

# Fast Algorithm for Generalized Multinomial Models with Ranking Data

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## Generalized multinomial models

Consider  $d$  basic cells  $c_1, \dots, c_d$ , where  $c_i$  is assigned with cell probability  $p_i$  ( $\sum_{i=1}^d p_i = 1$ ). Suppose cell  $c_i$  is chosen for  $a_i$  times ( $i = 1, \dots, d$ ), then the log-likelihood function is

$$\ell(\mathbf{p}) = \sum_{i=1}^d a_i \log p_i. \quad (1)$$

Completeness condition: Sets of basic cells for selection are always  $\{c_1\}, \dots, \{c_d\}$ . (choose 1 from  $d$ )

If completeness condition is violated:

- Union of sets for selection includes only a fraction of basic cells. (choose 1 from  $k < d$ )
- Sets for selection consists of more than one basic cell. (choose  $l > 1$  from  $d$ )

→ Incomplete multinomial models.

## Log-likelihood function

$$\ell(\mathbf{p}) = \sum_{j=1}^n \left\{ \log\left(\sum_{c_i \in A^j} p_i\right) - \log\left(\sum_{c_i \in \mathcal{C}^j} p_i\right) \right\}, \quad (2)$$

where  $\mathcal{C}^j$  is the union of sets for selection,  $A^j$  is the selected set in  $j$ -th record.

Examples:

- Placett–Luce model [3, 4];
- Bradley–Terry model [1];
- Contingency table model [2].

## Markov chain based algorithm

Denote  $W_i = \{j : c_j \in A^i\}$  and  $L_i = \{j : c_j \in (C^j \setminus A^i)\}$ ,  
 $q_j^+ = \sum_{c_i \in A^i} p_i$  and  $q_j^* = \sum_{c_i \in C^j} p_i$ ,

$$\ell(\mathbf{p}) = \sum_{j=1}^n \left\{ \log(q_j^+) - \log(q_j^*) \right\}.$$

Letting  $\frac{\partial \ell(\mathbf{p})}{\partial p_i} = 0$ , we have

$$\sum_{i' \neq i} p_{i'} \left[ \sum_{j \in W_i \cap L_{i'}} \frac{p_i}{q_j^+ q_j^*} \right] = \sum_{i' \neq i} p_i \left[ \sum_{j \in L_i \cap W_{i'}} \frac{p_{i'}}{q_j^+ q_j^*} \right], \quad (3)$$

$(i = 1, \dots, d).$

# Markov chain based algorithm

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**Algorithm 1** Markov chain based algorithm

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**Input:** Observations  $\{(A^j, \mathcal{C}^j) : j = 1, \dots, n\}$  and calculate  $\{W_i, L_i\}$  for each  $c_i$ .

Initialize  $\mathbf{p} = (1/d, \dots, 1/d)^T$ .

Initialize  $\Sigma(\mathbf{p}) = \mathbf{0}_{d \times d}$ .

**repeat**

**for**  $i \in \{1, \dots, d\}$  **do**

**for**  $i' \in \{1, \dots, d\} \setminus \{i\}$  **do**

      Compute

$$\sigma_{ii'}(\mathbf{p}) \leftarrow \sum_{j \in L_i \cap W_{i'}} \frac{p_{i'}}{q_j^+ q_j^*}$$

**end for**

**end for**

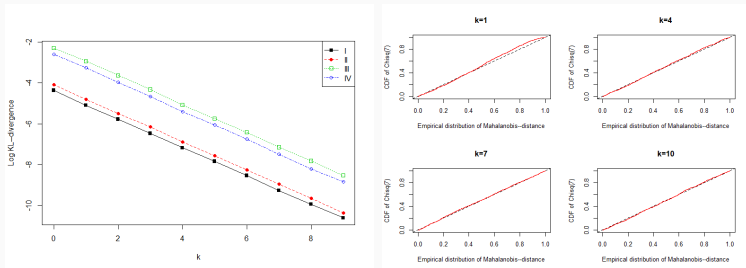
  Compute  $\sigma_{ii}(\mathbf{p})$  ( $i = 1, \dots, d$ ) and then normalize

$\Sigma(\mathbf{p})$  so that  $\forall i, \sum_{i'=1}^d \sigma_{ii'}(\mathbf{p}) = 1$ .

$\mathbf{p} \leftarrow T(\mathbf{p})$  under the transition matrix  $\Sigma(\mathbf{p})$ .

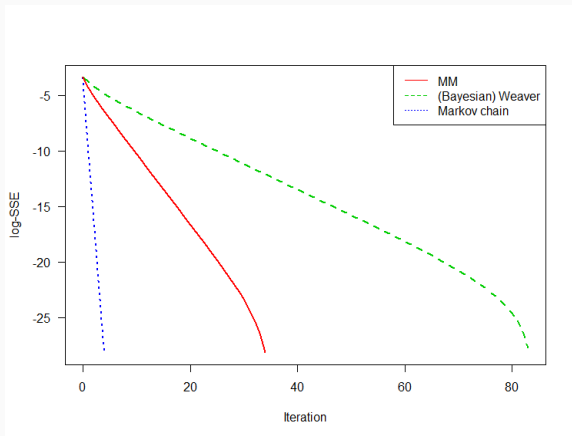
**until** convergence.

# Experiments: Convergence



Estimator obtained by our algorithm is close to the MLE, indicating our algorithm's convergence to the MLE.

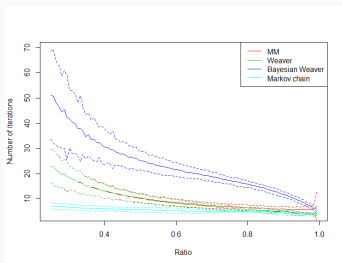
## Experiments: Convergence rate



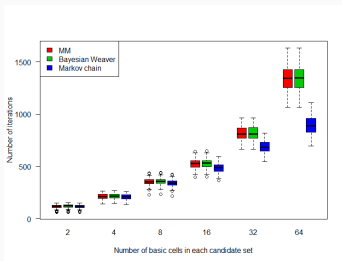
**Figure 1:** Path of iterations for three algorithms on sushi data

# Experiments: Computational efficiency

- Choose 1 from  $k < d$ :



- Choose  $l > 1$  from  $d$ :





- Our algorithm obtain the MLE efficiently than existing methods.
- Further improvement. (Especially in situation "choose  $l > 1$  from  $d''$ ")

## Reference

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End

Thank you for listening.