

```

> rm(list=ls())
> ###2
> ##d
>
> #Yes
>
> #Yes. There is an association between final grade and study amount.
> #For difference proportions, CI is [0.65946,0.92388] which is close
to 1
> #suggesting that there is a strong association
>
> #For relative risk, CI is [7.09559,15.53782] which is quite larger t
han 1
> #suggesting that there is a strong association
>
> #For odds ratio, CI is [8.17106,215.88374] which is quite larger tha
t 1
> #suggesting that there is a strong association
>
>
>
> ##e
>
>
> CI=matrix(0,nrow=100,ncol=2)
> t=0
> theta=(70*12)/(2*10)
> for(i in 1:100){
+   data=runif(500,0,1)
+   n11=0
+   n12=0
+   n21=0
+   n22=0
+   for(j in 1:500){
+     if(data[j]<70/94){
+       n11=n11+1
+     } else if(data[j]<80/94){
+       n12=n12+1
+     } else if(data[j]<82/94){
+       n21=n21+1
+     } else{
+       n22=n22+1
+     }
+   }
+   theta_hat=(n11*n22)/(n12*n21)
+   CI[i,1]=exp(log(theta_hat)-qnorm(1-(.05/2))*(1/n11+1/n12+1/n21+1/
n22)^0.5)

```

```

+   CI[i,2]=exp(log(theta_hat)+qnorm(1-(.05/2))*(1/n11+1/n12+1/n21+1/
n22)^0.5)
+   if((theta>CI[i,1])&(theta<CI[i,2])){
+     t=t+1
+   }
+ }
>
> t
[1] 96
>
> #change n = 100
>
> CI=matrix(0,nrow=100,ncol=2)
> t=0
> theta=(70*12)/(2*10)
> for(i in 1:100){
+   data=runif(100,0,1)
+   n11=0
+   n12=0
+   n21=0
+   n22=0
+   for(j in 1:100){
+     if(data[j]<70/94){
+       n11=n11+1
+     } else if(data[j]<80/94){
+       n12=n12+1
+     } else if(data[j]<82/94){
+       n21=n21+1
+     } else{
+       n22=n22+1
+     }
+   }
+   theta_hat=(n11*n22)/(n12*n21)
+   CI[i,1]=exp(log(theta_hat)-qnorm(1-(.05/2))*(1/n11+1/n12+1/n21+1/
n22)^0.5)
+   CI[i,2]=exp(log(theta_hat)+qnorm(1-(.05/2))*(1/n11+1/n12+1/n21+1/
n22)^0.5)
+   if((theta>CI[i,1])&(theta<CI[i,2])){
+     t=t+1
+   }
+ }
Error in if ((theta > CI[i, 1]) & (theta < CI[i, 2])) { :
  missing value where TRUE/FALSE needed
>
> t
[1] 1
>

```

```

> ###4
> ##a
> rm(list=ls())
>
> x=matrix(c(192,75,8,459,586,471),nrow = 2,ncol = 3,byrow=TRUE)
> colnames(x)=c("PhD", "Employed", "Unemployed")
> rownames(x)=c("Yes", "No")
> x
      PhD Employed Unemployed
Yes 192      75      8
No  459     586     471
>
> pi_hat=x/sum(x)
> pi_hat
      PhD   Employed  Unemployed
Yes 0.1072027 0.04187605 0.004466778
No  0.2562814 0.32719151 0.262981575
>
> x_marg=apply(x,1,sum)
> y_marg=apply(x,2,sum)
>
> mu_hat=(x_marg %*% t(y_marg))/sum(x)
> mu_hat
      PhD Employed Unemployed
[1,] 99.95812 101.4936  73.5483
[2,] 551.04188 559.5064 405.4517
>
> #X2 test
> X2=sum((x-mu_hat)^2/mu_hat)
> X2
[1] 177.3124
>
> qchisq(0.95,((dim(x)[1]-1)*(dim(x)[2]-1)))
[1] 5.991465
>
> #177.3124>5.991465,reject H0,which means that they are not independent
>
> chi2=chisq.test(x)
> chi2

```

Pearson's Chi-squared test

data: x

X-squared = 177.31, df = 2, p-value < 2.2e-16

```

>
> #p-value is 2.2e-16, which means that assuming they are independent,

```

```

> #we have extremely small chance to get this sample of data.
>
> #G2 test
> G2=2*sum(x*log(x/mu_hat))
> G2
[1] 197.3944
>
> qchisq(0.95,((dim(x)[1]-1)*(dim(x)[2]-1)))
[1] 5.991465
>
>
>
> #197.3944>5.991465,reject H0,which means that they are not independent
>
> library(DescTools)
> LRT=GTest(x)
> LRT

```

Log likelihood ratio (G-test) test of independence without correction

data: x  
G = 197.39, X-squared df = 2, p-value < 2.2e-16

```

>
> #p-value is 2.2e-16, which means that assuming they are independent,
> #we have extremely small chance to get this sample of data.
>
>
> ##b
>
> chisq.test(x)$stdres
      PhD  Employed Unemployed
Yes  12.54209 -3.598555  -9.70631
No   -12.54209  3.598555   9.70631
>
> ##c
> #PhD and employed
> GTest(matrix(c(192,75,459,586),nrow=2,byrow=TRUE))

```

Log likelihood ratio (G-test) test of independence without correction

data: matrix(c(192, 75, 459, 586), nrow = 2, byrow = TRUE)  
G = 68.448, X-squared df = 1, p-value < 2.2e-16

```

>

```

```
> #combined and unemployed
> GTest(matrix(c(192+75,8,459+586,471),nrow=2,byrow=TRUE))
```

Log likelihood ratio (G-test) test of independence without correction

```
data: matrix(c(192 + 75, 8, 459 + 586, 471), nrow = 2, byrow = TRUE)
G = 128.95, X-squared df = 1, p-value < 2.2e-16
```

```
>
> #Results shows that being PhD or employed is quite dependent to whether taking
> #6231 or not and being unemployed or not is also quite dependent to
> #whether taking 6231 or not.
>
>
> ###5
> ##a b
> rm(list=ls())
> data = as.data.frame(UCBAdmissions)
>
>
>
> G=matrix(0,nrow = 9,ncol = 3)
> rownames(G)=(c("AGD","AG,AD,GD","AD","GD","AG",
+               "AD,GD","AG,GD","AG,AD","A,G,D"))
> colnames(G)=(c("G2","df","p-value"))
> #model1: Most general (AGD)
> #
>
> model1=glm(Freq~Admit*Gender*Dept,data=data,family = poisson())
> G[1,]=c(model1$deviance,model1$df.residual,
+         "NULL")
>
>
> #model2: Homogeneous association (AG,AD,GD)
> #
>
> model2=glm(Freq~Admit+Gender+Dept+Admit*Gender+Admit*Dept+
+           Gender*Dept,data=data,family = poisson())
> G[2,]=c(model2$deviance,model2$df.residual,
+         1-pchisq(model2$deviance,model2$df.residual))
>
> #model3: Joint y-x,z (AD)
> #
> model3=glm(Freq~Admit+Gender+Dept+Admit*Dept,data=data,family = poisson())
> G[3,]=c(model3$deviance,model3$df.residual,
```

```

+      1-pchisq(model3$deviance,model3$df.residual))
>
> #model4: Joint x-y,z (GD)
> #
> model4=glm(Freq~Admit+Gender+Dept+Gender*Dept,data=data,family = poisson())
> G[4,]=c(model4$deviance,model4$df.residual,
+      1-pchisq(model4$deviance,model4$df.residual))
>
> #model5: Joint z-x,y (AG)
> #
> model5=glm(Freq~Admit+Gender+Dept+Admit*Gender,data=data,family = poisson())
> G[5,]=c(model5$deviance,model5$df.residual,
+      1-pchisq(model5$deviance,model5$df.residual))
>
> #model6: Conditional x,y-z (AD,GD)
> #
> model6=glm(Freq~Admit+Gender+Dept+Admit*Dept+
+      Gender*Dept,data=data,family = poisson())
> G[6,]=c(model6$deviance,model6$df.residual,
+      1-pchisq(model6$deviance,model6$df.residual))
>
> #model7: Conditional x,z-y (AG,GD)
> #
> model7=glm(Freq~Admit+Gender+Dept+Admit*Gender+
+      Gender*Dept,data=data,family = poisson())
> G[7,]=c(model7$deviance,model7$df.residual,
+      1-pchisq(model7$deviance,model7$df.residual))
>
> #model8: Conditional y,z-x (AG,AD)
> #
> model8=glm(Freq~Admit+Gender+Dept+Admit*Gender+Admit*Dept
+      ,data=data,family = poisson())
> G[8,]=c(model8$deviance,model8$df.residual,
+      1-pchisq(model8$deviance,model8$df.residual))
>
> #model9: Mutual Independent (A,G,D)
> #
> model9=glm(Freq~Admit+Gender+Dept,data=data,family = poisson())
> G[9,]=c(model9$deviance,model9$df.residual,
+      1-pchisq(model9$deviance,model9$df.residual))
> G

```

	G2	df	p-value
AGD	"1.11910480882207e-13"	"0"	"NULL"
AG,AD,GD	"20.2042753272416"	"5"	"0.00114407845139408"
AD	"1242.35030597543"	"11"	"0"
GD	"877.056413219775"	"11"	"0"

```

AG      "2004.22180522134"      "15" "0"
AD,GD   "21.7355067781329"      "6"  "0.00135199265317087"
AG,GD   "783.607006024044"      "10" "0"
AG,AD   "1148.9008987797"       "10" "0"
A,G,D   "2097.67121241707"      "16" "0"
>
> ##c
> #I would choose saturated model(model1 (AGD)), because in the table
> #all p-value is less than 0.01, which means we reject every reduced
> #model.
>
> ##d
>
> GG=matrix(0,nrow = 3,ncol = 3)
> rownames(GG)=c("AD,GD|AG,AD,GD","AG,GD|AG,AD,GD","AG,AD|AG,AD,GD")
> colnames(GG)=(c("G2","df","p-value"))
>
> GG[1,]=c(model6$deviance-model2$deviance,
+          model6$df.residual-model2$df.residual,
+          1-pchisq(model6$deviance-model2$deviance
+                  ,model6$df.residual-model2$df.residual))
>
> GG[2,]=c(model7$deviance-model2$deviance,
+          model7$df.residual-model2$df.residual,
+          1-pchisq(model7$deviance-model2$deviance
+                  ,model7$df.residual-model2$df.residual))
>
> GG[3,]=c(model8$deviance-model2$deviance,
+          model8$df.residual-model2$df.residual,
+          1-pchisq(model8$deviance-model2$deviance
+                  ,model8$df.residual-model2$df.residual))
>
> GG

```

	G2	df	p-value
AD,GD AG,AD,GD	1.531231	1	0.2159277
AG,GD AG,AD,GD	763.402731	5	0.0000000
AG,AD AG,AD,GD	1128.696623	5	0.0000000