# Markov bases and subbases for bounded contingency tables 

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| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | Serum Cholesterol $(\mathrm{mg} / 100 \mathrm{ml})$ |  |  |  |  |  |  |  |
|  | Blood | 1 | 2 | 3 | 4 | 5 | 6 | 7 |  |
|  | Pressure | $<200$ | $200-209$ | $210-219$ | $220-244$ | $245-259$ | $260-284$ | $>284$ |  |
| 1 | $<117$ | $2 / 53$ | $0 / 21$ | $0 / 15$ | $0 / 20$ | $0 / 14$ | $1 / 22$ | $0 / 11$ |  |
| 2 | $117-126$ | $0 / 66$ | $2 / 27$ | $1 / 25$ | $8 / 69$ | $0 / 24$ | $5 / 22$ | $1 / 19$ |  |
| 3 | $127-136$ | $2 / 59$ | $0 / 34$ | $2 / 21$ | $2 / 83$ | $0 / 33$ | $2 / 26$ | $4 / 28$ |  |
| 4 | $137-146$ | $1 / 65$ | $0 / 19$ | $0 / 26$ | $6 / 81$ | $3 / 23$ | $2 / 34$ | $4 / 23$ |  |
| 5 | $147-156$ | $2 / 37$ | $0 / 16$ | $0 / 6$ | $3 / 29$ | $2 / 19$ | $4 / 16$ | $1 / 16$ |  |
| 6 | $157-166$ | $1 / 13$ | $0 / 10$ | $0 / 11$ | $1 / 15$ | $0 / 11$ | $2 / 13$ | $4 / 12$ |  |
| 7 | $167-186$ | $3 / 21$ | $0 / 5$ | $0 / 11$ | $2 / 27$ | $2 / 5$ | $6 / 16$ | $3 / 14$ |  |
| 8 | $>186$ | $1 / 5$ | $0 / 1$ | $3 / 6$ | $1 / 10$ | $1 / 7$ | $1 / 7$ | $1 / 7$ |  |

Source : [Cornfield, 1962]
Data on coronary heart disease incidence in Framingham, Massachusetts [Cornfield, 1962, Agresti, 1990]. A sample of male residents, aged 40 through 50, were classified on blood pressure and serum cholesterol concentration. $2 / 53$ in the $(1,1)$ cell means that there are 53 cases, of whom 2 exhibited heart disease.

## Imcomplete contingency table

Table 1: Effects of decision alternatives on the verdicts and social perceptions of simulated jurors.

|  | Condition |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Alternative | 1 | 2 | 3 | 4 | 5 | 6 | 7 |
| First degree | 11 | $[0]$ | $[0]$ | 2 | 7 | $[0]$ | 2 |
| Second degree | $[0]$ | 20 | $[0]$ | 22 | $[0]$ | 11 | 15 |
| Manslaughter | $[0]$ | $[0]$ | 22 | $[0]$ | 16 | 13 | 5 |
| Not guilty | 13 | 4 | 2 | 0 | 1 | 0 | 2 |

Source: [Vidmar, 1972]
This table refers to the possible effects on decision making of limiting the number of alternatives available to the number of a jury panel.
[0] refers to the structural zero on the cell.

## Independence model

Let $\mathbf{X}=\left\{X_{i j}\right\}$ be a $I \times J$ table $X_{i j} \in \mathbb{N}, i=1, \ldots, I, j=1, \ldots, J$.
An observed table $X^{o b s}=\left\{x_{i j}^{o b s}\right\}, x_{i j}^{o b s} \in \mathbb{N}$, and $1 \leq I, 1 \leq J$.

$$
X_{i j} \sim \operatorname{Poi}\left(\mu_{i j}\right) \mathrm{iid}
$$

where $\mu_{i j}=\ln \left(\theta_{i j}\right)$.
Consider the generalized linear model with a canonical linear predictor of the form:

$$
\theta_{i j}=\lambda+\lambda_{i}^{R}+\lambda_{j}^{C}+\lambda_{i j}^{R C}
$$

for $i=1, \ldots, I$ and $j=1, \ldots, J$.
Independence model is a special case such that

$$
\lambda_{i j}^{R C}=0 \text { for } 1 \leq i \leq I, 1 \leq j \leq J
$$

## Hypothesis

The sufficient statistics for independence model include the row and column margins. Hence, the conditional distribution of the table counts given the margins is the same regardless of the values of the parameters in the model.

We have the following hypothesis test:

$$
\begin{gathered}
H_{0}: \lambda_{i j}^{R C}=0 \text { no interaction. } \\
H_{1}: \lambda_{i j}^{R C} \text { not constant over all cells. }
\end{gathered}
$$

## Exact p-value computation

Let $\hat{\mathbf{X}}$ be the MLE of the data under the model. Then Pearson's $\chi^{2}$ statistics is

$$
f(X)=\sum_{i=1}^{I} \sum_{j=1}^{J} \frac{\left(\hat{X}_{i j}-X_{i j}\right)^{2}}{\hat{X}_{i j}}
$$

An exact permutation test based on the $\chi^{2}$ statistic is constructed as follows. The $p$-value of this test is:

$$
p=E_{\mathbf{p}}\left[I_{\{f(\mathbf{X}) \geq f(\mathbf{x})\}} \mid \text { satisfying margins }\right]
$$

where $\mathbf{x}$ is an observed table and $\mathbf{p}$ is the hypergeometric distribution.

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In general we approximate the expected value by generating random draws from the hypergeometric distribution and estimate

$$
\hat{p}=\frac{1}{N} \sum_{i=1}^{N} I_{\left\{f\left(\mathrm{x}^{i}\right) \geq f(\mathrm{x})\right\}}
$$

where $N$ is the number of draws $\mathbf{x}^{1}, \cdots, \mathbf{x}^{N}$ iid from the hypergoemetric conditional on the sufficient statistics under $H_{0}$.

Note: This is the only possible method in situations where counts are very small or the number of tables satisfying margins is very small.

Question: How can we generate random draws from this distribution?
Answer: Apply Diaconis-Sturmfels algorithm to the MCMC technique. Diaconis-Sturmfels algorithm is the only method guaranteed to connect the MC.

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What is a set of moves which connect all feasible contingency tables satisfying these margins?

For unbounded tables under independence model, we know the set of moves which connect all feasible contingency tables satisfying margins.

Note: We can generalize this problem by adding a bound for each cell of a table in addition to row and column sums.

Note: If some of the bounds are zeros, then it is a incomplete table, i.e., table with structural zeros.

Question 1: Finding a set of moves which connect all feasible bounded 2-way contingency tables satisfying the row sums and column sums.

Question 2: If we know these bounds are non-zero, i.e., it is not an incomplete table, then what is a set of moves connect all feasible bounded 2-way contingency tables satisfying the row sums and column sums?

## Exact p-value computation

Note that the row sums and column sums are the sufficient statistics under $H_{0}$. For example, we have

|  |  |  |  | Total |
| :---: | :---: | :---: | :---: | :---: |
|  | $x_{1,1}$ | $x_{1,2}$ | $x_{1,3}$ | 6 |
|  | $x_{2,1}$ | $x_{2,2}$ | $x_{2,3}$ | 6 |
| Total | 4 | 4 | 4 |  |

and each cell is bounded by 2 , i.e., $x_{i, j} \leq 2$ for $i=1,2$ and $j=1,2,3$.

From the constraints we can set up the system of linear equations and inequalities.
e.g. For our $2 \times 3$ table, we have:
where $Z_{+}=\{0,1,2, \cdots\}$.

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By introducing slack variables we have the system of equations.

$$
\begin{array}{rllllll}
x_{1,1} & & & +x_{2,1} & & & \\
& x_{1,2} & & & +x_{2,2} & & =4 \\
& & x_{1,3} & & & +x_{2,3} & =4 \\
x_{1,1}+x_{1,2} & +x_{1,3} & & & & =6 \\
& & & x_{2,1} & +x_{2,2} & +x_{2,3} & = \\
& & & x_{i, j} & +y_{i, j} & = & 2 \\
& & & & & x_{i, j} & \in \mathbb{Z}_{+}
\end{array}
$$

This is equivalent with $2 \times 3 \times 2$ tables with constraints $[A, C],[B, C]$, $[A, B]$ for factors $A, B, C$, which would arise for example in case-control data with two factors $A$ and $B$ at three levels each.

In general, we can set up a system $\left\{x \in \mathbb{Z}_{+}^{d} \mid A x=b\right\}$ for any tables.
Note: Thus, moves connect all integral points inside a feasible region $P_{b}=\left\{x \in \mathbb{R}^{d} \mid A x=b, x \geq 0\right\} \neq \emptyset$.

## What is a Markov Basis??

Suppose $P_{b}=\left\{x \in \mathbb{R}^{d} \mid A x=b, x \geq 0\right\} \neq \emptyset$ and let $M$ be a finite set such that $M \subset\left\{x \in \mathbb{Z}^{d} \mid A x=0\right\}$.

We define the graph $G_{b}$ such that:

- Nodes of $G_{b}$ are the lattice points inside $P_{b}$.
- We draw an undirected edge between a node $u$ and a node $v$ iff $u-v \in M$.

Definition : $M$ is called a Markov basis if $G_{b}$ is a connected graph for all $b$ with $P_{b} \neq \emptyset$.

Why do we care?: A Markov basis is the only known set of moves which guarantees to connect all tables with any constraints.

## Example

To make it simple we just removed bounds.

|  |  |  |  | Total |
| :---: | :---: | :---: | :---: | :---: |
|  | ? ? ? | ? ? ? | ? ? ? | 6 |
|  | ? ? ? | ? ? ? | ? ? ? | 6 |
| Total | 4 | 4 | 4 |  |

Table 2: $2 \times 3$ tables with 1 -marginals.

There are 19 tables satisfying these margins. We counted using a software LattE.


$\pm$| 1 | 0 | -1 |
| :---: | :---: | :---: |
| -1 | 0 | 1 |

There are 3 elements in a Markov basis modulo signs.
In fact such moves are called basic moves.

Note: A Gröbner basis of a toric idea $\mathcal{I}_{A}$ associate to a matrix $A$ with any term order is a Markov basis associate to a matrix $A$. So one can compute a Markov basis from a Gröbner basis of $\mathcal{I}_{A}$ with any term order.

Note: There are several nice software to compute Gröbner bases (such as 4ti2).

However: Computing a Gröbner basis is very hard in general.

## Notation

Without loss of generality, we represent a table by a vector of counts $\mathbf{n}=\left(n_{1}, \ldots, n_{k}\right)$. Let $\mathcal{X}=\{1, \ldots, k\}$ be the sample space of the contingency table. In the special case of two-way $I \times J$ tables, we will also denote the sample space with $\mathcal{X}=\{1, \ldots, I\} \times\{1, \ldots, J\}$.

The fiber of an observed table $\mathbf{n}_{\text {obs }}$ with respect to a function $T: \mathbb{N}^{k} \longrightarrow \mathbb{N}^{s}$ is the set

$$
\mathcal{F}_{T}\left(\mathbf{n}_{\mathrm{obs}}\right)=\left\{\mathbf{n} \mid \mathbf{n} \in \mathbb{N}^{k}, T(\mathbf{n})=T\left(\mathbf{n}_{\mathrm{obs}}\right)\right\}
$$

When the dependence on the specific observed table is irrelevant, we will write simply $\mathcal{F}_{T}$ instead of $\mathcal{F}_{T}\left(\mathbf{n}_{\text {obs }}\right)$.

In mathematical statistics framework, the function $T$ is usually the minimal sufficient statistic of some statistical model.

Definition: A Universal Gröbner basis of an ideal is the Gröbner basis with respect to every term order.

Let a $s \times k$-matrix $A_{T}$ be a configuration of $T$ and $\mathcal{I}_{A_{T}}$ be a toric ideal associate with $A_{T}$.

Theorem [Rapallo and Rogantin, 2007] A Universal Gröbner basis of the toric ideal $\mathcal{I}_{A_{T}}$ is a Markov basis of bounded tables under the given model.

If we know a Universal Gröbner basis for $A_{T}$, then we can compute a MB for incomplete tables.

## Computing a MB for incomplete tables

Let $\mathcal{X}_{0} \subset \mathcal{X}$ be the set of structural zeros of the table, let $T^{\prime}$ be the function $T$ restricted to $\mathcal{X}^{\prime}=\mathcal{X} \backslash \mathcal{X}_{0}$ and let $\mathcal{I}_{A_{T}}^{\prime}$ be the toric ideal associated with $A_{T^{\prime}}$

Definition A Markov basis computed through a Universal Gröbner basis is a Universal Markov basis.

Theorem [Rapallo and $\mathrm{Y} ., 2009$ ] Let $\mathbf{n}$ be a contingency table and let $\mathcal{F}_{T}^{\mathrm{b}}$ be its bounded fiber under the bound $\mathbf{n} \leq \mathbf{b}$. Let $\mathcal{X}_{0}$ be the set of structural zeros. Then a Universal Markov basis for $\mathcal{F}_{T^{\prime}}^{b}$ is obtained from a Universal Markov basis for $\mathcal{F}_{T}^{b}$ by removing the moves involving the cells in $\mathcal{X}_{0}$.

## Example

Let us consider $4 \times 4$ contingency tables with fixed marginal totals. Without structural zeros, the Universal Markov basis is formed by 204 binomials: 36 moves involving 4 cells: 96 moves involving 6 cells: and 72 moves involving 8 cells.

Suppose that the cell $(1,1)$ is a structural zero. This kind of table is depicted below, where [0] means a structural zero, while the symbol • denotes a non-zero cell.

$$
\left(\begin{array}{cccc}
{[0]} & \bullet & \bullet & \bullet \\
\bullet & \bullet & \bullet & \bullet \\
\bullet & \bullet & \bullet & \bullet \\
\bullet & \bullet & \bullet & \bullet
\end{array}\right)
$$

Applying Theorem, we remove: 9 moves involving 4 cells: 36 moves involving 6 cells: and 36 moves involving 8 cells. The Universal Markov basis in this case has 123 moves.

## Example cont...

Suppose now that the whole main diagonal contains structural zeros, as in the figure below.

$$
\left(\begin{array}{cccc}
{[0]} & \bullet & \bullet & \bullet \\
\bullet & {[0]} & \bullet & \bullet \\
\bullet & \bullet & {[0]} & \bullet \\
\bullet & \bullet & \bullet & {[0]}
\end{array}\right)
$$

In this situation we remove: 30 moves involving 4 cells: 80 moves involving 6 cells: and 66 moves involving 8 cells. Finally, the Universal Markov basis has only 28 moves.

However, the Universal Gröbner basis of the toric ideal $\mathcal{I}_{A_{T}}$ is, in general, much bigger than a Gröbner basis of the toric ideal $\mathcal{I}_{A_{T}}$ with respect to a given term order. So in general it is very hard to compute.

Just to give the idea of such increase, we present in the following table the number of moves of the standard Markov basis for square $I \times I$ tables for the first $I$ 's.

|  | 2 | 3 | 4 | 5 | 6 | 7 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Standard Markov basis | 1 | 9 | 36 | 100 | 225 | 441 |
| Universal Gröbner basis | 1 | 15 | 204 | 3,940 | 113,865 | $4,027,161$ |

Thus, we consider the set of connecting moves.

## Markov subbases

Definition: [Chen et. al., 2007] A Markov subbasis $M_{A_{T}, \mathbf{n}_{\text {obs }}}$ for $\mathbf{n}_{\text {obs }} \in \mathbb{N}^{k}$ and integer matrix $A_{T}$ is a finite subset of $\operatorname{ker}\left(A_{T}\right) \cap \mathbb{Z}^{k}$ such that, for each pair of vectors $\mathbf{u}, \mathbf{v} \in \mathcal{F}_{T}$, there is a sequence of vectors $\mathbf{m}_{i} \in M_{A_{T}, \mathbf{n}_{\mathrm{obs}}}, i=$ $1, \ldots, l$, such that

$$
\begin{gathered}
\mathbf{u}=\mathbf{v}+\sum_{i=1}^{l} \mathbf{m}_{i}, \\
0 \leq \mathbf{v}+\sum_{i=1}^{j} \mathbf{m}_{i}, j=1, \ldots, l .
\end{gathered}
$$

The connectivity through nonnegative lattice points only is required to hold for this specific $\mathbf{n}_{\text {obs }}$.

Note: $M_{A_{T}, \mathbf{n}_{\text {obs }}}$ for every $\mathbf{n}_{\text {obs }} \in \mathbb{N}^{k}$ and for a given $A_{T}$ is a Markov basis $\mathcal{M}$ for $A_{T}$.

## Markov subbases for tables with positive bounds

We consider Markov subbases $M_{A_{T}, \mathbf{n}_{\mathrm{obs}}}$ for any bounded two-way contingency tables $\mathbf{n}_{\text {obs }} \in \mathbb{N}^{k}$ with positive bounds, i.e., no structural zeros, under independence model.

Theorem [Rapallo and Y., 2009] Consider $I \times J$ tables with row and column sums fixed and with all cells bounded. If these bounds are positive, then a Markov subbasis for the tables is the standard Markov basis for $I \times J$ tables with row and column sums fixed without bounds, i.e., the set of basic moves of all $2 \times 2$ minors.

## Example

Consider now $4 \times 4$ tables with fixed row and column sums, and with all bounded cells.

The constraint matrix that fixes row and column sums in a $4 \times 4$ table gives a toric ideal with a $\binom{4}{2} \times\binom{ 4}{2}$ element Gröbner basis, i.e., a Markov basis is formed by the basic moves of the form $\left(\begin{array}{ll}+1 & -1 \\ -1 & +1\end{array}\right)$ for all $2 \times 2$ minors of the table.

The full Markov basis for bounded tables has 204 moves. However, by the theorem above the Markov subbasis for this table is the standard Markov basis for a $4 \times 4$ table with fixed row and column sums fixed without bounds.

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## Thank you....

The paper is available at http://arxiv.org/abs/0905.4841.

