



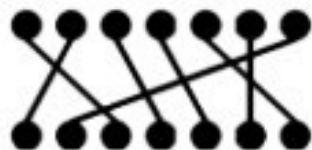
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# A branch and bound method to compute a median permutation

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# A permutation problem in voting theory

- Given a profile  $\Pi = (\sigma_1, \sigma_2, \dots, \sigma_m)$  of  $m$  permutations (i.e. linear orders)  $\sigma_i$  ( $1 \leq i \leq m$ ) on a set  $X$  of  $n = |X|$  elements, how to aggregate them into a unique permutation which summarizes  $\Pi$  as accurately as possible?
- In voting theory (Condorcet, 1784): we want to rank  $n$  candidates from the rankings provided by  $m$  voters.

# Example

- $X = \{a, b, c, d, e, f\}, m = 5$

voter 1:  $\sigma_1 = a > b > c > f > d > e$

voter 2:  $\sigma_2 = a > c > f > b > d > e$

voter 3:  $\sigma_3 = e > d > a > f > b > c$

voter 4:  $\sigma_4 = b > c > d > e > f > a$

voter 5:  $\sigma_5 = c > f > b > e > a > d.$

# A combinatorial optimization problem

- Symmetric difference distance  $d$  between  $R$  and  $R'$ :  
$$d(R, R') = |\{(x, y) \in X^2 \text{ with } [xRy \text{ and not } xR'y] \\ \text{or } [\text{not } xRy \text{ and } xR'y]\}|.$$
- Let  $\Sigma$  be the set of all the permutations defined on  $X$ .  
Then, for  $\Pi = (\sigma_1, \sigma_2, \dots, \sigma_m)$ :  
$$\text{Minimize } \rho_{\Pi}(\sigma) = \sum_{i=1}^m d(\sigma, \sigma_i) \text{ for } \sigma \in \Sigma$$

(cf. J.-P. Barthélemy, B. Monjardet, 1981)

- $d(R, R')$  measures the number of disagreements between  $R$  and  $R'$ .
- $\rho_{\Pi}(\sigma)$  (*= remoteness of  $\sigma$  from  $\Pi$* ) measures the total number of disagreements between  $\sigma$  and  $\Pi$ .
- $\sigma^*$  minimizing  $\rho_{\Pi}$  over  $\Sigma$  is called a *median permutation* (or a *median linear order*) of  $\Pi$ .
- **Theorem** (J.J. Bartholdi III *et alii*, 1989; O. Hudry, 1989; C. Dwork *et alii*, 2001):  
The computation of  $\sigma^*$  is NP-hard.

# A 0-1 linear programming problem

- $\sigma = (\sigma_{xy})_{(x,y) \in X^2}$  with  $\sigma_{xy} = 1$  if  $\sigma$  ranks  $x$  better than  $y$  ( $x >_{\sigma} y$ ) and  $\sigma_{xy} = 0$  otherwise.
- $m_{xy} = m - 2|\{i: 1 \leq i \leq m \text{ and } x >_{\sigma_i} y\}| = -m_{yx}$
- Then:  $\rho_{\Pi}(\sigma) = C + \sum_{(x,y) \in X^2} m_{xy} \sigma_{xy}$

with :

$$\forall x \in X, \sigma_{xx} = 1 \quad (\text{reflexivity})$$

$$\forall (x, y) \in X^2, x \neq y, \sigma_{xy} + \sigma_{yx} = 1 \quad (\text{antisymmetry})$$

$$\forall (x, y, z) \in X^3, \sigma_{xy} + \sigma_{yz} - \sigma_{xz} \leq 1 \quad (\text{transitivity})$$

$$\forall (x, y) \in X^2, \sigma_{xy} \in \{0, 1\} \quad (\text{binarity})$$

# Lagrangian relaxation

- Relaxation of the transitivity constraints:

$$\forall (x, y, z) \in X^3, \sigma_{xy} + \sigma_{yz} - \sigma_{xz} \leq 1$$

- Lagrangian function  $L$  for  $\sigma = (\sigma_{xy})_{(x,y) \in X^2}$  with  $\sigma_{xy} \in \{0, 1\}$ ,  $\sigma_{xx} = 1$ ,  $\sigma_{xy} + \sigma_{yx} = 1$ , and  $\Lambda = (\lambda_{xyz})_{(x,y,z) \in X^3}$  with  $\lambda_{xyz} \geq 0$ :

$$L(\sigma, \Lambda) = \rho_{\Pi}(\sigma) + \sum_{(x,y,z) \in X^3} \lambda_{xyz} (\sigma_{xy} + \sigma_{yz} - \sigma_{xz} - 1)$$

$$= C + \sum_{(x,y) \in X^2} a_{xy}(\Lambda) \sigma_{xy} - \sum_{(x,y,z) \in X^3} \lambda_{xyz}$$

with

$$a_{xy}(\Lambda) = m_{xy} + \sum_{z \in X} (\lambda_{xyz} + \lambda_{zxy} - \lambda_{xzy})$$

# Lagrangian relaxation (end)

- Dual function for  $\Lambda = (\lambda_{xyz})_{(x,y,z) \in X^3}$  with  $\lambda_{xyz} \geq 0$ :

$$D(\Lambda) = \min \{L(\sigma, \Lambda) \text{ with } \sigma \in \mathbf{A}\}$$

with  $\mathbf{A} = \{\text{reflexive and antisymmetric relations defined on } X\}$ .

- Dual problem: maximize  $D(\Lambda)$  for  $\Lambda \geq 0$ .
- The maximum of  $D$  gives a lower bound of the minimum of  $\rho_{\Pi}$ .
- Computation of  $D(\Lambda)$  for a given  $\Lambda$ :  
if  $a_{xy} \geq 0$ , set  $\sigma_{xy} = 0$ , and  $\sigma_{xy} = 1$  otherwise.
- Resolution of the dual problem by subgradient methods.



# The components of the BB algorithm

- **Initial bound**: found by a metaheuristic (a self-tuned noising method; I. Charon and O. Hudry, 1993, 2009)
- **Evaluation function**: provided by the Lagrangean relaxation.
- **Branching rule** (J.-P. Barthélemy, A. Guénoche, O. Hudry, 1989; I. Charon, A. Guénoche, O. Hudry, F. Woirgard, 1996):

The root of the BB-tree contains all the permutations defined on  $X$ .

A node of the BB-tree contains the permutations sharing a given *beginning section*  $S$  (i.e. a permutation of a subset of  $X$ ):

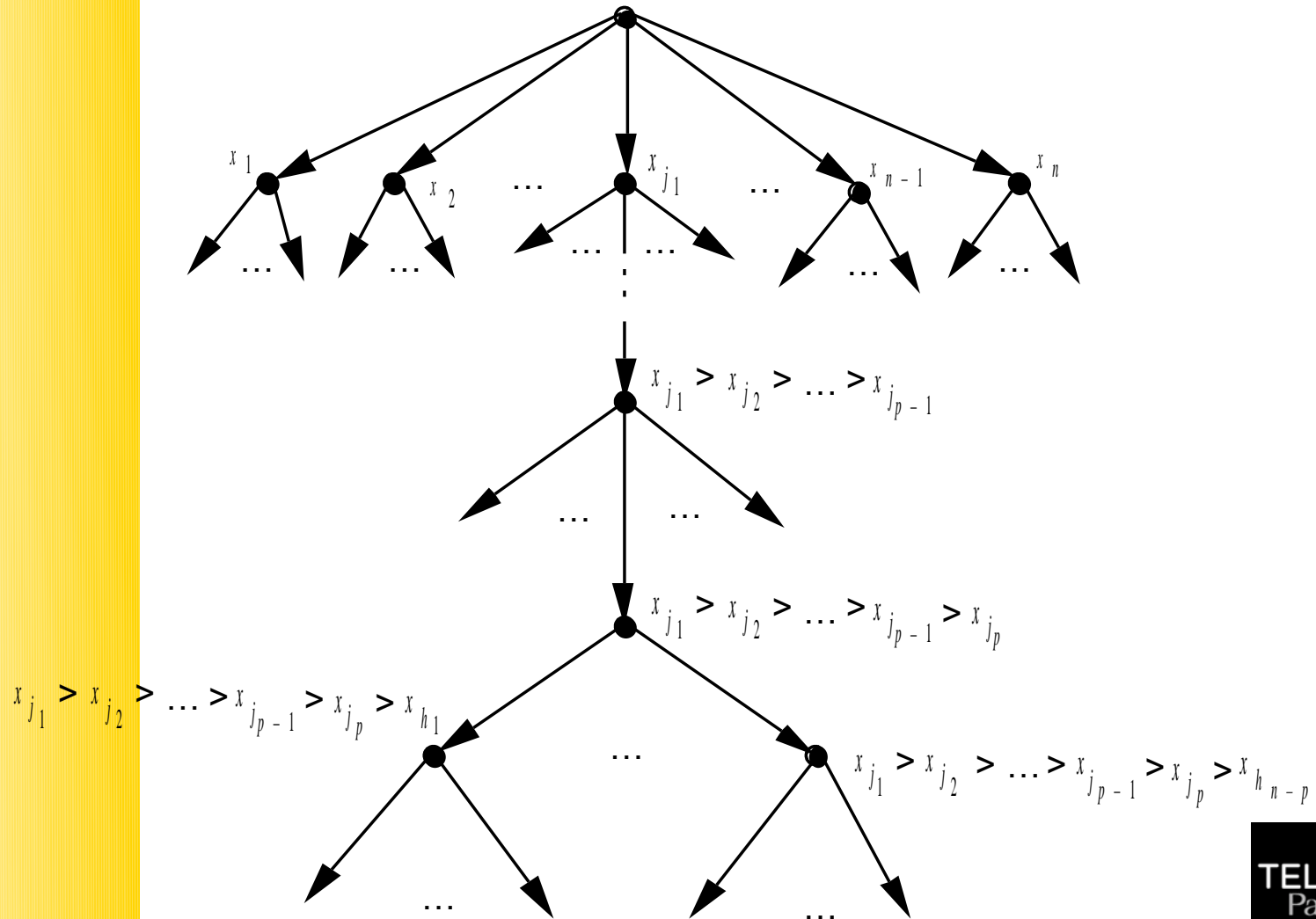
$$S(x_{j1}, x_{j2}, \dots, x_{jp}) = x_{j1} >_{\sigma} x_{j2} >_{\sigma} \dots >_{\sigma} x_{jp}.$$

The branching principle consists in expanding this beginning section:

$$S(x_{j1}, x_{j2}, \dots, x_{jp}, x) = x_{j1} >_{\sigma} x_{j2} >_{\sigma} \dots x_{jp} >_{\sigma} x$$

with  $x \notin \{x_{j1}, x_{j2}, \dots, x_{jp}\}$ .

# Shape of the BB-tree



# Other components to prune the BB-tree

- **Hamiltonian permutations.**
  - \* We may summarize a profile  $\Pi$  of permutations by a tournament  $T$  (weighted by  $-m_{xy} > 0$ ): there is an arc  $(x, y)$  if a majority of voters prefer  $x$  to  $y$  (we assume that there is no tie).
  - \* We say that a *permutation*  $\sigma$  is *Hamiltonian* if it induces a Hamiltonian path in  $T$ .
  - \* **Theorem** (R. Remage and W.A. Thompson, 1966): a median permutation is Hamiltonian.
  - $x_{j1} >_{\sigma} x_{j2} >_{\sigma} \dots >_{\sigma} x_{jp}$  is expanded into  $x_{j1} >_{\sigma} \dots >_{\sigma} x_{jp} >_{\sigma} x$  only if a majority of voters prefer  $x_{jp}$  to  $x$ .

# Example

- $X = \{a, b, c, d, e, f\}$

$$\sigma_1 = a > b > c > f > d > e$$

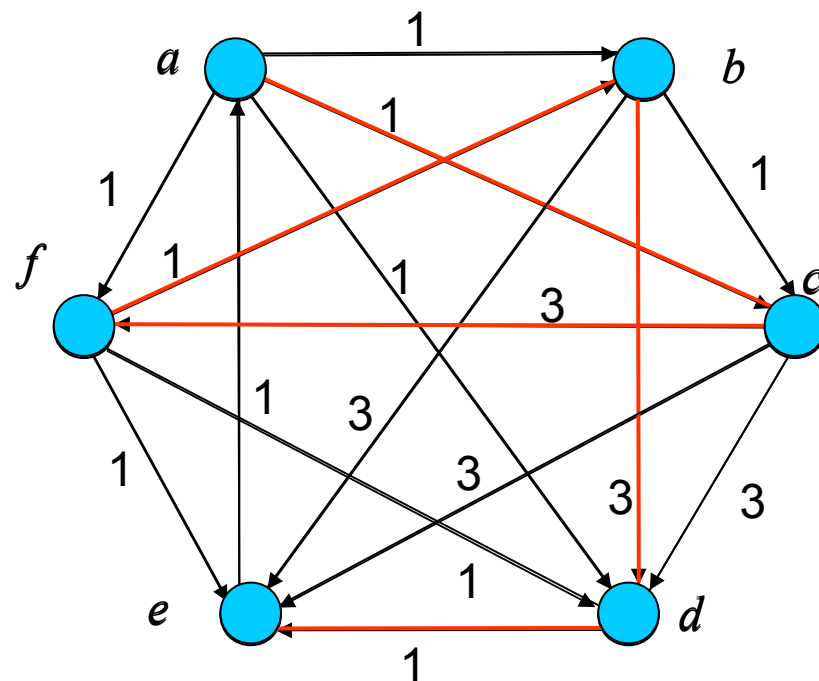
$$\sigma_2 = a > c > f > b > d > e$$

$$\sigma_3 = e > d > a > f > b > c$$

$$\sigma_4 = b > c > d > e > f > a$$

$$\sigma_5 = c > f > b > e > a > d.$$

Here,  $a > c > f > b > d > e$  is a median permutation and induces a **Hamiltonian path**.



# Other components to prune the BB-tree

- We compute the variation of  $\rho_{\Pi}$  when, from a permutation  $\sigma$  beginning with  $S = x_{j1} >_{\sigma} x_{j2} >_{\sigma} \dots >_{\sigma} x_{jp}$ , we take an interval  $x_{jh} >_{\sigma} \dots >_{\sigma} x_{jp}$  ( $1 \leq h \leq p$ ) and we shift it at the end of  $\sigma$ , after the elements of  $X - S$  ( $= OS = \ll \text{out of section} \gg$ ):

$$\sigma = x_{j1} >_{\sigma} x_{j2} >_{\sigma} \dots x_{jh-1} >_{\sigma} x_{jh} >_{\sigma} \dots >_{\sigma} x_{jp} >_{\sigma} (OS)$$

becomes

$$\sigma' = x_{j1} >_{\sigma'} x_{j2} >_{\sigma'} \dots x_{jh-1} >_{\sigma'} (OS) >_{\sigma'} x_{jh} >_{\sigma'} \dots >_{\sigma'} x_{jp}.$$

If  $\rho_{\Pi}$  decreases, we do not keep the node associated with  $S$ .

*OSmoves* will count this kind of cuts.

# Other components to prune the BB-tree (end)

- When we deal with a new beginning section

$$S = x_{j1} >_{\sigma} x_{j2} >_{\sigma} \dots x_{jh-1} >_{\sigma} \textcolor{red}{x_{jh}} >_{\sigma} \dots >_{\sigma} \textcolor{red}{x_{jp}} >_{\sigma} \textcolor{red}{x},$$

we consider the beginning sections that we can get by moving, inside  $S$ , an “interval” of  $S$  including  $x$ , i.e., the beginning sections with the following shape:

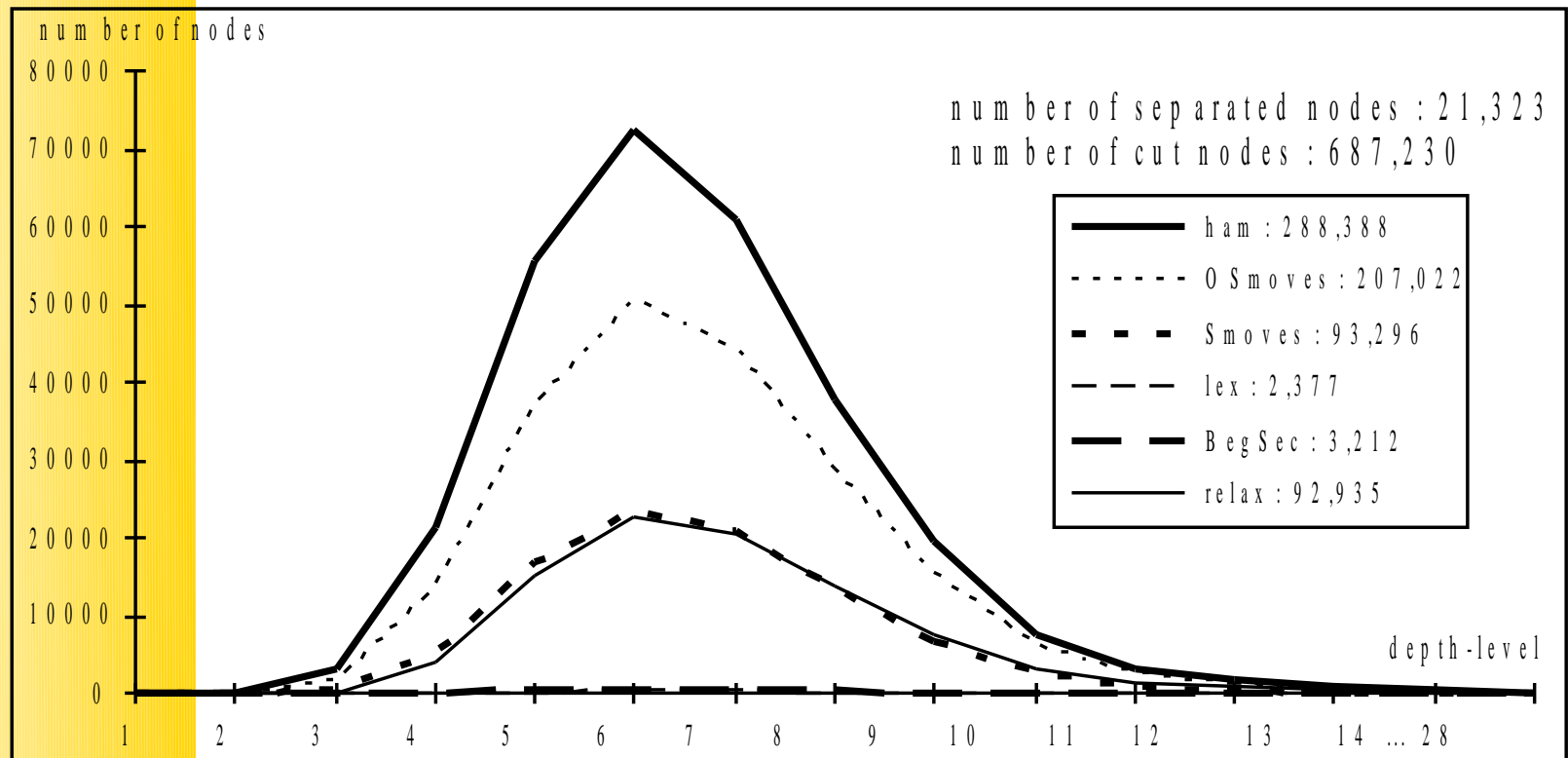
$$\textcolor{red}{x_{jh}} >_{\sigma'} \dots >_{\sigma'} \textcolor{red}{x_{jp}} >_{\sigma'} \textcolor{red}{x} >_{\sigma'} x_{j1} >_{\sigma'} x_{j2} >_{\sigma'} \dots x_{jh-1}.$$

If  $\rho_{\Pi}$  decreases, we do not keep the node associated with  $S$ .

*Smoves* will count this kind of cuts.

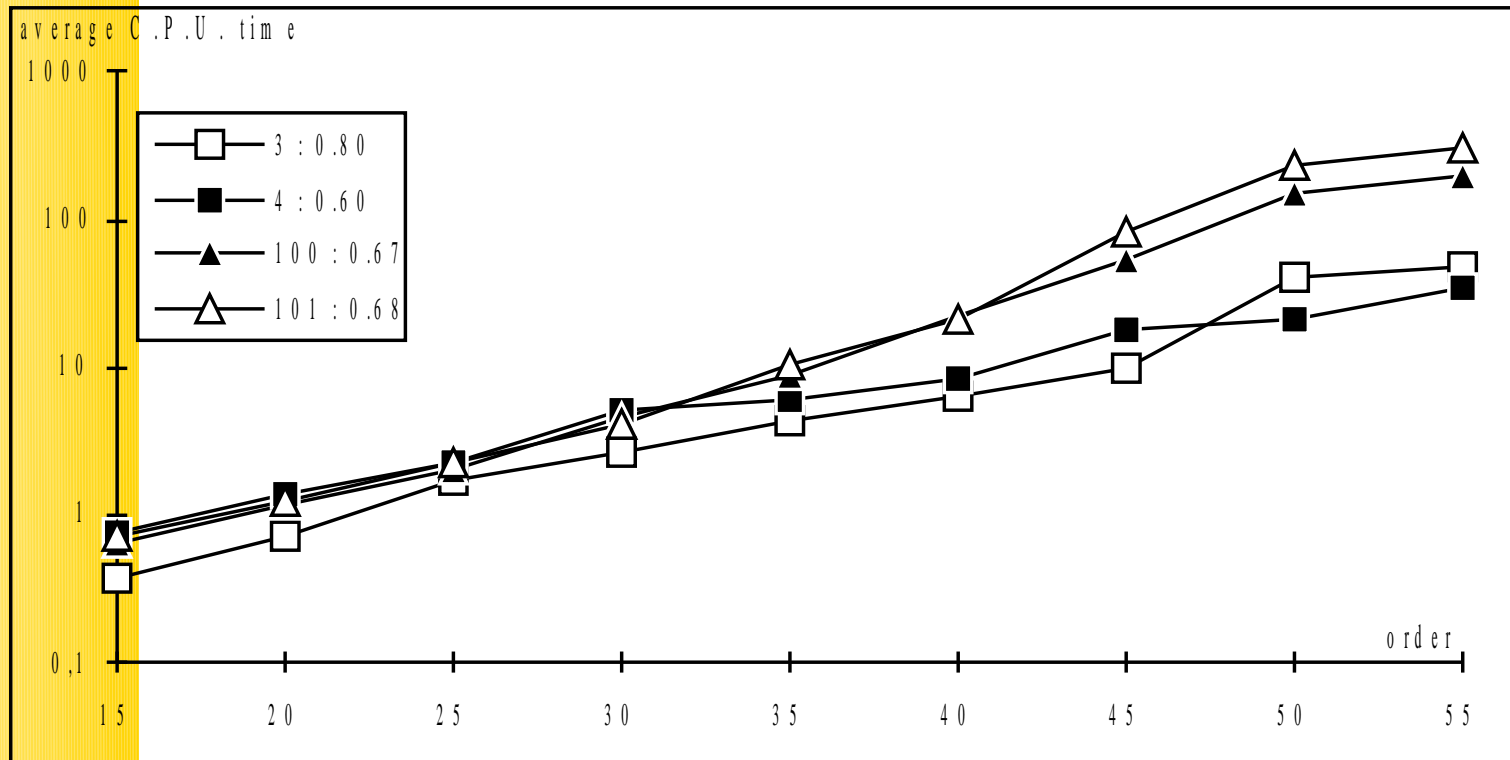
# An experimental result on the efficiency of the branch and bound components

- Numbers of cuts for an instance on 39 candidates



# CPU times for $m \in \{3, 4, 100, 101\}$

- CPU times in seconds (Rk: order =  $n$ ).





# Number of median permutations versus number of Hamiltonian permutations

- Let  $M(n)$  and  $H(n)$  denote respectively the maximum number of median permutations or of Hamiltonian permutations for instances on  $n$  candidates.
- If  $n$  is even with  $n \geq 2$ :  $M(n) = n!$
- If  $n$  is odd:  $M(n) \leq H(n)$ .
- **Theorem** (N. Alon, 1990):  $H(n) \leq (c \times n^{1.5} \times n!)/2^n$  where  $c$  is a constant.
- **Theorem** (I. Charon, O. Hudry, 2000): for  $n = 3^k$ ,

$$3^{0.75(n-1)}/n^2 \leq M(n).$$



Thank you for your attention!

