

# Data Mining 2020

## Counting $u$ -terms in graphical log-linear models

Being able to count  $u$ -terms is important for

1. determining the correct degrees of freedom for statistical tests, and
2. computing the correct complexity penalty for a model's AIC or BIC score.

Let  $X = (X_1, X_2, \dots, X_k)$  denote a vector of  $k$  discrete random variables. Furthermore, let  $d_i$  denote the size of the domain of  $X_i$ . For example, if  $X_i$  corresponds to the variable "Eye Color", with possible values {blue, brown, green, other}, then  $d_i = 4$ .

The full log-linear expansion of the probability distribution  $P_K$  of  $(X_1, X_2, \dots, X_k)$  is given by

$$\log P_K(x) = \sum_{a \subseteq K} u_a(x_a)$$

where the sum is taken over all possible subsets  $a$  of  $K = \{1, 2, \dots, k\}$ .

The total number of possible value assignments to  $(X_1, X_2, \dots, X_k)$  (the number of cells in the contingency table) is:

$$\text{ncell} = \prod_{i=1}^k d_i$$

The number of  $u$ -terms in the full log-linear expansion (the saturated model) is equal to ncell. It can alternatively be computed by the formula

$$\sum_{a \subseteq K} \left( \prod_{i \in a} (d_i - 1) \right),$$

where

$$\prod_{i \in \emptyset} (d_i - 1)$$

is taken to be equal to 1, corresponding to the constant  $u$ -term  $u_\emptyset$ . So apparently

$$\prod_{i=1}^k d_i = \sum_{a \subseteq K} \left( \prod_{i \in a} (d_i - 1) \right).$$

In general, a collection of  $u$ -terms  $u_a(x_a)$  contains

$$\prod_{i \in a} (d_i - 1)$$

$u$ -terms.

Consider as an example the random vector  $X = (X_1, X_2, X_3, X_4)$  with  $d_1 = 2$ ,  $d_2 = 3$ ,  $d_3 = 5$ , and  $d_4 = 2$ . We compute that

$$\text{ncell} = \prod_{i=1}^k d_i = 2 \times 3 \times 5 \times 2 = 60.$$

How many  $u$ -terms are there in the graphical model with the chordless 4-cycle (1–2–3–4–1) as its independence graph? Recall the following definition:

“given its independence graph  $G = (K, E)$ , the log-linear model for the random vector  $X$  is a *graphical model* for  $X$  if the distribution of  $X$  is *arbitrary* apart from constraints of the form that for all pairs of coordinates not in the edge set  $E$ , the  $u$ -terms containing the selected coordinates are equal to zero”.

The following edges are absent in the chordless 4-cycle: 1 – 3, and 2 – 4. Because of the absence of 1 – 3 we have to set the following collections  $u$ -terms to zero:  $u_{13}$ ,  $u_{123}$ ,  $u_{134}$  and  $u_{1234}$ . Likewise, because of the absence of 2 – 4 we have to set the following collections  $u$ -terms to zero:  $u_{24}$ ,  $u_{124}$ ,  $u_{234}$  and  $u_{1234}$ . The total number of  $u$ -terms that has to be removed is counted in the table below.

$u_a$	$\prod_{i \in a} (d_i - 1)$
$u_{13}$	$1 \times 4 = 4$
$u_{123}$	$1 \times 2 \times 4 = 8$
$u_{134}$	$1 \times 4 \times 1 = 4$
$u_{1234}$	$1 \times 2 \times 4 \times 1 = 8$
$u_{24}$	$2 \times 1 = 2$
$u_{124}$	$1 \times 2 \times 1 = 2$
$u_{234}$	$2 \times 4 \times 1 = 8$
Total	36

So to test the chordless 4-cycle against the saturated model, we have to use a chi-square distribution with 36 degrees of freedom. Also, to compute the AIC or BIC score, we should use  $\dim(M) = 60 - 36 = 24$ .

Alternatively, we could directly have counted the  $u$ -terms present in the model. These are:

$u_a$	$\prod_{i \in a} (d_i - 1)$
$u_\emptyset$	1
$u_1$	1
$u_2$	2
$u_3$	4
$u_4$	1
$u_{12}$	$1 \times 2 = 2$
$u_{23}$	$2 \times 4 = 8$
$u_{34}$	$4 \times 1 = 4$
$u_{14}$	$1 \times 1 = 1$
Total	24