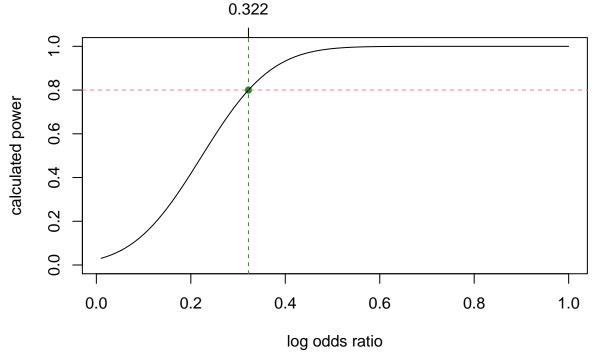


Sample size curves





Estimated log odds ratio at the power target



Power vs. log odds ratio When Sample Size (N) = 603 0.277

