

## **Ejemplo:Cigarrillo–Alcohol–Marihuana**

Los datos corresponden a un estudio realizado en la Wright State University School of Medicine and the United Health Services en Dayton Ohio, realizado a fin de relacionar el uso de alcohol (A), marihuana (M) y cigarrillos (C) en los estudiantes. Se realizó una encuesta entre los estudiantes del último año y se les preguntó si alguna vez habían consumido alguna de las tres cosas.

**Table 7.3. Alcohol (*A*), Cigarette (*C*), and Marijuana (*M*) Use for High School Seniors**

Alcohol Use	Cigarette Use	Marijuana Use	
		Yes	No
Yes	Yes	911	538
	No	44	456
No	Yes	3	43
	No	2	279

Table 7.4. Fitted Values for Loglinear Models Applied to Table 7.3

Alcohol Use	Cigarette Use	Marijuana Use	Loglinear Model				
			(A, C, M)	(AC, M)	(AM, CM)	(AC, AM, CM)	(ACM)
Yes	Yes	Yes	540.0	611.2	909.24	910.4	911
		No	740.2	837.8	438.84	538.6	538
	No	Yes	282.1	210.9	45.76	44.6	44
		No	386.7	289.1	555.16	455.4	456
No	Yes	Yes	90.6	19.4	4.76	3.6	3
		No	124.2	26.6	142.16	42.4	43
	No	Yes	47.3	118.5	0.24	1.4	2
		No	64.9	162.5	179.84	279.6	279

*(AM, CM)*

$$AC \quad 1.0 = \frac{909.24 \times 0.24}{45.76 \times 4.76} = \frac{438.84 \times 179.84}{555.16 \times 142.16}$$

marginal *AC*

$$2.7 = \frac{(909.24 + 438.84)(0.24 + 179.84)}{(45.76 + 555.16)(4.76 + 142.16)}$$

**Table 7.5. Estimated Odds Ratios for Loglinear Models in Table 7.4**

Model	Conditional Association			Marginal Association		
	<i>AC</i>	<i>AM</i>	<i>CM</i>	<i>AC</i>	<i>AM</i>	<i>CM</i>
<i>(A, C, M)</i>	1.0	1.0	1.0	1.0	1.0	1.0
<i>(AC, M)</i>	17.7	1.0	1.0	17.7	1.0	1.0
<i>(AM, CM)</i>	1.0	61.9	25.1	2.7	61.9	25.1
<i>(AC, AM, CM)</i>	7.8	19.8	17.3	17.7	61.9	25.1
<i>(ACM) level 1</i>	13.8	24.3	17.5	17.7	61.9	25.1
<i>(ACM) level 2</i>	7.7	13.5	9.7			

**Table 7.7. Goodness-of-Fit Tests for Loglinear Models Relating Alcohol (*A*), Cigarette (*C*), and Marijuana (*M*) Use**

Model	$G^2$	$\chi^2$	$df$	$P$ -value*
( <i>A, C, M</i> )	1286.0	1411.4	4	<0.001
( <i>A, CM</i> )	534.2	505.6	3	<0.001
( <i>C, AM</i> )	939.6	824.2	3	<0.001
( <i>M, AC</i> )	843.8	704.9	3	<0.001
( <i>AC, AM</i> )	497.4	443.8	2	<0.001
( <i>AC, CM</i> )	92.0	80.8	2	<0.001
( <i>AM, CM</i> )	187.8	177.6	2	<0.001
( <i>AC, AM, CM</i> )	0.4	0.4	1	0.54
( <i>ACM</i> )	0.0	0.0	0	—

\*  $P$ -value for  $G^2$  statistic.

## Planteamos primero el modelo (A,C,M)

```
* A = Alcohol
* C = Cigarrillo
* M = Marihuana
man 3
dim 2 2 2
lab A C M
mod {A,C,M}
dat [ 911 538
      44 456
      3 43
      2 279]
```

Number of iterations = 2  
Converge criterion = 0.0000000000  
X-squared = 1411.3860 (0.0000)  
L-squared = 1286.0200 (0.0000)  
Cressie-Read = 1316.2009 (0.0000)  
Dissimilarity index = 0.2875  
Degrees of freedom = 4  
Log-likelihood = -3950.08844  
Number of parameters = 3 (+1)  
Sample size = 2276.0  
BIC(L-squared) = 1255.0993  
AIC(L-squared) = 1278.0200  
BIC(log-likelihood) = 7923.3674  
AIC(log-likelihood) = 7906.1769  
Eigenvalues information matrix  
2220.4205 2051.8191 1119.8361

\*\*\* FREQUENCIES \*\*\*

A	C	M	observed	estimated	std. res.
1	1	1	911.000	539.983	15.966
1	1	2	538.000	740.226	-7.433
1	2	1	44.000	282.091	-14.176
1	2	2	456.000	386.700	3.524
2	1	1	3.000	90.597	-9.203
2	1	2	43.000	124.194	-7.286
2	2	1	2.000	47.329	-6.589
2	2	2	279.000	64.880	26.583



\*\*\* LOG-LINEAR PARAMETERS \*\*\*

\* TABLE ACM [or P(ACM)] \*

effect	beta	std err	z-value	exp(beta)	Wald	df	prob
main	5.2320			187.1738			
A							
1	0.8926	0.0299	29.868	2.4414			
2	-0.8926			0.4096	892.12	1	0.000
C							
1	0.3247	0.0221	14.706	1.3836			
2	-0.3247			0.7228	216.26	1	0.000
M							
1	-0.1577	0.0212	-7.431	0.8541			
2	0.1577			1.1708	55.23	1	0.000

\*\*\* (CONDITIONAL) PROBABILITIES \*\*\*

\* P(ACM) \*

1	1	1	0.2373	(0.0071)
1	1	2	0.3252	(0.0081)
1	2	1	0.1239	(0.0048)
1	2	2	0.1699	(0.0060)
2	1	1	0.0398	(0.0023)
2	1	2	0.0546	(0.0031)
2	2	1	0.0208	(0.0013)
2	2	2	0.0285	(0.0018)

## **Probamos con (AC,AM,CM)**

man 3

dim 2 2 2

lab A C M

mod {AC,AM,CM}

dat [ 911 538  
      44 456  
      3 43  
      2 279]

```

Number of iterations = 6
  Converge criterion   = 0.0000000246
  X-squared           = 0.4011 (0.5265)
  L-squared           = 0.3740 (0.5408)
  Cressie-Read        = 0.3913 (0.5316)
  Dissimilarity index = 0.0011
  Degrees of freedom  = 1
  Log-likelihood       = -3307.26545
  Number of parameters = 6 (+1)
  Sample size         = 2276.0
  BIC(L-squared)      = -7.3562
  AIC(L-squared)      = -1.6260
  BIC(log-likelihood) = 6660.9120
  AIC(log-likelihood) = 6626.5309
  Eigenvalues information matrix
5310.35  3346.091  1750.952  500.779  176.449
26.072

```

\*\*\* FREQUENCIES \*\*\*

A	C	M	observed	estimated	std. res.
1	1	1	911.000	910.383	0.020
1	1	2	538.000	538.617	-0.027
1	2	1	44.000	44.617	-0.092
1	2	2	456.000	455.383	0.029
2	1	1	3.000	3.617	-0.324
2	1	2	43.000	42.383	0.095
2	2	1	2.000	1.383	0.524
2	2	2	279.000	279.617	-0.037

\*\*\* LOG-LINEAR PARAMETERS \*\*\*

\* TABLE ACM [or P(ACM)] \*

effect	beta	std err	z-value	exp(beta)	Wald	df	prob
main	4.2515			70.2130			
A							
1	1.5040	0.1138	13.215	4.4996			
2	-1.5040			0.2222	174.64	1	0.000
C							
1	0.2823	0.0549	5.140	1.3261			
2	-0.2823			0.7541	26.42	1	0.000
M							
1	-1.1960	0.1185	-10.093	0.3024			
2	1.1960			3.3070	101.87	1	0.000
AC							
1 1	0.5136	0.0435	11.802	1.6714			
1 2	-0.5136			0.5983			
2 1	-0.5136			0.5983			
2 2	0.5136			1.6714	139.30	1	0.000
AM							
1 1	0.7465	0.1162	6.426	2.1096			
1 2	-0.7465			0.4740			
2 1	-0.7465			0.4740			
2 2	0.7465			2.1096	41.29	1	0.000
CM							
1 1	0.7120	0.0410	17.381	2.0380			
1 2	-0.7120			0.4907			
2 1	-0.7120			0.4907			
2 2	0.7120			2.0380	302.10	1	0.000

\*\*\* (CONDITIONAL) PROBABILITIES \*\*\*

\* P(ACM) \*

1 1 1	0.4000	(0.0103)
1 1 2	0.2367	(0.0089)
1 2 1	0.0196	(0.0029)
1 2 2	0.2001	(0.0084)
2 1 1	0.0016	(0.0007)
2 1 2	0.0186	(0.0028)
2 2 1	0.0006	(0.0003)
2 2 2	0.1229	(0.0069)

## Probamos de quitar AM

- \* A = Alcohol
- \* C = Cigarrillo
- \* M = Marihuana

```
man 3
dim 2 2 2
lab A C M
mod {AC,CM}
dat [ 911 538
      44 456
      3 43
      2 279]
```



Number of iterations = 2  
Converge criterion = 0.0000000000  
X-squared = 80.8148 (0.0000)  
L-squared = 92.0184 (0.0000)  
Cressie-Read = 82.0368 (0.0000)  
Dissimilarity index = 0.0349  
Degrees of freedom = 2  
Log-likelihood = -3353.08764  
Number of parameters = 5 (+1)  
Sample size = 2276.0  
BIC(L-squared) = 76.5580  
AIC(L-squared) = 88.0184  
BIC(log-likelihood) = 6744.8262  
AIC(log-likelihood) = 6716.1753  
Eigenvalues information matrix  
4190.794 2743.776 1399.651 351.217 183.293

## Ahora quitamos AC

\* A = Alcohol

\* C = Cigarrillo

\* M = Marihuana

man 3

dim 2 2 2

lab A C M

mod {AM,CM}

dat [ 911 538  
      44 456  
      3 43  
      2 279]

```

Number of iterations = 2
  Converge criterion = 0.000000000000
  X-squared          = 177.6149 (0.0000)
  L-squared          = 187.7543 (0.0000)
  Cressie-Read      = 178.1090 (0.0000)
  Dissimilarity index = 0.0887
  Degrees of freedom = 2
  Log-likelihood     = -3400.95561
  Number of parameters = 5 (+1)
  Sample size       = 2276.0
  BIC(L-squared)    = 172.2940
  AIC(L-squared)    = 183.7543
  BIC(log-likelihood) = 6840.5621
  AIC(log-likelihood) = 6811.9112
  Eigenvalues information matrix
    4814.082  2582.42  1755.695  272.398  26.286

```

## Ahora quitamos CM

- \* A = Alcohol
- \* C = Cigarrillo
- \* M = Marihuana

```
man 3
dim 2 2 2
lab A C M
mod {AC,AM}
dat [ 911 538
      44 456
      3 43
      2 279 ]
```

```

Number of iterations = 2
  Converge criterion = 0.000000000000
  X-squared          = 443.7611 (0.0000)
  L-squared          = 497.3693 (0.0000)
  Cressie-Read       = 453.7193 (0.0000)
  Dissimilarity index = 0.1786
  Degrees of freedom = 2
  Log-likelihood      = -3555.76310
  Number of parameters = 5 (+1)
  Sample size         = 2276.0
  BIC(L-squared)      = 481.9089
  AIC(L-squared)      = 493.3693
  BIC(log-likelihood) = 7150.1771
  AIC(log-likelihood) = 7121.5262
  Eigenvalues information matrix
3897.094  3302.761  1803.663  263.962  26.369

```

Ninguno de los 3 modelos (AC,AM) (AC,CM)  
(AM,CM) ajusta bien  
Nos quedamos entonces con (AC,AM,CM)