

Iterated Proportional Fitting Procedure and infinite products of stochastic matrices

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Journées de probabilités

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Initial problem.

Let $p \geq 2$ and $q \geq 2$ be two integers. We are given:

- two target marginals $a \in (\mathbf{R}_+^*)^p$, $b \in (\mathbf{R}_+^*)^q$ such that $a_1 + \cdots + a_p = b_1 + \cdots + b_q = 1$,
- a matrix $X_0 \in \mathcal{M}_{p,q}(\mathbf{R}_+)$. One may assume that the entries add up to 1.

Call $\Gamma(a, b, X_0)$ the set of all matrices $X \in \mathcal{M}_{p,q}(\mathbf{R}_+)$ with support contained in $\text{Supp}(X_0)$ such that

$$\forall i \in [1, p], \quad X(i, +) := \sum_{j=1}^q X(i, j) \text{ equals } a_i,$$

$$\forall j \in [1, q], \quad X(+, j) := \sum_{i=1}^p X(i, j) \text{ equals } b_j.$$

We look for a matrix $X \in \Gamma(a, b, X_0)$ as close to X_0 as possible:

- we want X to be diagonally equivalent to X_0 , namely $X = D_1 X_0 D_2$ for some diagonal matrices $D_1 \in \mathcal{M}_p(\mathbf{R}_+^*)$ and $D_2 \in \mathcal{M}_q(\mathbf{R}_+^*)$ (biproportional fitting);
- equivalently, we want to minimize the relative entropy

$$D(X||X_0) := \sum_{(i,j) \in \text{Supp}(X_0)} X(i,j) \ln \frac{X(i,j)}{X_0(i,j)}.$$

Concrete application: estimation of origin/destination matrix on a public bus line

$X(i, j)$ = proportion of population going from station i to station j .

$X_0(i, j)$ = rough estimation.

a_i = proportion of users getting in at station i (well known)

b_j = proportion of users getting out at station j (well known)

Probabilistic interpretation We view a , b and X_0 as probability measures on $[1, p]$, $[1, q]$ and $[1, p] \times [1, q]$. So we look at couplings of a and b with support contained in $\text{Supp}(X_0)$.

- Existence of such couplings?
- Existence and uniqueness of a coupling X minimizing $D(X||X_0)$?
- How to approach it?

Note that set $\Gamma(a, b, X_0)$ is a compact (convex) polyedron in $\mathcal{M}_{p,q}(\mathbf{R})$, possibly empty. Fourier or Farkas' lemma applies.

For $A \subset [1, p]$ and $B \subset [1, q]$, we set

$$a(A) := \sum_{i \in A} a_i, \quad b(B) := \sum_{j \in B} b_j.$$

Cause of incompatibility: two (non-empty) subsets $A \subset [1, p]$ and $B \subset [1, q]$ such that X_0 is null on $A \times B$ and $a(A) + b(B) > 1$.

$$X_0 = \left(\begin{array}{c|c} & 0 \\ \hline - & - \\ \hline & \end{array} \right) \begin{array}{l} \} A \\ \\ \} A^c \end{array}$$

$\underbrace{\hspace{2em}}_{B^c} \quad \underbrace{\hspace{2em}}_B$

Cause of criticality: two non-empty subsets $A \subset [1, p]$ and $B \subset [1, q]$ such that X_0 is null on $A \times B$ and $a(A) + b(B) = 1$. This forces matrices in $\Gamma(a, b, X_0)$ to be null on $A^c \times B^c$.

Theorem (Bacharach, Pukelsheim)

Three situations may occur.

- 1 (Nice case) *If there is no cause of incompatibility neither criticality, then $\Gamma(a, b, X_0)$ contains some matrix with same support as X_0 .*
- 2 (Critical case) *If there is no cause of incompatibility but some cause of criticality, then $\Gamma(a, b, X_0)$ is not empty and contains only matrices with support strictly included in $\text{Supp}(X_0)$.*
- 3 (Case of incompatibility) *If there is some cause of incompatibility, then, $\Gamma(a, b, X_0)$ is empty.*

The IPFP, introduced in 1937 by Kruithof, works as follows.
 Call $\Gamma(*, *, *)$ the set of all non-negative $p \times q$ matrices having positive sum on each row or column.

For $X \in \Gamma(*, *, *)$, $i \in [1, p]$ and $j \in [1, q]$, set

$$R_i(X) := a_i^{-1} X(i, +), \quad C_j(X) := b_j^{-1} X(+, j).$$

Define $T_R : \Gamma(*, *, *) \rightarrow \Gamma(a, *, *)$, $T_C : \Gamma(*, *, *) \rightarrow \Gamma(*, b, *)$ by

$$T_R(X)(i, j) := R_i(X)^{-1} X(i, j), \quad T_C(X) := C_j(X)^{-1} X(i, j).$$

Set $X_{2n+1} = T_R(X_{2n})$, $X_{2n+2} = T_C(X_{2n+1})$ for every $n \geq 0$.

Does the sequence $(X_n)_{n \geq 0}$ thus defined converge?

If yes, the limit X_∞ belongs to $\Gamma(a, b, X_0)$.

Theorem (Bacharach, Bregman, Sinkhorn, Csiszar, Pretzel...)

(Nice case: fast convergence) Assume that $\Gamma(a, b, X_0)$ contains some matrix with same support as X_0 . Then

- 1 *The sequences $(R_i(X_{2n})_{n \geq 0})$ and $(C_j(X_{2n+1})_{n \geq 0})$ converge to 1 at an at least geometric rate.*
- 2 *The sequence $(X_n)_{n \geq 0}$ converges to some matrix X_∞ which has the same support as X_0 . The rate of convergence is at least geometric.*
- 3 *The limit X_∞ is the only matrix in $\Gamma(a, b, X_0)$ which is diagonally equivalent to X_0 , or equivalently, which minimizes of $D(Y||X_0)$ over all $Y \in \Gamma(a, b, X_0)$.*

Theorem (Bacharach, Bregman, Sinkhorn, Csiszar, Pretzel...)

Assume that $\Gamma(a, b, X_0)$ is not empty. Then

- 1 The sequences $(R_i(X_{2n}))_{n \geq 0}$ and $((C_j(X_{2n+1}))_{n \geq 0}$ are $1 + o(n^{-1/2})$.
- 2 The sequence $(X_n)_{n \geq 0}$ converges to some matrix X_∞ whose support contains the support of every matrix in $\Gamma(a, b, X_0)$.
- 3 The limit X_∞ is the unique matrix achieving the minimum of $D(Y||X_0)$ over all $Y \in \Gamma(a, b, X_0)$.

Remark: if $\Gamma(a, b, X_0)$ contains only matrices with support strictly included in $\text{Supp}(X_0)$, the additional zeroes in X_∞ are given by the causes of criticality. One gets the same limit with a faster convergence by replacing the corresponding entries in X_0 with 0.

Theorem (Gietl and Reffel, Aas, L.)

There exist $r \in [1, \min(p, q)]$ ($r \geq 2$ iff $\Gamma(a, b, X_0)$ is empty) and partitions $\{I_1, \dots, I_r\}$ of $[1, p]$ and $\{J_1, \dots, J_r\}$ of $[1, q]$ such that:

- 1 The ratios $\lambda_k := b(J_k)/a(I_k)$ increase with $k \in [1, r]$;
- 2 $(R_i(X_{2n}))_{n \geq 0}$ converges to λ_k whenever $i \in I_k$;
- 3 $(C_j(X_{2n+1}))_{n \geq 0}$ converges to λ_k^{-1} whenever $j \in J_k$;
- 4 When $k < k'$, $X_n(i, j) = 0$ for every $n \geq 0$, $i \in I_k$ and $j \in J_{k'}$;
- 5 When $k < k'$, $X_n(i, j) \rightarrow 0$ for every $n \geq 0$, $i \in I_k$ and $j \in J_{k'}$;
- 6 $(X_{2n})_{n \geq 0}$ converges to $\arg \min_{Y \in \Gamma(a', b, X_0)} D(Y \| X_0)$, where $a'_i = \lambda_k a_i$ whenever $i \in I_k$;
- 7 $(X_{2n+1})_{n \geq 0}$ converges to $\arg \min_{Y \in \Gamma(a, b', X_0)} D(Y \| X_0)$, where $b'_j = \lambda_k^{-1} b_j$ whenever $j \in J_k$;

The partitions $\{I_1, \dots, I_r\}$ and $\{J_1, \dots, J_r\}$ are algorithmically determined by the causes of incompatibility and depend only on a , b and $\text{Supp}(X_0)$.

Replacing a by a' or b by b' transforms causes of incompatibility into causes of criticality.

Let $X \in \Gamma(*, b, *)$. Then for every $j \in [1, q]$,

$$C_j(T_R(X)) = \frac{1}{b_j} \sum_{i=1}^p R_i(X)^{-1} X(i, j) = \sum_{i=1}^p \frac{X(i, j)}{b_j} R_i(X)^{-1}.$$

In the same way, each $R_i(T_C(T_R(X)))$ is a weighted mean of the quantities $C_j(T_R(X))^{-1}$, therefore a weighted hybrid mean of the quantities $R_k(X)$... In particular,

$$\begin{aligned} [\underline{R}(X), \overline{R}(X)] &\supset [\underline{C}(T_R(X))^{-1}, \overline{C}(T_R(X))^{-1}] \\ &\supset [\underline{R}((T_C(T_R(X))), \overline{R}((T_C(T_R(X))))]. \end{aligned}$$

This can be applied to X_2, X_4, X_6, \dots

Lemma

Let $X \in \Gamma(*, b, *)$. Call $R(X)$ the column vector with entries $R(X_1), \dots, R(X_p)$. Then $R(T_C(T_R(X))) = P(X)R(X)$, where $P(X)$ is the $p \times p$ stochastic matrix given by

$$\forall i, k \in [1, p], \quad P(X)(i, k) = \sum_{j=1}^q \frac{T_R(X)(i, j) T_R(X)(k, j)}{a_i b_j C_j(T_R(X))}.$$

Moreover,

$$\forall i \in [1, p], \quad P(X)(i, i) \geq \frac{\underline{a}}{\bar{b} \bar{C}(T_R(X))q}$$

and

$$\forall i, k \in [1, p], \quad P(X)(k, i) \leq \frac{\bar{a}}{\underline{a}} P(X)(i, k).$$

Thus, the convergence of $R(X_{2n}) = P(X_{2n-2}) \cdots P(X_2)R(X_2)$ follows from the next theorem, which improves on Lorenz' stabilization theorem.

Theorem (L.)

Let $(M_n)_{n \geq 1}$ be some sequence of $d \times d$ stochastic matrices. Assume that there exists some constants $\gamma > 0$, and $\rho \geq 1$ such that for every $n \geq 1$ and i, j in $[1, d]$, $M_n(i, i) \geq \gamma$ and $M_n(i, j) \leq \rho M_n(j, i)$.

Then the sequence $(M_n \cdots M_1)_{n \geq 1}$ has a finite variation, so it converges to some stochastic matrix L . Moreover, the series $\sum_n M_n(i, j)$ and $\sum_n M_n(j, i)$ converge whenever the rows of L with indexes i and j are different.