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Exercises

1. log-linear regression models

$$\mathbb{P}(y_i | \alpha, \beta, \mathbf{x}_i) \sim \mu_i^{y_i} \exp(-\mu_i)$$
$$\mu(\mathbf{x}) = \exp\left(\alpha + \sum_j \beta^{(j)} x^{(j)}\right)$$
$$\log(\mu(\mathbf{x})) = \alpha + \sum_j \beta^{(j)} x^{(j)}$$

data

1	y_1	$x_1^{(1)}$...	$x_1^{(k)}$
\vdots				
n	y_n	$x_n^{(1)}$		$x_n^{(k)}$

Likelihood:

$$\log L(\alpha, \beta | y_i, \mathbf{x}_i) = \sum_{i=1}^n y_i \left(\alpha + \sum_{j=1}^k \beta^{(j)} x_i^{(j)} \right) - \exp\left(\alpha + \sum_{j=1}^k \beta^{(j)} x_i^{(j)} \right)$$

log-linear models and contingency tables

for the following data:

age-group	cat1($y^{(1)}$)	cat2($y^{(2)}$)	cat3($y^{(3)}$)
1	$y_1^{(1)}$	$y_1^{(2)}$	$y_1^{(3)}$
2	$y_2^{(1)}$	$y_2^{(2)}$	$y_2^{(3)}$
\vdots			
I	$y_I^{(1)}$	$y_I^{(2)}$	$y_I^{(3)}$

- specify Poisson reg model and interpret the coefficients
- write likelihood function

- c) write likelihood-ratio test for hypothesis of age-group independence of cat
 - d) discuss equivalence of the log-linear and multinomial models for the given grouped data
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2. three-dimensional contingency tables and log-linear models

- a) specify a 3-dimensional contingency table in (group) variables (factors) R, C, L and write an equivalent (full) log-linear model
 - b) interpret triple interaction as differences between pairwise interactions of any two variables at different levels of the third variable
 - c) interpret the pairwise interactions in terms of (conditional) odds
 - d) describe the LR-test for the hypothesis that triple interactions are negligible.
 - in the homogenous odds ratio model:
 - e) interpret the pairwise interactions in terms of odds.
 - f) discuss the relation between pairwise interactions and conditional independence properties.
 - g) discuss the relations between the reduced model and a multinomial regression of first on second and third factors (hint: Table 7.2)
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- 3. find an example to show

$$\mathbb{P}(x | y, z) = \mathbb{P}(x | z) \not\Rightarrow \mathbb{P}(z | y, x) = \mathbb{P}(z | x)$$

- 4. define

$$OR(X | Y) = \frac{\mathbb{P}(X = x_{i_1} | y_{j_1}) / \mathbb{P}(X = x_{i_2} | y_{j_1})}{\mathbb{P}(X = x_{i_1} | y_{j_2}) / \mathbb{P}(X = x_{i_2} | y_{j_2})}$$

show

$$OR(X | Y) = OR(Y | X)$$

5. define

$$\begin{aligned} OR(X | Y, Z) &= \frac{\mathbb{P}(X = x_{i_1} | y_{j_1}, z_{k_1}) / \mathbb{P}(X = x_{i_2} | y_{j_1}, z_{k_1})}{\mathbb{P}(X = x_{i_1} | y_{j_1}, z_{k_1}) / \mathbb{P}(X = x_{i_2} | y_{j_2}, z_{k_2})} = \frac{p_{i_1 | j_1 k_1} / p_{i_2 | j_1 k_1}}{p_{i_1 | j_2 k_2} / p_{i_2 | j_2 k_2}} \\ &= \frac{p_{i_1 j_1 k_1} / p_{i_2 j_1 k_1}}{p_{i_1 j_2 k_2} / p_{i_2 j_2 k_2}} \end{aligned}$$

note that in loglinear models with only pairwise interactions:

$$\begin{aligned} \log OR(X | Y, Z) &= \left(\alpha_{i_1} + \beta_{j_1} + \gamma_{k_1} + (\alpha\beta)_{i_1 j_1} + (\alpha\gamma)_{i_1 k_1} + (\beta\gamma)_{j_1 k_1} \right) \\ &\quad - \left(\alpha_{i_2} + \beta_{j_1} + \gamma_{k_1} + (\alpha\beta)_{i_2 j_1} + (\alpha\gamma)_{i_2 k_1} + (\beta\gamma)_{j_1 k_1} \right) \\ &\quad - \left(\alpha_{i_1} + \beta_{j_2} + \gamma_{k_2} + (\alpha\beta)_{i_1 j_2} + (\alpha\gamma)_{i_1 k_2} + (\beta\gamma)_{j_2 k_2} \right) \\ &\quad + \left(\alpha_{i_2} + \beta_{j_2} + \gamma_{k_2} + (\alpha\beta)_{i_2 j_2} + (\alpha\gamma)_{i_2 k_2} + (\beta\gamma)_{j_2 k_2} \right) \\ &= \left((\alpha\beta)_{i_1 j_1} + (\alpha\gamma)_{i_1 k_1} + (\beta\gamma)_{j_1 k_1} \right) - \left((\alpha\beta)_{i_2 j_1} + (\alpha\gamma)_{i_2 k_1} + (\beta\gamma)_{j_1 k_1} \right) \\ &\quad - \left((\alpha\beta)_{i_1 j_2} + (\alpha\gamma)_{i_1 k_2} + (\beta\gamma)_{j_2 k_2} \right) + \left((\alpha\beta)_{i_2 j_2} + (\alpha\gamma)_{i_2 k_2} + (\beta\gamma)_{j_2 k_2} \right) \\ &= \left((\alpha\beta)_{i_1 j_1} + (\alpha\gamma)_{i_1 k_1} \right) - \left((\alpha\beta)_{i_2 j_1} + (\alpha\gamma)_{i_2 k_1} \right) \\ &\quad - \left((\alpha\beta)_{i_1 j_2} + (\alpha\gamma)_{i_1 k_2} \right) + \left((\alpha\beta)_{i_2 j_2} + (\alpha\gamma)_{i_2 k_2} \right) \end{aligned}$$

6. if $(X \perp Y) | Z \implies OR(X | Y, Z) = OR(X | Z)$

- a) by symmetry find the corresponding OR property for $(Z \perp Y) | X$
- b) show that if $(X \perp Y) | Z$ or $(Z \perp Y) | X$ the X, Y, Z table is collapsible in Y .

3. corresponding to the data

data	A	B	C	D
cat1	42	44	40	126
cat2	33	133	43	37
cat3	37	43	127	46
cat4	48	42	38	121

interpret the following outputs:

a) model

Coefficients:

	Estimate	Std. Error	z value	Pr(> z)	
(Intercept)	3.68888	0.07906	46.661	< 2e-16	***
xB	0.49317	0.10033	4.915	8.86e-07	***
xC	0.43825	0.10140	4.322	1.55e-05	***
xD	0.72392	0.09633	7.515	5.71e-14	***\bigskip

b) model

Coefficients:

	Estimate	Std. Error	z value	Pr(> z)	
(Intercept)	3.69685	0.09601	38.504	< 2e-16	***
xB	0.49317	0.10033	4.915	8.86e-07	***
xC	0.43826	0.10140	4.322	1.55e-05	***
xD	0.72392	0.09633	7.515	5.71e-14	***
ycat2	-0.02410	0.08963	-0.269	0.788	
ycat3	0.00396	0.08900	0.044	0.965	
ycat4	-0.01198	0.08936	-0.134	0.893	\bigskip

c) model

Coefficients:

	Estimate	Std. Error	z value	Pr(> z)	
(Intercept)	3.73767	0.15430	24.223	< 2e-16	***
xB	0.04652	0.21572	0.216	0.829264	
xC	-0.04879	0.22093	-0.221	0.825216	
xD	1.09861	0.17817	6.166	7.01e-10	***
ycat2	-0.24116	0.23262	-1.037	0.299868	
ycat3	-0.12675	0.22547	-0.562	0.574002	
ycat4	0.13353	0.21129	0.632	0.527396	
xB:ycat2	1.34732	0.29045	4.639	3.50e-06	***
xC:ycat2	0.31348	0.31995	0.980	0.327192	
xD:ycat2	-0.98420	0.29846	-3.298	0.000975	***
xB:ycat3	0.10376	0.31116	0.333	0.738779	
xC:ycat3	1.28206	0.28933	4.431	9.37e-06	***
xD:ycat3	-0.88089	0.28375	-3.104	0.001906	**
xB:ycat4	-0.18005	0.30196	-0.596	0.550991	
xC:ycat4	-0.18482	0.30977	-0.597	0.550743	
xD:ycat4	-0.17402	0.24667	-0.706	0.480498	
