统计软件 HW11

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贮存可靠性评估问题

```
reliab <- read.csv("store-reliab-data.csv")
Z1 <- reliab[reliab['testid'] == 1, c('times', 'delta')]
theta <- reliab[reliab['testid'] == 1, 'true_theta'][1] # 数据中的 true theta 都是 10
P_theta_Z <- function(P, delta)
prod(ifelse(delta, P, 1 - P))
```

直接求值法

```
G.direct <- function(theta, Z0) {
    N <- nrow(Z0)
    P <- exp(-Z0[, 1] / theta)
    n_0 <- sum(Z0[, 2])
    G.sum <- 0

# 满足条件 1 的 Zn

if (n_0 < N) {
    for (n in (n_0 + 1):N) {
        delta.combn <- combn(N, n)

        G.temp <- vapply(1:ncol(delta.combn), function(i) {
            delta1 <- rep(0, N)
            delta1[delta.combn[, i]] <- 1
            return (P_theta_Z(P, delta1))
        }, FUN.VALUE = 1)
```

```
G.sum <- G.sum + sum(G.temp)</pre>
    }
  }
  # 满足条件 2 的 Zn (nO==N)
  delta.combn <- combn(N, n_0)
  G.temp <- vapply(1:ncol(delta.combn), function(i) {</pre>
    delta1 <- rep(0, N)
    delta1[delta.combn[, i]] <- 1</pre>
    if (sum(ZO[, 1] * delta1) >= sum(ZO[, 1] * ZO[, 2]))
      return (P_theta_Z(P, delta1))
    else
      return (0)
  }, FUN.VALUE = 1)
  G.sum <- G.sum + sum(G.temp)</pre>
  G.sum
}
G.direct <- Vectorize(G.direct, vectorize.args = "theta") # 向量化
```

随机模拟法

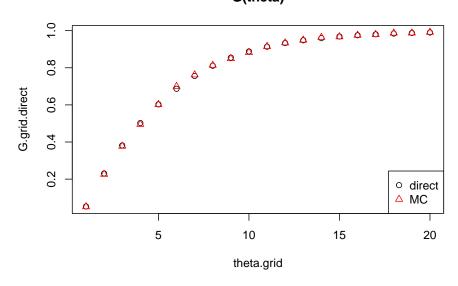
```
G.MC <- function(theta, ZO, N = 1e4) {
    n0 <- sum(ZO[, 2])
    G.MC.single <- function(theta, ZO) {
    # 返回 I(Zn(i)>Zn)
    X <- rexp(nrow(ZO), 1 / theta)
    delta <- X > ZO[, 1]
    # 满足条件 1 的 Zn
    if (sum(delta) > nO)
        return (TRUE)
    if (sum(delta * ZO[, 1]) >= sum(ZO[, 2] * ZO[, 1]))
        return (TRUE)
    }
```

```
return (FALSE)
}
x <- replicate(N,G.MC.single(theta, ZO))
c(mean = mean(x), sd = sd(x))
}
G.MC <- Vectorize(G.MC, vectorize.args = "theta")</pre>
```

试验两方法求解出的 θ 是否相同

```
#对 testid 为 1 的数据进行试验,选取不同的 $\theta$
ptm <- proc.time()</pre>
G.direct(10, Z1)
## [1] 0.8880072
proc.time() - ptm
##
      user system elapsed
      0.01
              0.00
                      0.02
##
ptm <- proc.time()</pre>
G.MC(10, Z1, 1e4)
##
             [,1]
## mean 0.8952000
        0.3063109
## sd
proc.time() - ptm
      user system elapsed
##
##
      0.35
              0.00
                      0.34
```

G(theta)



随机模拟不同于随机误差。可以看到在 n=8 使,如果随机模拟的次数设得较高,反而比直接求解慢。但当 n 变大时,直接求解得时间复杂度指数上升,随机模拟法只是线性上升,优势就体现出来了,如下面的试验:

```
Z2 <- reliab[reliab['testid'] <= 2, c('times', 'delta')]
ptm <- proc.time()
G.direct(10, Z2)</pre>
```

[1] 0.640969

```
proc.time() - ptm
      user system elapsed
      0.98
              0.00
##
                      0.99
ptm <- proc.time()</pre>
G.MC(10, Z2, 1e4)
##
             [,1]
## mean 0.6395000
## sd
        0.4801696
proc.time() - ptm
##
      user system elapsed
      0.27
              0.00
##
                      0.26
    可以看到,n=16 时,MC 方法得效率优势已经体现。
求解方程 G(\theta) - \alpha = 0 (\alpha = 0.05)
```

```
直接法求 G,二分法求根
```

```
if (G(theta.max) < alpha) {</pre>
    print("theta.max is not large enough!")
    return (FALSE)
  while (abs(theta.max - theta.min) > eps) {
    theta.new <- (theta.min + theta.max) / 2</pre>
    if (G(theta.new) > alpha)
      theta.max <- theta.new
    else
      theta.min <- theta.new
  }
  (theta.min + theta.max) / 2
}
# 直接法求 G(theta)
ptm=proc.time()
solve.theta(0.05, Z1)
## [1] 0.9813935
proc.time()-ptm
      user system elapsed
      0.12
              0.00
                      0.13
# 随机模拟法求 G(theta)
ptm=proc.time()
solve.theta(
 0.05,
 Z1,
  G = function(theta)
    G.MC(theta, Z1)[1]
## [1] 1.001075
```

```
proc.time()-ptm
```

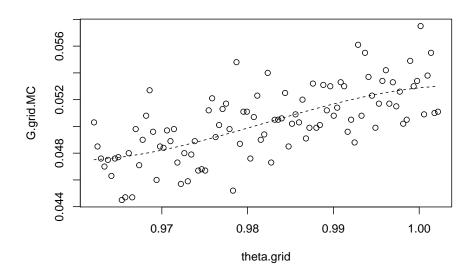
```
## user system elapsed
## 5.13 0.00 5.14
```

可以看到,随机模拟法由于每次计算有随机误差,用 naive 得二分法求根结果不够准确,需要对求根算法做一些改进。下面是利用样条回归改进后的算法:

```
solve.theta <- function(alpha,</pre>
                        G.method = "direct",
                        N = NULL, spline.plot =TRUE,
                        theta.min = 0,
                        theta.max = 50,
                        eps = 1e-4) {
  if (sum(Z0==1)==0) return(0) # 所有 n 次试验都失效的情形
  # 二分法求根
  if (alpha >= 1) {
   print ("Illegal value of alpha!")
   return(FALSE)
  }
  if (G.method=="direct"){
    G<-function(theta) G.direct(theta, Z0)</pre>
 }
  else if (G.method=="MC"){
    G<-function(theta) G.MC(theta,ZO,ifelse(is.null(N),1e4,N))[1]
  }
  else {
    print("Illegal G.method!")
   return(FALSE)
  if (G(theta.max) < alpha) {</pre>
   print("theta.max is not large enough!")
```

```
return (FALSE)
  }
  while (abs(theta.max - theta.min) > eps) {
    theta.new <- (theta.min + theta.max) / 2
    if (G(theta.new) > alpha)
     theta.max <- theta.new
    else
      theta.min <- theta.new
  }
  if (G.method=="direct") (theta.min + theta.max) / 2
  else if (G.method=="MC"){
    #并不直接返回,而是在其领域内做样条回归,在线上找最接近 alpha 的点
    theta.grid<-seq(theta.min-200*eps,theta.max+200*eps,length.out = 100)
   G.grid.MC<-vapply(theta.grid,G,FUN.VALUE = 1)</pre>
   require(splines)
   reg.spline.fitted<-lm(G.grid.MC~bs(theta.grid))$fitted.values
   if (spline.plot){
     plot(theta.grid,G.grid.MC)
     lines(theta.grid,reg.spline.fitted,lty="dashed")
   }
   theta.grid[which.min(abs(reg.spline.fitted-alpha))]
  #(theta.min + theta.max) / 2
}
ptm=proc.time()
solve.theta(0.05, Z1, G.method = "MC")
```

Loading required package: splines



[1] 0.980724

```
proc.time()-ptm

## user system elapsed
## 24.30 0.01 24.44
```

可以看到,用改进后的算法虽然会花费更多时间,但是很大程度上减小了随机误差造成的影响,更加准确。

求解 store-reliab-data.csv 中问题

```
ptm=proc.time()
result <- vapply(1:10, function(testid) {
   Z <- reliab[reliab['testid'] == testid, c('times', 'delta')]
   theta <- reliab[reliab['testid'] == testid, 'true_theta'][1]
   alpha <- reliab[reliab['testid'] == testid, 'alpha'][1]
   c(direct = solve.theta(alpha, Z),</pre>
```

```
MC = solve.theta(alpha, Z, G.method = "MC", spline.plot = F))
formula = 0.5, 0.5
proc.time()-ptm
##
     user system elapsed
           0.05 236.81
## 235.88
result
##
              [,1]
                      [,2]
                             [,3]
                                       [,4] [,5] [,6]
                                                             [,7]
## direct 0.9813786 5.269861 4.315329 4.904222 26.02706
                                                       0 15.64975
         0.9832704 5.265162 4.285385 4.891751 26.54250
                                                       0 15.63808
##
             [,8] [,9]
                           [,10]
## direct 2.610636 5.088663 4.026270
## MC
        2.624348 5.078317 4.037358
```