

Algebra of Reversible Markov Chains

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Abstract and a few references

A Markov Chain with stationary un-normalized probability π and transitions P_{xy} , $x, y \in V$, is reversible if the detailed balance conditions

$$\pi(x)P_{xy} = \pi(y)P_{yx}, \quad x \neq y,$$

are satisfied.

In turn, the detailed balance conditions generate a toric ideal of the ring $\mathbb{Q}[\pi(x), P_{xy}]$, so that reversible Markov Chains are part of a toric variety.

Elimination of the indeterminate π produces again a toric ideal, whose binomial generators are the Kolmogorov's equations on the closed paths of the undirected graph of the transition matrix P .

We discuss the parameterization of this model and the effect of imposing a toric model on the π 's.

- R.L. Dobrushin, Y.M. Sukhov, Ī. Fritts, *Uspekhi Mat. Nauk* **43**(6(264)), 167 (1988), ISSN 0042-1316, <http://dx.doi.org/10.1070/RM1988v043n06ABEH001985>
- Chapter 5 in D.W. Strook, *An Introduction to Markov Processes*, Number 230 in Graduate Texts in Mathematics (Springer-Verlag, Berlin, 2005)
- J.L. Lebowitz, H. Spohn, *J. Statist. Phys.* **95**(1-2), 333 (1999), ISSN 0022-4715, <http://dx.doi.org/10.1023/A:1004589714161>

2-reversible processes

The stochastic process $(X_n)_{n \geq 0}$ with state space V is *2-reversible* if

$$P(X_n = x, X_{n+1} = y) = P(X_n = y, X_{n+1} = x), \quad x, y \in V, n \geq 0$$

In particular, the process is 1-stationary:

$$P(X_n = x) = P(X_{n+1} = x) = \pi(x), \quad x \in V, n \geq 0.$$

Let $V_1 = \binom{N}{1}$, $V_2 = \binom{V}{2}$. In

- P. Diaconis, S.W.W. Rolles, Ann. Statist. **34**(3), 1270 (2006), ISSN 0090-5364, <http://dx.doi.org/10.1214/009053606000000290>

the following parameterization is considered:

$$\begin{aligned}\theta_{\{x,y\}} &= 2P(X_n = x, X_{n+1} = y), \quad \{x,y\} \subset V_2; \\ \theta_{\{x\}} &= P(X_n = x, X_{n+1} = x), \quad x \in V_1.\end{aligned}$$

We have:

$$1 = \sum_{x,y \in V} P(X_n = x, X_{n+1} = y) = \sum_{\{x\} \in V_1} \theta_{\{x\}} + \sum_{\{x,y\} \in V_2} \theta_{\{x,y\}},$$

so that $\theta = (\theta_{\{x\}} : \{x\} \in V_1, \theta_{\{x,y\}} : \{x,y\} \in V_2)$ belongs to the simplex $\Delta(V_1 \cup V_2)$.

Restriction on a graph

We consider here a special case. We assume we are given the undirected connected graph $\mathcal{G} = (V, \mathcal{E})$ such that $\theta_{\{x,y\}} = 0$ if $\{x,y\} \notin \mathcal{E}$. The the vector of parameters $\theta = (\theta_{\{x\}}: \{x\} \in V, \theta_{\{x,y\}}: \{x,y\} \in \mathcal{E})$ belong to the simplex $\Delta(V \cup \mathcal{E})$.

The probability $\pi(x)$ can is a linear function of the θ parameters:

$$\pi(x) = \sum_{y \in V} P(X_n = x, X_{n+1} = y) = \theta_{\{x\}} + \frac{1}{2} \sum_{y: \{x,y\} \in \mathcal{E}} \theta_{\{x,y\}}$$

or, in matrix form,

$$\pi = \theta_V + \frac{1}{2} \Gamma \theta_{\mathcal{E}},$$

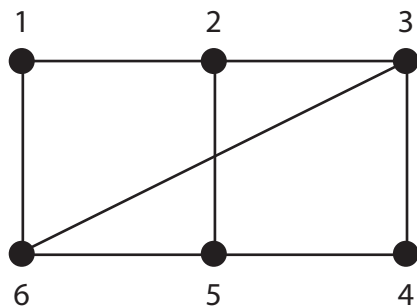
where Γ is the incidence matrix of the graph \mathcal{G} . The map

$$\gamma: \Delta(V \cup \mathcal{E}) \ni \theta = \begin{bmatrix} \theta_V \\ \theta_{\mathcal{E}} \end{bmatrix} \mapsto \pi = [I_V \quad \frac{1}{2} \Gamma] \begin{bmatrix} \theta_V \\ \theta_{\mathcal{E}} \end{bmatrix} \in \Delta(V)$$

is a surjective Markov map. In fact, the image of the convex set $\Delta(V \cup \mathcal{E})$ is the convex hull of the extreme points e_i , $i \in V$, and $e_{\{x,y\}}$, $\{x,y\} \in \mathcal{E}$, hence

$\gamma(\Delta(V \cup \mathcal{E}))$ is the convex hull of the columns of the matrix $[I_V \quad \frac{1}{2} \Gamma]$. The image of $(\theta_V, 0)$, $\theta_V \in \Delta(V)$, is full; the image of $(0, \theta_{\mathcal{E}})$, $\theta_{\mathcal{E}} \in \Delta(\mathcal{E})$, is the convex hull in $\Delta(V)$ of the half points of each edge of the graph \mathcal{G} .

Example 1A



$$\Gamma = \begin{matrix} & \{1,2\} & \{2,3\} & \{1,6\} & \{2,5\} & \{3,4\} & \{5,6\} & \{4,5\} & \{3,6\} \\ \begin{matrix} 1 \\ 2 \\ 3 \\ 4 \\ 5 \\ 6 \end{matrix} & \begin{bmatrix} 1 & 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 \\ 1 & 1 & 0 & 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 1 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 1 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 & 0 & 1 & 1 & 1 & 0 \\ 0 & 0 & 1 & 0 & 0 & 1 & 1 & 0 & 1 \end{bmatrix} \end{matrix}$$

Joint 2-distributions with a given stationary π

- Given π , the fiber $\gamma^{-1}(\pi)$ is contained in an affine space parallel to the subspace $\theta_{\{x\}} + (1/2) \sum_{y: \{x,y\} \in \mathcal{E}} \theta_{\{x,y\}} = 0$.
- Each fiber contains special solutions. One is the zero transition case $(\pi, 0_{\mathcal{E}})$.
- If the graph has full connections, $\mathcal{G} = (V, V_2)$, there is the independence solution $\theta_{\{x\}} = \pi(x)^2$, $\theta_{\{x,y\}} = 2\pi(x)\pi(y)$.
- If $\pi(x) > 0$, $x \in V$, a strictly positive solution is obtained as follows. Let $d(x) = \#\{y: \{x,y\} \in \mathcal{E}\}$ be the degree of the vertex x and define a transition probability by $A(x,y) = 1/2d(x)$ if $\{x,y\} \in \mathcal{E}$, $A(x,x) = 1/2$, and $A(x,y) = 0$ otherwise. A is the transition matrix of a random walk on the graph \mathcal{G} , stopped with probability $1/2$. Define a probability on $V \times V$ with $Q(x,y) = \pi(x)A(x,y)$. If $Q(x,y) = Q(y,x)$, we have a 2-reversible probability with marginal π . Otherwise, there is a general construction as follows.

Proposition

Let Q be a probability on $V \times V$, strictly positive on \mathcal{E} , and let $\pi(x) = \sum_y Q(x, y)$. If $f :]0, 1[\times]0, 1[\rightarrow]0, 1[$ is a symmetric function such that $f(u, v) \leq u \wedge v$ then

$$P(x, y) = \begin{cases} f(Q(x, y), Q(y, x)) & \{x, y\} \in \mathcal{E} \\ \pi(x) - \sum_{y: y \neq x} P(x, y) & x = y \\ 0 & \text{otherwise,} \end{cases}$$

is a 2-reversible probability on \mathcal{E} such that $\pi(x) = \sum_y P(x, y)$, positive if Q is positive.

The proposition applies to

- $f(u, v) = u \wedge v$. This is the standard Metropolis case: $u \wedge v = u(1 \wedge \frac{v}{u})$
- $f(u, v) = uv$. In fact, as $v \leq 1$, we have $uv \leq u$. For our purposes, this case is interesting, because it is an algebraic function.

Proof.

For $\{x, y\} \in \mathcal{E}$ we have $P(x, y) = P(y, x) > 0$. As $P(x, y) \leq Q(x, y)$, $x \neq y$, it follows

$$\begin{aligned} P(x, x) &= \pi(x) - \sum_{y: y \neq x} P(x, y) \\ &\geq \sum_y Q(x, y) - \sum_{y: y \neq x} Q(x, y) \\ &= Q(x, x) > 0. \end{aligned}$$

We have $\sum_y P(x, y) = \pi(x)$ by construction and, in particular, P is a probability on $V \times V$. □

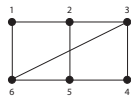
- Given a positive Q , the corresponding parameters for P

$$\theta_{\{x, y\}} = 2P(x, y), \quad \theta_{\{x\}} = P(x, x)$$

are strictly positive. We have shown the existence of a mapping from the interior of $\Delta(V)$ to the interior of $\Delta(V_1 \cup \mathcal{E})$.

- The mapping $\theta \mapsto (\pi, P_{xy} = \frac{P(x, y)}{\pi(x)})$ is a rational mapping from $\Delta(V_1 \cup V_2)$ into $\Delta(V) \otimes \Delta(V)^{\otimes V}$.

Example 1B



$$Q = \begin{matrix} & \begin{matrix} 1 & 2 & 3 & 4 & 5 & 6 \end{matrix} \\ \begin{matrix} 1 \\ 2 \\ 3 \\ 4 \\ 5 \\ 6 \end{matrix} & \begin{bmatrix} 1 & 2 & 3 & 4 & 5 & 6 \\ 0 & \frac{1}{2}\pi(1) & 0 & 0 & 0 & \frac{1}{2}\pi(1) \\ \frac{1}{3}\pi(2) & 0 & \frac{1}{3}\pi(2) & 0 & \frac{1}{3}\pi(2) & 0 \\ 0 & \frac{1}{3}\pi(3) & 0 & \frac{1}{3}\pi(3) & 0 & \frac{1}{3}\pi(3) \\ 0 & 0 & \frac{1}{2}\pi(4) & 0 & \frac{1}{2}\pi(4) & 0 \\ 0 & \frac{1}{3}\pi(5) & 0 & \frac{1}{3}\pi(5) & 0 & \frac{1}{3}\pi(5) \\ \frac{1}{3}\pi(6) & 0 & \frac{1}{3}\pi(6) & 0 & \frac{1}{3}\pi(6) & 0 \end{bmatrix} \end{matrix}$$

$$P = \begin{matrix} & \begin{matrix} 1 & 2 & 3 & 4 & 5 & 6 \end{matrix} \\ \begin{matrix} 1 \\ 2 \\ 3 \\ 4 \\ 5 \\ 6 \end{matrix} & \begin{bmatrix} P(11) & \frac{1}{6}\pi(1)\pi(2) & 0 & 0 & 0 & \frac{1}{6}\pi(1)\pi(6) \\ \frac{1}{6}\pi(1)\pi(2) & P(22) & \frac{1}{9}\pi(2)\pi(3) & 0 & \frac{1}{9}\pi(2)\pi(5) & 0 \\ 0 & \frac{1}{9}\pi(2)\pi(3) & P(33) & \frac{1}{6}\pi(3)\pi(4) & 0 & \frac{1}{9}\pi(3)\pi(6) \\ 0 & 0 & \frac{1}{6}\pi(3)\pi(4) & P(44) & \frac{1}{6}\pi(4)\pi(5) & 0 \\ 0 & \frac{1}{9}\pi(2)\pi(5) & 0 & \frac{1}{6}\pi(4)\pi(5) & P(55) & \frac{1}{9}\pi(5)\pi(6) \\ \frac{1}{6}\pi(1)\pi(6) & 0 & \frac{1}{9}\pi(3)\pi(6) & 0 & \frac{1}{9}\pi(5)\pi(6) & P(66) \end{bmatrix} \end{matrix}$$

Example 1C

edges

$$9\theta_{\mathcal{E}} = \begin{array}{l} \{1, 2\} \\ \{2, 3\} \\ \{1, 6\} \\ \{2, 5\} \\ \{3, 4\} \\ \{5, 6\} \\ \{4, 5\} \\ \{3, 6\} \end{array} \begin{bmatrix} 3\pi(1)\pi(2) \\ 2\pi(2)\pi(3) \\ 3\pi(1)\pi(6) \\ 2\pi(2)\pi(5) \\ 3\pi(3)\pi(4) \\ 2\pi(5)\pi(6) \\ 3\pi(4)\pi(5) \\ 2\pi(3)\pi(6) \end{bmatrix} \quad \text{and} \quad \theta_V = \pi - \frac{1}{2}\Gamma\theta_{\mathcal{E}}$$

$$\log \bar{\theta}_{\mathcal{E}} = \text{const} + \Gamma^t \pi$$

$$\bar{\theta}_V = \pi - \frac{1}{2}\Gamma\bar{\theta}_{\mathcal{E}} \iff \theta = \bar{\theta} + \delta \in \gamma^{-1}(\pi)$$

$$\delta_V + \frac{1}{2}\Gamma\delta_{\mathcal{E}} = 0$$

Example 1D

$$\pi = \text{Binomial}(5, p) + 1 \implies$$

edges

$$9\theta_{\mathcal{E}} = \begin{matrix} \{1, 2\} \\ \{2, 3\} \\ \{1, 6\} \\ \{2, 5\} \\ \{3, 4\} \\ \{5, 6\} \\ \{4, 5\} \\ \{3, 6\} \end{matrix} \begin{bmatrix} 3 \binom{5}{0} p^0 (1-p)^5 \binom{5}{1} p^1 (1-p)^4 \\ 2 \binom{5}{1} p^1 (1-p)^4 \binom{5}{2} p^2 (1-p)^3 \\ 3 \binom{5}{0} p^0 (1-p)^5 \binom{5}{5} p^5 (1-p)^0 \\ 2 \binom{5}{1} p^1 (1-p)^4 \binom{5}{4} p^4 (1-p)^1 \\ 3 \binom{5}{2} p^2 (1-p)^3 \binom{5}{3} p^3 (1-p)^2 \\ 2 \binom{5}{4} p^4 (1-p)^1 \binom{5}{5} p^5 (1-p)^0 \\ 3 \binom{5}{3} p^3 (1-p)^2 \binom{5}{4} p^4 (1-p)^1 \\ 2 \binom{5}{2} p^2 (1-p)^5 \binom{5}{5} p^5 (1-p)^0 \end{bmatrix} = \begin{bmatrix} 3 \binom{5}{0} \binom{5}{1} p^1 (1-p)^9 \\ 2 \binom{5}{1} \binom{5}{2} p^3 (1-p)^4 \\ 3 \binom{5}{0} \binom{5}{5} p^5 (1-p)^5 \\ 2 \binom{5}{1} \binom{5}{4} p^5 (1-p)^5 \\ 3 \binom{5}{2} \binom{5}{3} p^5 (1-p)^5 \\ 2 \binom{5}{4} \binom{5}{5} p^9 (1-p)^1 \\ 3 \binom{5}{3} \binom{5}{4} p^7 (1-p)^3 \\ 2 \binom{5}{2} \binom{5}{5} p^7 (1-p)^3 \end{bmatrix}$$

Reversible Markov chain

Assume the 2-reversible process $(X_n)_{n \in \mathbb{N}}$ is a Markov chain and consider the undirected graph $\mathcal{G} = (V, \mathcal{E})$ such that $\{x, y\} \in \mathcal{E}$ if, and only if, $\theta_{\{x,y\}} > 0$. The transition probability are:

$$p_{xy} = \frac{P(X_n = x, X_{n+1} = y)}{P(X_n = x)} = \frac{\theta_{\{x,y\}}}{\sum_{y: \{x,y\} \in \mathcal{E}} \theta_{\{x,y\}}}$$
$$p_{yx} = \frac{P(X_n = y, X_{n+1} = x)}{P(X_n = y)} = \frac{\theta_{\{x,y\}}}{\sum_{y: \{x,y\} \in \mathcal{E}} \theta_{\{x,y\}}}$$

so that, denoting $\sum_y \theta_{\{x,y\}}$ by $k(x)$, we have the detailed balance conditions

$$k(x)P_{xy} = k(y)P_{yx}, \quad x \neq y.$$

Vice-versa, if there exist positive constants $k(x)$, $x \in V$ such that the detailed balance conditions hold, then the process is 2-reversible with $\pi \propto k$.

Reversibility

Let $\omega = v_0 \cdots v_n$ be a path in the connected graph $\mathcal{G} = (V, \mathcal{E})$ and let $-\omega = v_n \cdots v_0$ be the reverse path. Then

$$\pi(v_0)P_{v_0v_1} = \pi(v_1)P_{v_1v_0}$$

$$\pi(v_1)P_{v_1v_2} = \pi(v_2)P_{v_2v_1}$$

$$\vdots$$

$$\pi(v_{n-1})P_{v_{n-1}v_n} = \pi(v_n)P_{v_nv_{n-1}}$$

Proposition

If the detailed balance holds with $\prod_V \pi(v) \neq 0$, the reversibility condition

$$P(\omega) = P(-\omega)$$

holds for all path ω .

Kolmogorov's condition

- We denote now by ω a circuit, that is a path on the graph such that the last vertex coincides with the first one, $\omega = v_0 v_1 \dots v_n v_1$, and by $-\omega$ the reverse circuit $-\omega = v_0 v_n \dots v_1 v_0$.
- A 'basis' of circuits is obtained by considering a spanning tree and adding the remaining vertices one by one.

Theorem

Let then the Markov chain $(X_n)_{n \in \mathbb{N}}$ have support on the connected graph \mathcal{G} . If the process is reversible, for all circuit ω

$$P(\omega | X_0 = v_0) = P(-\omega | X_0 = v_0).$$

If the equality is true on a basis of circuits, then the process is reversible.

- A 'basis' of circuits is a basis of the polynomial (binomial) ideal.

Proof.

- If $P(\omega) = P(-\omega)$, then for a circuit $\omega = v v_1 \cdots v_{n-1} v$, we have $P(\omega | X_0 = v) = P(-\omega | X_n = v)$.
- Vice-versa, assume that all the circuit have the displayed property. We denote by x and y the first and the next to last vertices, respectively. By summing on the intermediate vertices on all circuits with same x and y , we obtain:

$$\sum_{v_2 v_3 \dots v_{n-1}} P_{xv_2} P_{v_2 v_3} \cdots P_{yx} = \sum_{v_2 v_3 \dots v_{n-1}} P_{xy} \cdots P_{v_3 v_2} P_{v_2 x}$$

and

$$P_{xy}^{(n-2)} P_{yx} = P_{xy} P_{xy}^{(n-2)}$$

where $P_{xy}^{(n-2)}$ denotes the $(n-2)$ -step transition probability. If $n \rightarrow \infty$, then $\pi(j) P_{yx} = P_{xy} \pi(x)$, and the chain is reversible.



CoCoA: Square

```
Use S:=Q[k[1..4],p[1..4,1..4]];S;
Set Indentation;
NI:=4;M=[];
Define Lista(L,NI);
    For I:=1 To NI Do
        For J:=1 To I-1 Do
            Append(L,k[I]p[I,J]-k[J]p[J,I]);
        EndFor;
    EndFor;
Return L;
EndDefine;
N:=Lista(M,NI);
LL:=Product([k[I]|I In 1..NI])-1;-- non-zero k's
J:=Ideal(p[1,3]^1,p[3,1]^1,p[2,4]^1,p[4,2])^1;
Append(N,LL);N;
I:=Ideal(N);
K:=I+J;
E:=Elim(k,K);E;
```


CoCoA: Square out

```
Q[k[1..4],p[1..4,1..4]]
```

```
-----  
[ -k[1]p[1,2] + k[2]p[2,1],  
  -k[1]p[1,3] + k[3]p[3,1],  
  -k[2]p[2,3] + k[3]p[3,2],  
  -k[1]p[1,4] + k[4]p[4,1],  
  -k[2]p[2,4] + k[4]p[4,2],  
  -k[3]p[3,4] + k[4]p[4,3],  
  k[1]k[2]k[3]k[4] - 1]
```

```
-----  
Ideal(  
  p[4,2],  
  p[2,4],  
  p[3,1],  
  p[1,3],  
  p[1,2]p[2,3]p[3,4]p[4,1] - p[1,4]p[2,1]p[3,2]p[4,3])
```

CoCoA: Square with diag 13

```
Use S:=Q[k[1..4],p[1..4,1..4]];
Set Indentation;
NI:=4;M=[];
Define Lista(L,NI);
    For I:=1 To NI Do
        For J:=1 To I-1 Do
            Append(L,k[I]p[I,J]-k[J]p[J,I]);
        EndFor;
    EndFor;
    Return L;
EndDefine;
N:=Lista(M,NI);
LL:=Product([k[I]|I In 1..NI])-1;-- non zero k's
J:=Ideal(p[2,4]^1,p[4,2]^1); -- diagonal 13
Append(N,LL);
I:=Ideal(N);
K:=I+J;K;
E:=Elim(k,K);E;
```

CoCoA: Square with diag 13 out

```
Q[k[1..4],p[1..4,1..4]]
```

```
-----  
Ideal(  
  -k[1]p[1,2] + k[2]p[2,1],  
  -k[1]p[1,3] + k[3]p[3,1],  
  -k[2]p[2,3] + k[3]p[3,2],  
  -k[1]p[1,4] + k[4]p[4,1],  
  -k[2]p[2,4] + k[4]p[4,2],  
  -k[3]p[3,4] + k[4]p[4,3],  
  k[1]k[2]k[3]k[4] - 1,  
  p[2,4],  
  p[4,2])
```

```
-----  
Ideal(  
  p[4,2],  
  p[2,4],  
  p[1,3]p[3,4]p[4,1] - p[1,4]p[3,1]p[4,3],  
  p[1,2]p[2,3]p[3,1] - p[1,3]p[2,1]p[3,2],  
  p[1,2]p[2,3]p[3,4]p[4,1] - p[1,4]p[2,1]p[3,2]p[4,3])
```