

# threeway

*Heejung Shim*

4991 Wisconsin male high school seniors according to socio-economic status (low, lower middle, upper middle, and high), the degree of parental encouragement they receive (low and high) and whether or not they have plans to attend college (no, yes). Fienberg (1977, p. 101)

```
encouraged <- gl(2, 1, 16, labels=c("low", "high"))
soc_stratum <- gl(4, 2, 16, labels=c("lower", "lower middle", "upper middle", "higher"))
plans <- gl(2, 8, 16, labels=c("no", "yes"))
counts <- c(749, 233, 627, 330, 420, 374, 153, 266, 35, 133, 38, 303, 37, 467, 26, 800)
(wisconsin <- data.frame(counts, encouraged, soc_stratum, plans))
```

##	counts	encouraged	soc_stratum	plans
## 1	749	low	lower	no
## 2	233	high	lower	no
## 3	627	low	lower middle	no
## 4	330	high	lower middle	no
## 5	420	low	upper middle	no
## 6	374	high	upper middle	no
## 7	153	low	higher	no
## 8	266	high	higher	no
## 9	35	low	lower	yes
## 10	133	high	lower	yes
## 11	38	low	lower middle	yes
## 12	303	high	lower middle	yes
## 13	37	low	upper middle	yes
## 14	467	high	upper middle	yes
## 15	26	low	higher	yes
## 16	800	high	higher	yes

```
wt <- xtabs(counts ~ soc_stratum + encouraged + plans, wisconsin)
fable(wt)
```

##		plans	no	yes
## soc_stratum	encouraged			
## lower	low		749	35
##	high		233	133
## lower middle	low		627	38
##	high		330	303
## upper middle	low		420	37
##	high		374	467
## higher	low		153	26
##	high		266	800

a 20-year follow-up study on the effects of smoking

Appleton, French and Vanderpump (1996)

In the period 1972 - 74, a larger study categorized women into smokers and nonsmokers and according to their age group. In the follow-up, the researchers recorded whether the subjects were dead or still alive. Only smokers or women who had never smoked are presented here. Relatively few smokers quit and these women have been excluded from the data. The cause of death is not reported here. Here is the data.

```
library(faraway)
data(femsmoke)
str(femsmoke)
```

```
## 'data.frame': 28 obs. of 4 variables:
## $ y : num 2 1 3 5 14 7 27 12 51 40 ...
## $ smoker: Factor w/ 2 levels "yes","no": 1 2 1 2 1 2 1 2 1 2 ...
## $ dead : Factor w/ 2 levels "yes","no": 1 1 1 1 1 1 1 1 1 1 ...
## $ age : Factor w/ 7 levels "18-24","25-34",...: 1 1 2 2 3 3 4 4 5 5 ...
```

```
ct <- xtabs(y ~ smoker + dead + age, femsmoke)
ftable(ct)
```

```
##           age 18-24 25-34 35-44 45-54 55-64 65-74 75+
## smoker dead
## yes    yes      2      3     14     27     51     29    13
##        no      53     121    95    103     64      7     0
## no     yes      1      5      7     12     40    101    64
##        no      61     152   114     66     81     28     0
```

Question of interest is the relationship between smoking and risk of death. We can combine the data over age groups and produce:

```
(cta <- xtabs(y ~ smoker+dead, femsmoke))
```

```
##           dead
## smoker yes  no
##    yes 139 443
##    no  230 502
```

```
prop.table(cta, 1)
```

```
##           dead
## smoker    yes    no
##    yes 0.2388316 0.7611684
##    no  0.3142077 0.6857923
```

```
## pearson's chisquared test
summary(cta)
```

```
## Call: xtabs(formula = y ~ smoker + dead, data = femsmoke)
## Number of cases in table: 1314
## Number of factors: 2
## Test for independence of all factors:
##  Chisq = 9.121, df = 1, p-value = 0.002527
```

76% of smokers have survived for 20 years while only 69% of nonsmokers have survived. Thus, smoking appears to have beneficial effect on longevity. This dependency between smoking and risk of death is significant.

It seems that smoking reduce risk of death (significantly),  
but not for nearly all individual age groups (albeit insignificant).

```
cta <- xtabs(y ~ smoker+dead, femsmoke)
prop.table(cta, 1)
```

```
##      dead
## smoker    yes      no
##  yes 0.2388316 0.7611684
##   no 0.3142077 0.6857923
```

```
summary(cta)
```

```
## Call: xtabs(formula = y ~ smoker + dead, data = femsmoke)
## Number of cases in table: 1314
## Number of factors: 2
## Test for independence of all factors:
##  Chisq = 9.121, df = 1, p-value = 0.002527
```

```
ct1 <- xtabs(y ~ smoker+dead, femsmoke, subset=(age=="18-24"))
prop.table(ct1, 1)
```

```
##      dead
## smoker    yes      no
##  yes 0.03636364 0.96363636
##   no 0.01612903 0.98387097
```

```
summary(ct1)
```

```
## Call: xtabs(formula = y ~ smoker + dead, data = femsmoke, subset = (age ==
##      "18-24"))
## Number of cases in table: 117
## Number of factors: 2
## Test for independence of all factors:
##  Chisq = 0.4776, df = 1, p-value = 0.4895
##  Chi-squared approximation may be incorrect
```

```
ct2 <- xtabs(y ~ smoker+dead, femsmoke, subset=(age=="25-34"))
prop.table(ct2, 1)
```

```
##      dead
## smoker    yes      no
##   yes 0.02419355 0.97580645
##   no  0.03184713 0.96815287
```

```
summary(ct2)
```

```
## Call: xtabs(formula = y ~ smoker + dead, data = femsmoke, subset = (age ==
##      "25-34"))
## Number of cases in table: 281
## Number of factors: 2
## Test for independence of all factors:
##  Chisq = 0.14673, df = 1, p-value = 0.7017
##  Chi-squared approximation may be incorrect
```

```
ct3 <- xtabs(y ~ smoker+dead, femsmoke, subset=(age=="35-44"))
prop.table(ct3, 1)
```

```
##      dead
## smoker    yes      no
##   yes 0.12844037 0.87155963
##   no  0.05785124 0.94214876
```

```
summary(ct3)
```

```
## Call: xtabs(formula = y ~ smoker + dead, data = femsmoke, subset = (age ==
##      "35-44"))
## Number of cases in table: 230
## Number of factors: 2
## Test for independence of all factors:
##  Chisq = 3.444, df = 1, p-value = 0.06349
```

```
ct4 <- xtabs(y ~ smoker+dead, femsmoke, subset=(age=="45-54"))
prop.table(ct4, 1)
```

```
##      dead
## smoker    yes      no
##   yes 0.2076923 0.7923077
##   no  0.1538462 0.8461538
```

```
summary(ct4)
```

```
## Call: xtabs(formula = y ~ smoker + dead, data = femsmoke, subset = (age ==
##      "45-54"))
## Number of cases in table: 208
## Number of factors: 2
## Test for independence of all factors:
##  Chisq = 0.9278, df = 1, p-value = 0.3354
```

```
ct5 <- xtabs(y ~ smoker+dead, femsmoke, subset=(age=="55-64"))
prop.table(ct5, 1)
```

```
##      dead
## smoker   yes    no
##   yes 0.4434783 0.5565217
##   no  0.3305785 0.6694215
```

```
summary(ct5)
```

```
## Call: xtabs(formula = y ~ smoker + dead, data = femsmoke, subset = (age ==
##      "55-64"))
## Number of cases in table: 236
## Number of factors: 2
## Test for independence of all factors:
##  Chisq = 3.172, df = 1, p-value = 0.0749
```

```
ct6 <- xtabs(y ~ smoker+dead, femsmoke, subset=(age=="65-74"))
prop.table(ct6, 1)
```

```
##      dead
## smoker   yes    no
##   yes 0.8055556 0.1944444
##   no  0.7829457 0.2170543
```

```
summary(ct6)
```

```
## Call: xtabs(formula = y ~ smoker + dead, data = femsmoke, subset = (age ==
##      "65-74"))
## Number of cases in table: 165
## Number of factors: 2
## Test for independence of all factors:
##  Chisq = 0.08609, df = 1, p-value = 0.7692
```

```
ct7 <- xtabs(y ~ smoker+dead, femsmoke, subset=(age=="75+"))
prop.table(ct7, 1)
```

```
##      dead
## smoker yes no
##   yes   1  0
##   no   1  0
```

```
summary(ct7)
```

```
## Call: xtabs(formula = y ~ smoker + dead, data = femsmoke, subset = (age ==
##      "75+"))
## Number of cases in table: 77
## Number of factors: 2
## Test for independence of all factors:
##  Chisq = NaN, df = 1, p-value = NA
##  Chi-squared approximation may be incorrect
```

The marginal association where we add over the age groups is different from the conditional association observed within age groups.

This is an example of Simpson's paradox.

Let's see why it happens.

```
prop.table(xtabs(y ~ smoker+age, femsmoke), 2)

##      age
## smoker 18-24    25-34    35-44    45-54    55-64    65-74
##   yes 0.4700855 0.4412811 0.4739130 0.6250000 0.4872881 0.2181818
##   no  0.5299145 0.5587189 0.5260870 0.3750000 0.5127119 0.7818182
##      age
## smoker    75+
##   yes 0.1688312
##   no  0.8311688
```

We can see smokers are more concentrated in the younger age group and younger people are more likely to live for another 20 years. This explains why the marginal table gave an apparent advantage to smokers which is, in fact, not real because once we control for age, we see that smoking has a negative effect on longevity.

test if all three factors independent, clearly not!

```
modi <- glm(y ~ smoker + dead + age, femsmoke, family=poisson)
deviance(modi)
```

```
## [1] 735.0028
```

```
df.residual(modi)
```

```
## [1] 19
```

```
pchisq(735.0028, df=19, lower.tail = FALSE)
```

```
## [1] 1.362676e-143
```

test if age independent of smoking and death, clearly not!

```
modj <- glm(y ~ smoker*dead + age, femsmoke, family=poisson)
deviance(modj)
```

```
## [1] 725.8025
```

```
df.residual(modj)
```

```
## [1] 18
```

```
pchisq(725.8025, df=18, lower.tail = FALSE)
```

```
## [1] 1.889863e-142
```

are smoking and death conditionally independent given age? could be

```
modc <- glm(y ~ smoker*age + age*dead, femsmoke, family=poisson)
deviance(modc)
```

```
## [1] 8.326939
```

```
df.residual(modc)
```

```
## [1] 7
```

```
pchisq(deviance(modc), df.residual(modc), lower.tail=FALSE)
```

```
## [1] 0.3046493
```

table has some zeros which mean the chisq assumption may not hold

model comparison test is more reliable

```
modd <- glm(y ~ (smoker + dead + age)^2, femsmoke, family=poisson)
# modd <- glm(y ~ smoker*age + age*dead + smoker*dead, femsmoke, family=poisson) # equivalent
anova(modc, modd, test="Chi")
```

```
## Analysis of Deviance Table
```

```
##
```

```
## Model 1: y ~ smoker * age + age * dead
```

```
## Model 2: y ~ (smoker + dead + age)^2
```

```
##   Resid. Df Resid. Dev Df Deviance Pr(>Chi)
```

```
## 1         7      8.3269
```

```
## 2         6      2.3809  1      5.946  0.01475 *
```

```
## ---
```

```
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
# looks like smoking and death are related after all
```