

Parameterization of mixture independence models

E. Carlini

Dipartimento di Matematica
Politecnico di Torino

AMS meeting, University of Kentucky, March 2010



E.Carlini, F.Rapallo

Probability matrices, non-negative rank and parameterization of mixture models

to appear in LAA (2010)



Independence mixture models

$P \in \mathbb{R}^{I \times J}$ a $I \times J$ **probability matrix** (i.e. non-negative with sum of the entries one) of the form

$$P = \alpha_1 c_1 r_1^t + \dots + \alpha_k c_k r_k^t$$

where

$$c_i \in \mathbb{R}_{\geq 0}^I, \sum_j c_i(j) = 1, r_i \in \mathbb{R}_{\geq 0}^J, \sum_j r_i(j) = 1$$

and $\alpha_j \in \mathbb{R}_{\geq 0}, \sum \alpha_j = 1$.



$P = \sum_1^k \alpha_j c_j r_j^t$ is a probability matrix with $\text{rk}(P) \leq k$.

We are describing probability matrices of rank at most k using

$$k(I + J) - k - 1$$

parameters.

Questions

Can we use fewer parameters?

What is the least possible number of parameters?



Consider $\mathcal{M}_k \subset \mathbb{R}^{IJ}$ the variety of rank at most k matrices intersected with the simplex.

A **parameterization** of probability matrices with $\text{rk} \leq k$ using D parameters is an algebraic map

$$\mathbb{R}^D \supset U \longrightarrow \mathcal{M}_k$$

where U has non-empty interior.



To obtain almost all the elements of \mathcal{M}_k we need at least

$$D \geq \dim \mathcal{M}_k = k(I + J) - k^2 - 1$$

parameters.

Hence the parameterization

$$\mathbb{R}^{k(I+J)-k-1} \supset U \longrightarrow \mathcal{M}_k$$
$$(\alpha_1, \dots, \alpha_k, \mathbf{c}_1(1), \dots, \mathbf{r}_k(J)) \mapsto P = \sum \alpha_j \mathbf{c}_j \mathbf{r}_j^t$$

is **redundant**.



A matrix P such that

$$P = \sum_1^k \alpha_j c_j r_j^t$$

where $c_j \in \mathbb{R}_{\geq 0}^I$, $r_j \in \mathbb{R}_{\geq 0}^J$ and $\alpha_j \in \mathbb{R}_{\geq 0}$ has

non-negative rank at most k

and one writes $\text{rk}_+(P) \leq k$.



The non-negative rank and the rank can be different, i.e. there are matrices P such that

$$\text{rk}_+(P) \geq \text{rk}(P).$$

Consider, for example

$$P = \begin{pmatrix} 1 & 0 & 1 & 0 \\ 0 & 1 & 1 & 0 \\ 1 & 0 & 0 & 1 \\ 0 & 1 & 0 & 1 \end{pmatrix}$$

where $C_1 + C_2 = C_3 + C_4$ thus $\text{rk}(P) = 3$, **but** $\text{rk}_+(P) = 4$.



Hence the parameterization

$$P = \sum_1^k \alpha_j c_j r_j^t$$

is redundant and, in general, **non-surjective**.

Theorem (Cohen-Rothblum) If $\text{rk}(P) = 0, 1, 2$ then

$$\text{rk}_+(P) = \text{rk}(P).$$

Thus, for $k = 2$, $P = \sum_1^k \alpha_j c_j r_j^t$ is redundant but surjective.



Non-redundant parameterization for $\text{rk}=2$

We let $D = 2(I + J) - 5$ and we look for maps

$$\mathbb{R}^D \supset U \longrightarrow \mathcal{M}_2$$

with a dense image.

Idea:

- find $U_{j_1, j_2} \subset \mathbb{R}^D$ with non-empty interior;
- find maps $f_{j_1, j_2} : U_{j_1, j_2} \longrightarrow \mathcal{M}_2$ such that

$$\bigcup_{j_1, j_2} \text{Im}(U_{j_1, j_2}) = \mathcal{M}_2.$$



Non-redundant parameterization for $\text{rk}=2$

We consider the map

$$f_{1,2}(\mathbf{a}, \mathbf{b}, \mathbf{c}, \mathbf{d}, \alpha) = \alpha \mathbf{a}\mathbf{b}^t + (1 - \alpha)\mathbf{c}\mathbf{d}^t$$

where the vectors \mathbf{a}, \mathbf{c} have sum one and the vectors \mathbf{b}, \mathbf{d} have a fixed zero entry. The domain is

$$U_{1,2} = \left\{ (a_1, \dots, a_{I-1}, b_3, \dots, b_J, c_1, \dots, c_{I-1}, d_3, \dots, d_J, \alpha) \in \mathbb{R}^D : \right. \\ \left. 0 \leq a_i, b_i, c_i, d_i, \alpha \leq 1 \text{ and } 0 \leq \sum a_i, \sum b_i, \sum c_i, \sum d_i \leq 1 \right\}.$$



Non-redundant parameterization for $rk=2$

Where $f_{1,2} : U_{1,2} \rightarrow \mathcal{M}_2$ is defined as follows

$$\begin{aligned} f_{1,2}(a_1, \dots, a_{I-1}, b_3, \dots, b_J, c_1, \dots, c_{I-1}, d_3, \dots, d_J, \alpha) = \\ = \alpha \begin{pmatrix} a_1 \\ a_2 \\ \vdots \\ a_{I-1} \\ 1 - \sum a_i \end{pmatrix} (1 - \sum b_i \quad 0 \quad b_3 \quad \dots \quad b_J) + \\ + (1 - \alpha) \begin{pmatrix} c_1 \\ c_2 \\ \vdots \\ c_{I-1} \\ 1 - \sum c_i \end{pmatrix} (0 \quad 1 - \sum d_i \quad d_3 \quad \dots \quad d_J) \end{aligned}$$



Theorem

$$\text{Im}(U_{j_1, j_2}) = \mathcal{M}_2.$$

Proof Let P be in \mathcal{M}_2 . If $\text{rk}(P) = 1$ then we get P by setting $\alpha = 0$ or $\alpha = 1$. Otherwise $\text{rk}_+(P) = 2$ and we choose \mathbf{a} and \mathbf{c} accordingly.

Hence the maps

$$f_{j_1, j_2} : \mathbb{R}^{2(I+J)-5} \longrightarrow \mathcal{M}_2$$

give a **non-redundant** parameterization of the whole \mathcal{M}_2 and no fewer parameters can be used.



Given a map

$$F : \mathcal{M}_2 \longrightarrow \mathbb{R}$$

we want maximize/minimize F , i.e. we want to solve an optimization problem on rank two probability matrices.

E.g. F can be the likelihood function and we want to solve the **maximun likelihood** problem.



Theorem If $f_{j_1, j_2}(P') = P$ is a rank two matrix without two proportional columns, then

P is a maximum/minimum of F

iff

P' is a maximum/minimum of $F \circ f_{j_1, j_2}$.



If $f_{j_1, j_2}(P') = P \in U_{j_1, j_2}$ is a rank two matrix with at least two proportional columns then

P' is point of the boundary of U_{j_1, j_2}

but

$f_{j_1, j_2}(P')$ is **not** a point of the boundary of \mathcal{M}_2 .

Hence the study of $F \circ f_{j_1, j_2}$ is not enough.



Idea to maximize F on \mathcal{M}_2 :

- study F on rank one matrices
- maximize $F \circ f_{j_1, j_2}$ for all pairs (j_1, j_2) and make a list L of **candidates** extremal points
- if $P \in L$ has not two proportional columns, then $F(P)$ is a maximum
- if $P \in L$ has proportional columns it belongs to $\text{Im} f_{j_1, j_2}$ for different pairs (j_1, j_2)
- we study the extremal behaviour of all the functions

$$F \circ f_{j_1, j_2}$$

at all the point $P' \in U_{j_1, j_2}$ such that $f_{j_1, j_2}(P') = P$.

If these behaviors agree, then $F(P)$ is a maximum, otherwise it is not.

