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## INFERRING INEQUALITY Examples and RESULTS
The first part of this file contains examples to illustrate the functions specified in the Inferring Inequality Functions file.
Run all of the code in this file before running these experiments.

## The second part produces the results with the results published in Abul Naga, Stapenhurst and Yalonetzky "Inferring Inequality: Testing for Median Preserving Spreads in Ordinal Data":
# Happiness Inequality in the United States (table 5)
# Inequality in Self-assessed Health in Europe (figure 6)
# Size-boundary and power-locus curves (figures 3 and 5)

## Please send any questions or comments to Christopher Stapenhurst, cpstapenhurst.academic.wa

##### PART 1 EXAMPLES

### Input data (from section 3.1 worked example)
# Self-reported happiness response counts 2002 (x) and 1972 (y)
x <- c(184, 765, 450)
y <- c(218, 789, 599)

### Descriptive analysis
# sample sizes
nx<-sum(x)
ny<-sum(y)
c(nx,ny)

# empirical mass functions
f<-x/nx
g<-y/ny
rbind(f,g)

# empirical distribution functions
F<-cumsum(f)
G<-cumsum(g)
L<-cumsum(x+y)/(nx+ny)
rbind(F,G)

#MPS plot
mps.plot(F,G)

# median categories
find.m(F)
find.m(G)

# is F and mps of G?
is.mps(G,F)
# is G and mps of F?
is.mps(F,G)

### Null hypothesis: G is not an MPS of F.

### Test statistics
z.stat(x,y)
LR.stat(x,y)

#z stat worked example
signal<-sqrt((1-L)/(nx+ny))
((G-F)/signal/(nx+ny))*sqrt(nx+ny)[1]

#LR stat worked example
cmle<-find.CMLE(x,y)
ftilde<-cmle[1:3]
gtilde<-cmle[1:3+3]
2*log(dmultinom(x,prob=x)+dmultinom(y,prob=y)/dmultinom(x,prob=ftilde)/dmultinom(y,prob=gtilde))

##### Test functions
asyp.test(x,y, stat.fun="LR.stat")
asyp.test(x,y, stat.fun="z.stat")
boot.test(x,y, stat.fun="LR.stat",B=9999)
boot.test(x,y, stat.fun="z.stat",B=9999)

### PART 2 RESULTS

# create a new folder to store results and change directory.
setwd("Inference with ordered states/results")

### Happiness Inequality in the United States
# This code replicate the results of Inferring inequality table 5:
# i.e. the bootstrap p-value of the orderings displayed in Dutta and Foster (2013) table 2.

# Input data from Dutta and Foster (2013) table 1.
mat<-matrix(
  c(
    1972, 13.600, 49.100, 37.300, 1606,
    1973, 12.286, 50.932, 36.782, 1500,
    1974, 12.515, 49.194, 38.291, 1480,
    1975, 12.973, 53.630, 33.397, 1485,
    1976, 12.249, 52.920, 34.831, 1499,
    1977, 11.013, 53.242, 35.745, 1527,
    1978, 8.369, 56.189, 35.442, 1517,
    1980, 11.600, 52.000, 36.400, 1462,
    1982, 11.700, 53.499, 34.801, 1505,
    1983, 12.092, 56.239, 31.668, 1573,
    1984, 11.614, 52.091, 36.295, 1445,
    1985, 8.600, 58.400, 33.100, 1530,
    1986, 9.200, 55.800, 35.000, 1449,
    1987, 9.700, 53.300, 37.000, 1437,
    1988, 8.241, 55.695, 36.065, 1466,
    1989, 8.793, 56.737, 34.470, 1526,
    1990, 7.740, 56.527, 35.733, 1361,
    1991, 9.485, 58.004, 32.511, 1504,
    1993, 9.736, 56.865, 33.399, 1601,
    1994, 11.287, 58.216, 30.497, 2977,
    1996, 10.502, 57.357, 32.141, 2885,
    1998, 10.896, 55.851, 33.253, 2806,
    2000, 9.644, 56.435, 33.921, 2777,
    2002, 11.282, 56.846, 32.872, 1369,
    2004, 11.696, 54.718, 33.585, 1337,
    2006, 10.552, 55.901, 33.547, 2828,
    2008, 13.289, 54.749, 31.962, 1942,
    2010, 14.192, 57.010, 28.798, 2039)
  ,ncol=5,byrow=TRUE)

# format data to give samples x and y
mat1<-round(mat[,2:4]*mat[,5])/100
mat2<-cbind(matrix(rep(5*mat1),nrow(mat1)),ncol=3,byrow=TRUE),matrix(rep(mat1,each=nrow(mat1)),ncol=3))
set.seed(1)
# apply bootstrap LR test with B=499 bootstrap samples
res<-round(apply(mat2,1,function(z) boot.test(z[1:3],z[4:6], B=499)),2)
res1<-round(matrix(res,nrow=nrow(mat1)),2)
colnames(res1)<-mat[,1]
rownames(res1)<-mat[,1]

#display results in table 5
res1

#calculate the number of significant results at 1, 5 and 10% levels.
sum(res1<=0.01)
sum(res1<=0.05)
sum(res1<=0.1)
sum(res1<1)

### Inequality in Self-assessed Health in Europe
x<-round(c(0.01,0.04,0.19,0.54,0.22) * 13328) # netherlands
y<-round(c(0.03,0.06,0.21,0.45,0.25) * 5906) # denmark

nx<-sum(x)
ny<-sum(y)
k<-length(x)
f<-x/sum(x)
F<-cumsum(f)
g<-y/sum(y)
G<-cumsum(g)

# bar graph of mass functions f and g
pdf("EUMbarplot.pdf",height=7,width=7)
barplot(rbind(f,g), main="Self reported health",
  legend = c("Netherlands","Denmark"), xlab="Health category", col=c("darkblue","red"),
  names.arg=c("(1) very bad","(2) bad","(3) average","(4) good","(5) very good"),
  beside=TRUE)
dev.off()

# MPS plot of F and G.
pdf("EUMMPSplot.pdf",height=3,width=3)
mps.plot(F,G)
dev.off()

# find closest null distribution to (f,g)
cmle<-find.CMLE(f=nx,g=ny)
FO<-cmle[k:k]
GO<-cmle[k+1:k]
FO<-cumsum(FO)
GO<-cumsum(GO)

# Plot closest null distribution
pdf("EUMOMPSplot.pdf",height=3,width=3)
mps.plot(FO,GO)
dev.off()

# initialise data frame and size and power graph objects
df<-data.frame(l=0:1000/1000)
rng<-0.1 # plot range
grobo0<- ggplot()+ylab("Actual size")+xlab("Nominal size")+geom_line(data=df,aes(l,1))+coord_cartesian(ylim=c(0, rng),xlim=c(0, rng))
groboA0<- ggplot()+ylab("Power")+xlab("Actual size")+geom_line(data=df,aes(l,1))+coord_cartesian(ylim=c(0, rng),xlim=c(0, rng))

# set the seed
set.seed(2)
# 10000 monte carlo simulations
M<-10000
# Specify B=499 Bootstrap samples
B<-499

df$aLR0<-ecdf.fun(monte.carlo(asyp.test,f=f0,g=g0,nx=nx,ny=ny,M=M,stat.fun="LR.stat"))
df$aZ0<-ecdf.fun(monte.carlo(asyp.test,f=f0,g=g0,nx=nx,ny=ny,M=M,stat.fun="z.stat"))
df$aLR<-ecdf.fun(monte.carlo(asyp.test,f=f,g,g,nx=nx,ny=ny,M=M,stat.fun="LR.stat"))
df$aZ<-ecdf.fun(monte.carlo(asyp.test,f=f,g,g,nx=nx,ny=ny,M=M,stat.fun="z.stat"))
df$bLR0<-ecdf.fun(monte.carlo(boot.test,f=f0,g=g0,nx=nx,ny=ny,M=M,B=B,stat.fun="LR.stat"))
df$bZ0<-ecdf.fun(monte.carlo(boot.test,f=f0,g=g0,nx=nx,ny=ny,M=M,B=B,stat.fun="z.stat"))
df$bLR<-ecdf.fun(monte.carlo(boot.test,f=f,g,g,nx=nx,ny=ny,M=M,B=B,stat.fun="LR.stat"))
df$bZ<-ecdf.fun(monte.carlo(boot.test,f=f,g,g,nx=nx,ny=ny,M=M,B=B,stat.fun="z.stat"))

# add results to graph objects
grobo<-grobo0+geom_line(data=df,aes(l,bLR0),color="blue")+geom_line(data=df,aes(l,aLR0),color="red")+
  geom_line(data=df,aes(l,bZ0),color="blue4")+geom_line(data=df,aes(l,aZ0),color="red4")+
  geom_line(data=df,aes(l,bLR0+geom_line(data=df,aes(l,bLR0,bLR4),color="blue")+geom_line(data=df,aes(l,aLR0,aLR4),color="red4")+
  geom_line(data=df,aes(l,bZ0,bZA),color="blue4")+geom_line(data=df,aes(l,aZ0,aZA),color="red4"))

# produce and save size and power curves graph objects
pdf("EUsize.pdf",height=3,width=3)
grobo
dev.off()
pdf("EUpower.pdf",height=3,width=3)
groboA0
dev.off()

# arrange size and power curves in panel.
pars<-paste(c("f","paste(round(f,2), collapse = ",")"), g=c("paste(round(g,2), collapse = ",")"), ",nx=",nx,",ny=",ny), collapse = "")
pdf("EU2pan.pdf",height=3,width=6)
suppressWarnings(grid.arrange(grobo,groboA0,nrow=1,top="Inequality in self-assessed health")
dev.off()

##### Size-boundary and power-locus curves
set.seed(1)
M<-10000
B<-999

# Conduct experiments
data.10.10<-pcFun(10,10,M=M,B=B)
data.10.100<-pcFun(10,100,M=M,B=B)
data.10.1000<-pcFun(10,1000,M=M,B=B)
data.100.10<-pcFun(100,10,M=M,B=B)
data.1000.10<-pcFun(1000,10,M=M,B=B)
data.1000.100<-pcFun(1000,100,M=M,B=B)
data.1000.1000<-pcFun(1000,1000,M=M,B=B)
data.1000.1000<-pcFun(1000,1000,M=M,B=B)
data.1000.1000<-pcFun(1000,1000,M=M,B=B)

# Plot results

# size-boundary curves at 5% nominal level (figure 3)
size<-0.05
pdf("size05.pdf",height=7,width=7)
suppressWarnings(
  grid.arrange(
    size.plot.fun(data.10.10,size=size),
    size.plot.fun(data.10.100,size=size),
    size.plot.fun(data.10.1000,size=size),
    size.plot.fun(data.100.10,size=size),
    size.plot.fun(data.1000.10,size=size),
    size.plot.fun(data.1000.100,size=size),
    size.plot.fun(data.1000.1000,size=size),
    ncol=3,nrow=3)
)
dev.off()

# size-boundary curves at 1% nominal level
size<-0.01
pdf("size01.pdf",height=7,width=7)
suppressWarnings(
  grid.arrange(
    size.plot.fun(data.10.10,size=size),
    size.plot.fun(data.10.100,size=size),
    size.plot.fun(data.10.1000,size=size),
    size.plot.fun(data.100.10,size=size),
    size.plot.fun(data.1000.10,size=size),
    size.plot.fun(data.1000.100,size=size),
    size.plot.fun(data.1000.1000,size=size),
    ncol=3,nrow=3)
)
dev.off()

# size-boundary curves at 10% nominal level
size<-0.1
pdf("size1.pdf",height=7,width=7)
suppressWarnings(
  grid.arrange(
    size.plot.fun(data.10.10,size=size),
    size.plot.fun(data.10.100,size=size),
    size.plot.fun(data.10.1000,size=size),
    size.plot.fun(data.100.10,size=size),
    size.plot.fun(data.1000.10,size=size),
    size.plot.fun(data.1000.100,size=size),
    size.plot.fun(data.1000.1000,size=size),
    ncol=3,nrow=3)
)
dev.off()

# power-locus curves at 5% nominal level (figure 5)
size<-0.05
pdf("power05.pdf",height=7,width=7)
suppressWarnings(
  grid.arrange(
    power.plot.fun(data.10.10,size=size),
    power.plot.fun(data.10.100,size=size),
    power.plot.fun(data.10.1000,size=size),
    power.plot.fun(data.100.10,size=size),
    power.plot.fun(data.1000.10,size=size),
    power.plot.fun(data.1000.100,size=size),
    power.plot.fun(data.1000.1000,size=size),
    ncol=3,nrow=3)
)
dev.off()

# power-locus curves at 1% nominal level
size<-0.01
pdf("power01.pdf",height=7,width=7)
suppressWarnings(
  grid.arrange(
    power.plot.fun(data.10.10,size=size),
    power.plot.fun(data.10.100,size=size),
    power.plot.fun(data.10.1000,size=size),
    power.plot.fun(data.100.10,size=size),
    power.plot.fun(data.1000.10,size=size),
    power.plot.fun(data.1000.100,size=size),
    power.plot.fun(data.1000.1000,size=size),
    ncol=3,nrow=3)
)
dev.off()

# power-locus curves at 10% nominal level
size<-0.1
pdf("power1.pdf",height=7,width=7)
suppressWarnings(
  grid.arrange(
    power.plot.fun(data.10.10,size=size),
    power.plot.fun(data.10.100,size=size),
    power.plot.fun(data.10.1000,size=size),
    power.plot.fun(data.100.10,size=size),
    power.plot.fun(data.1000.10,size=size),
    power.plot.fun(data.1000.100,size=size),
    power.plot.fun(data.1000.1000,size=size),
    ncol=3,nrow=3)
)
dev.off()

##### END #####
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