

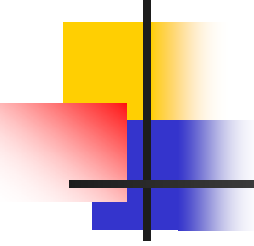
CS 886 Advanced Topics in Artificial Intelligence: Multiagent Systems

Rank Aggregation Methods for the Web

Cynthia Dwork Ravi Kumar Moni Naor D. Sivakumar

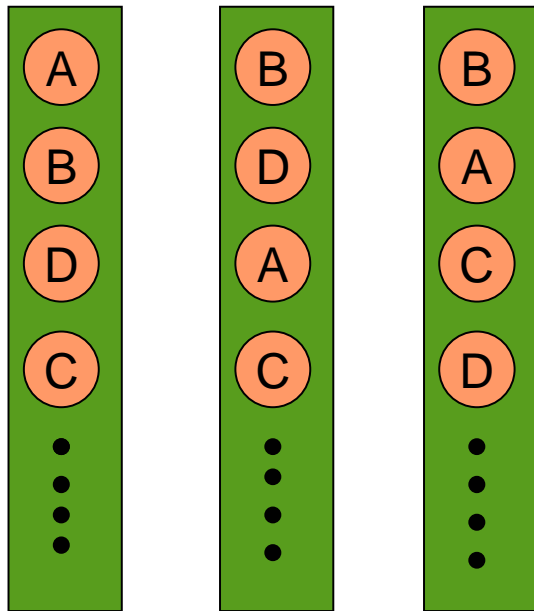
Presented by: Wanying Luo

Outline

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- What is rank aggregation problem
 - Motivation
 - Challenges
 - Preliminaries
 - First result: spam resistance in meta-search
 - Second result: Markov chain methods
 - Applications
 - Experiments
 - Conclusion

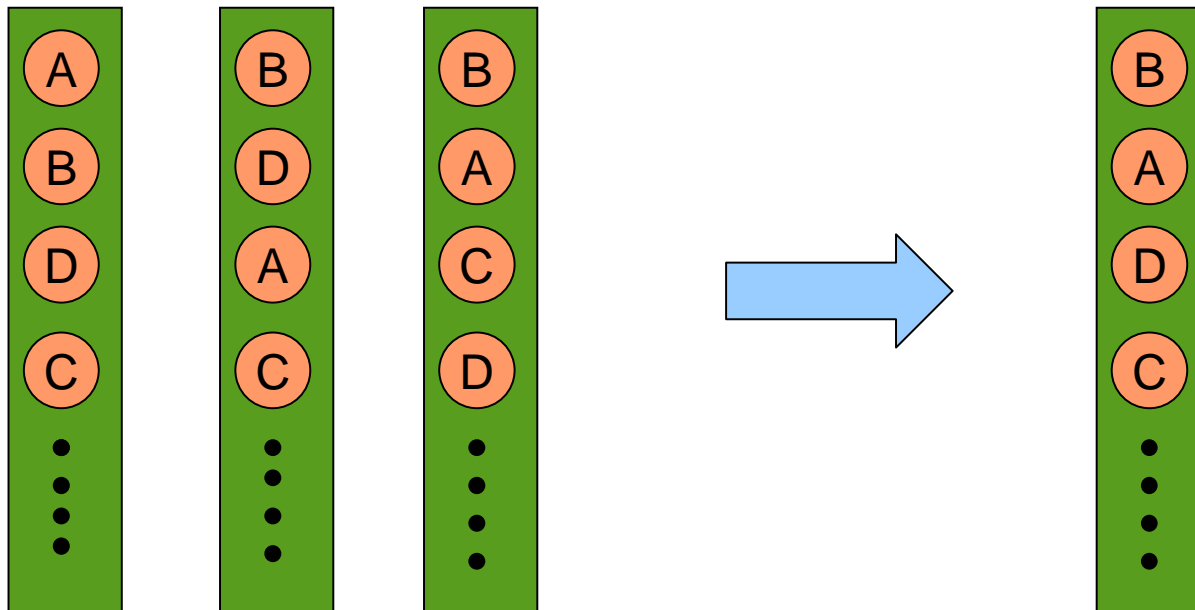
What is rank aggregation problem

Based on different ranking techniques and criteria, we may get different results

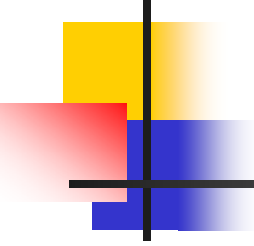


What is rank aggregation problem

Need to obtain a "consensus" ranking of all the individual rankings



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Motivation



- **Provide robust meta search**

- Examples of meta search engines

- ◆ **Clusty**

- ◆ **Dogpile**

- ◆ **Metacrawler**

- Spam

- ◆ <http://searchenginewatch.com/showPage.html?page=3483601>

- Commercial interests, e.g., sponsored links



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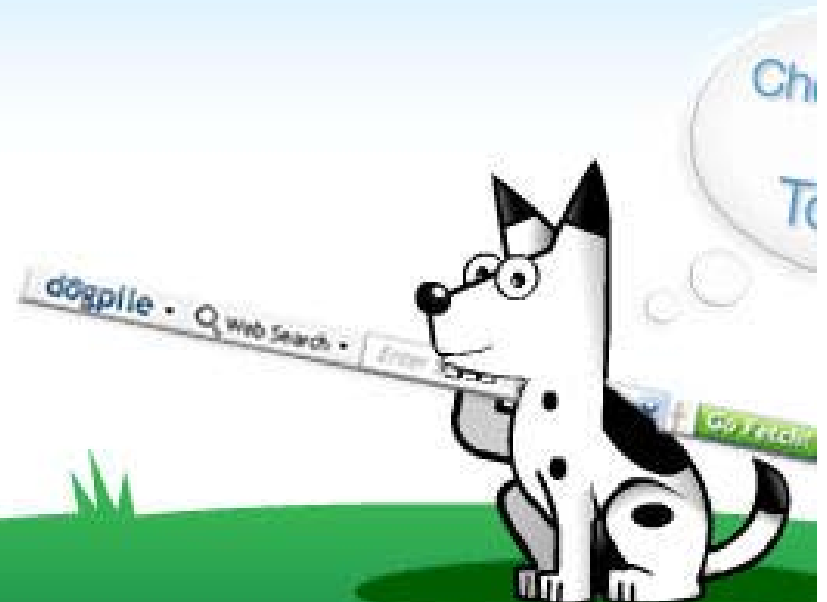
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Motivation



- **Provide robust meta search**

- Examples of meta search engines

- ◆ **Clusty**

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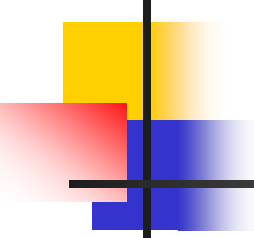
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Motivation

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- **User may provide a variety of searching criteria**

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Challenges



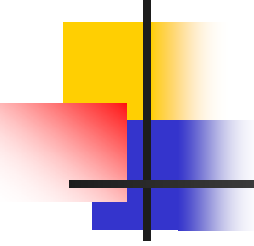
- **Unrealistic to rank the entire collection of pages on the web**
 - **29.7 billion pages on the World Wide Web as of February 2007 (<http://www.boutell.com/>)**
- **Most search engines rank only the top few hundred entries**

Challenges



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Preliminaries



- **Ordered list**

Given a universe U , an ordered list τ with respect to U is an ordering (aka ranking) of a subset $S \subseteq U$, i.e.

$$\tau = [\chi_1 \geq \chi_2 \geq \dots \geq \chi_d]$$

- **Full list**

τ contains all the elements in U

- **Partial list**

$$|\tau| < |U|$$

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- **Rank aggregation approach**

- **Goal: minimize the total disagreement between several rankings**

- Spearman footrule distance

 - ◆ Given two full lists σ and τ , $F(\sigma, \tau) = \sum_{i=1}^n |\sigma(i) - \tau(i)|$

- Kendall tau distance

 - ◆ Given two full lists σ and τ , $K(\sigma, \tau) = |\{(i, j) \mid i < j, \sigma(i) < \sigma(j) \text{ but } \tau(i) > \tau(j)\}|$

- These two measurements can be generalized to several lists

- Can also be generalized to partial lists

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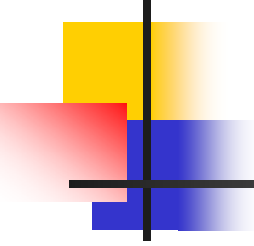
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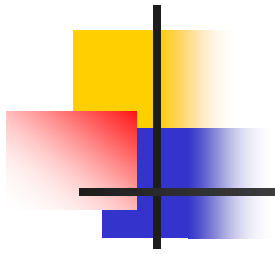
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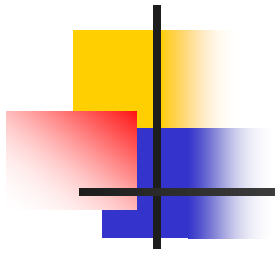
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First result: spam resistance in meta-search

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 - If there is a partition (C, C') of S such that for any $x \in C$ and $y \in C'$ the majority prefers x to y , then x must be ranked above y
 - ECC can be used to fight spam in meta-search
 - How to achieve ECC efficiently
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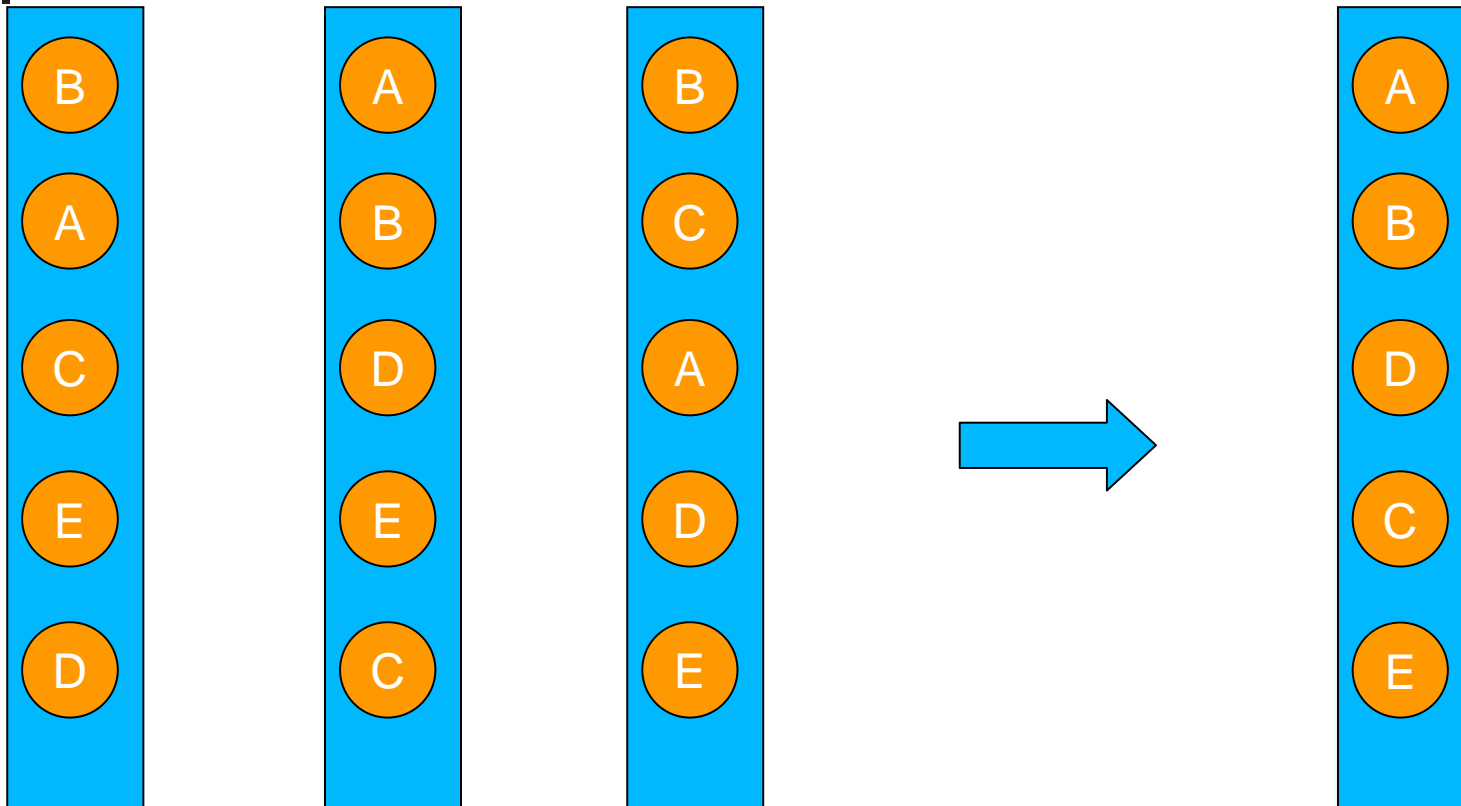
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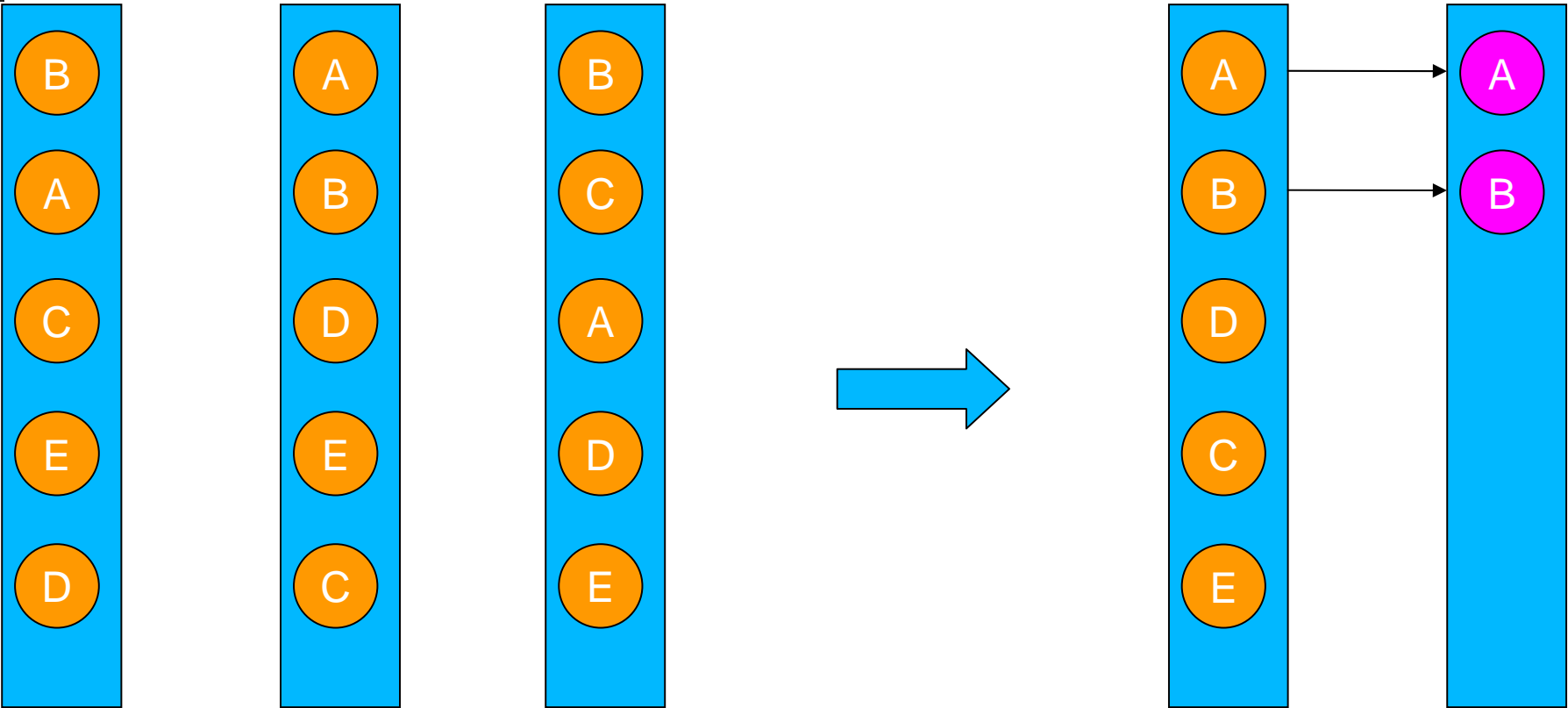
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- **An example to illustrate Local Kemenization ...**

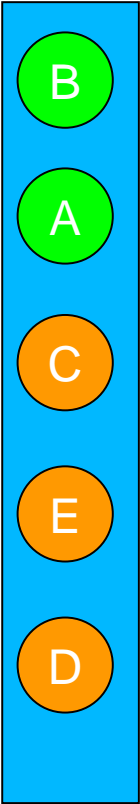
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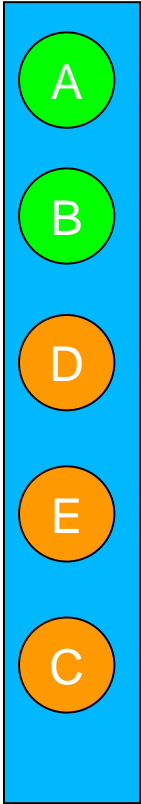
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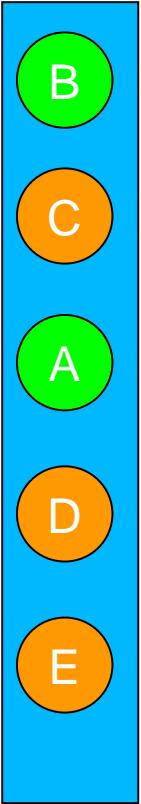
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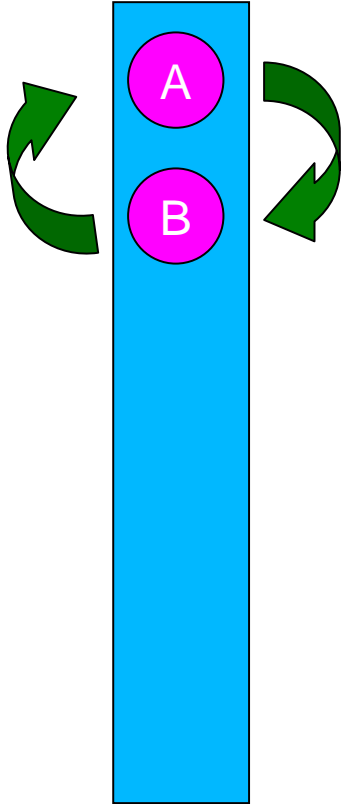
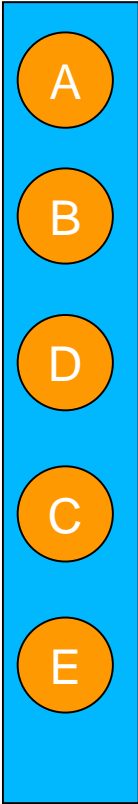
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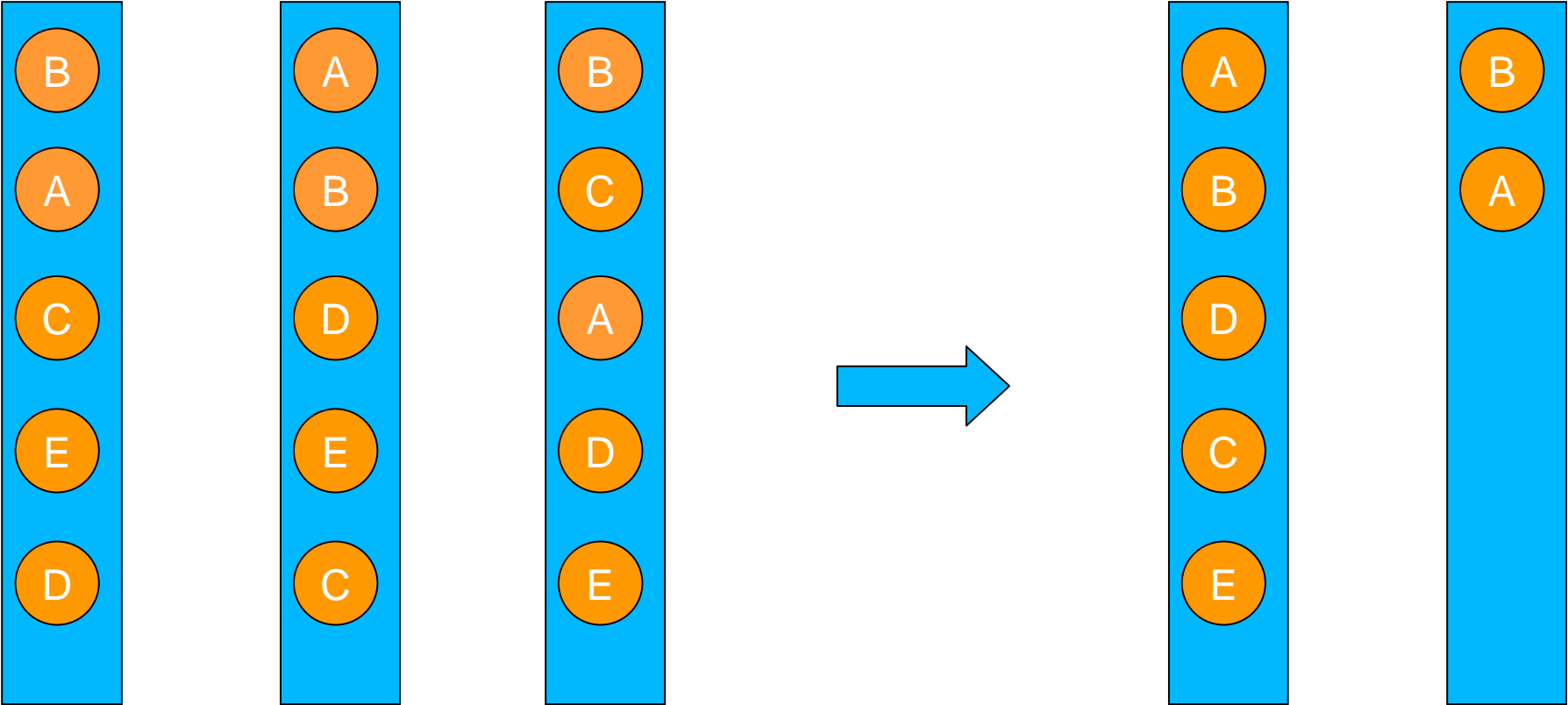
A>B



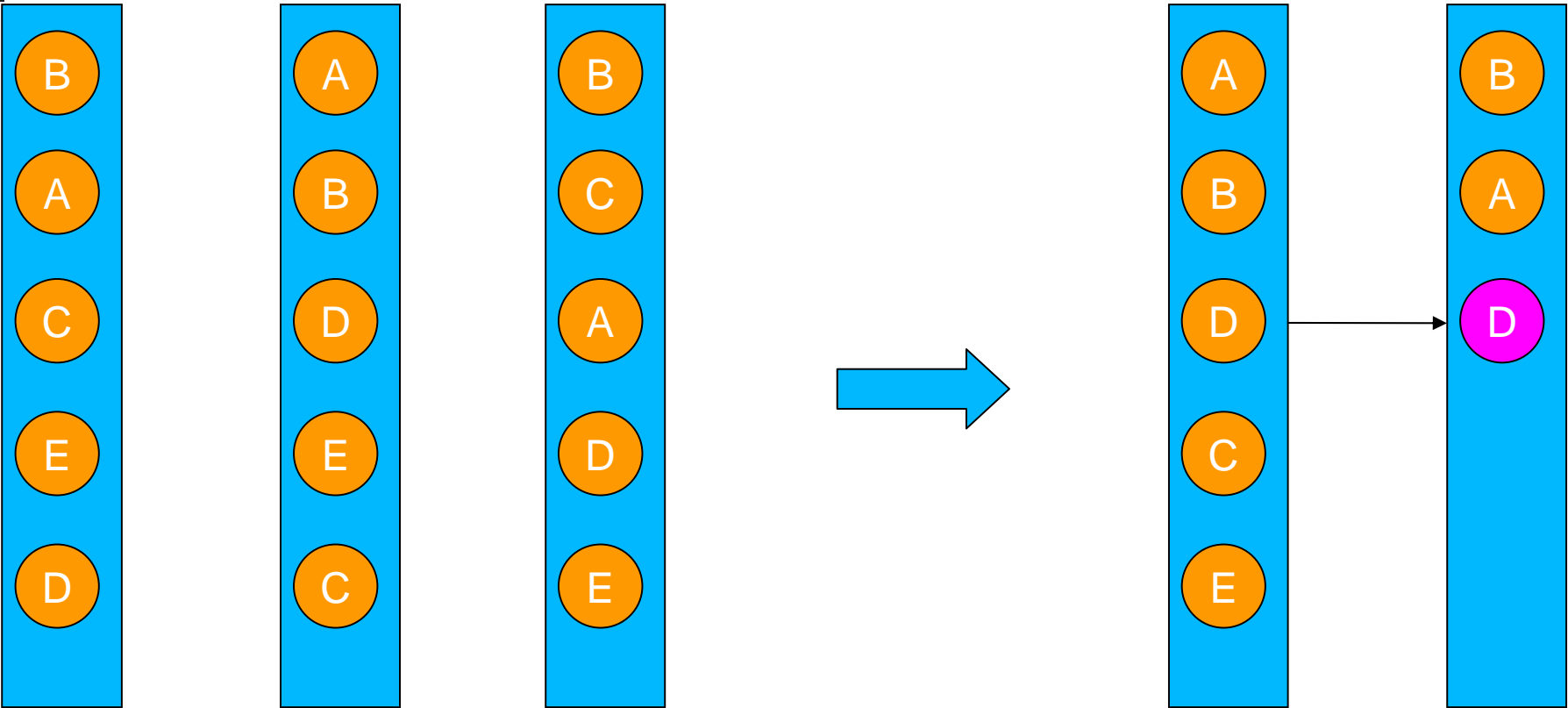
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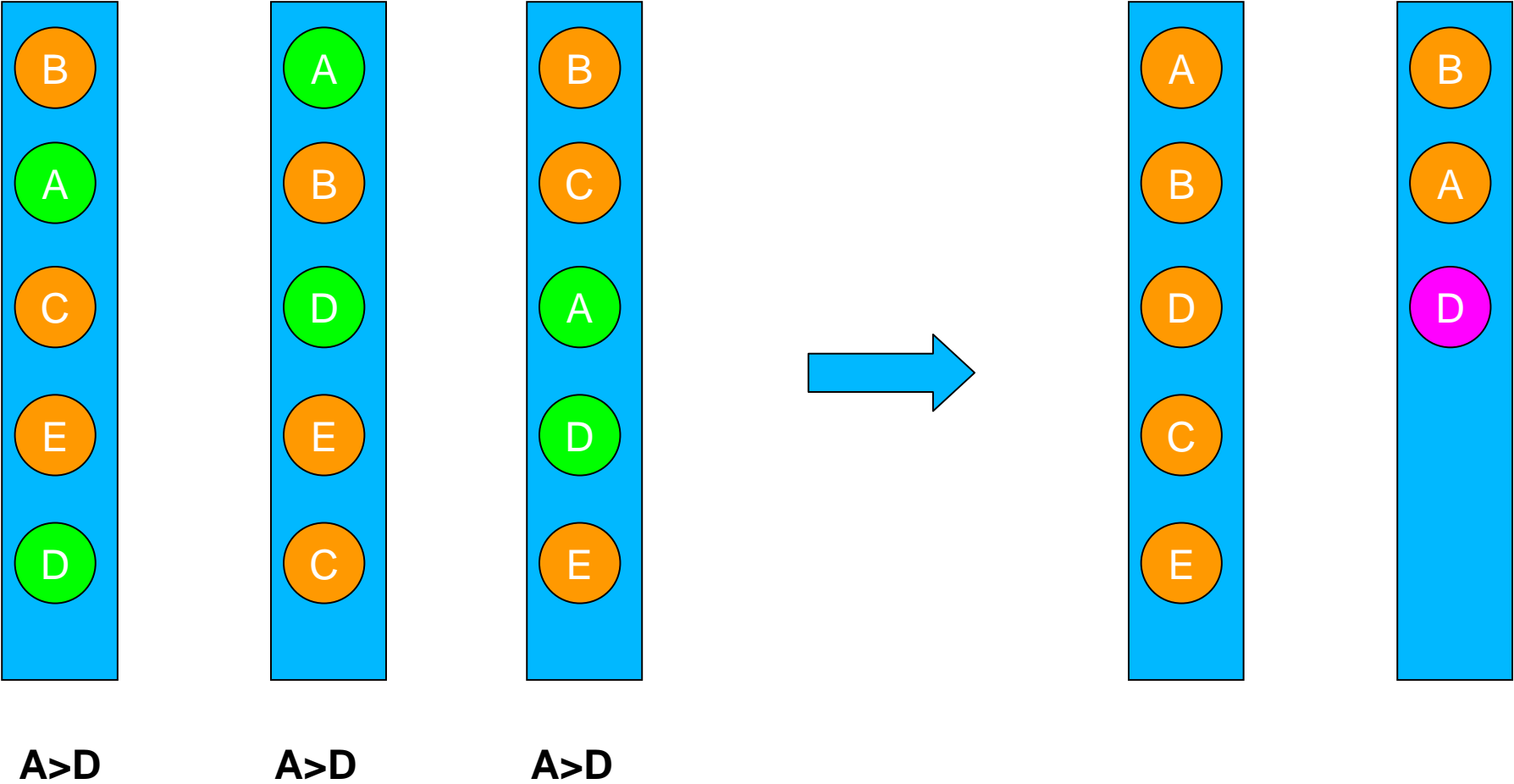
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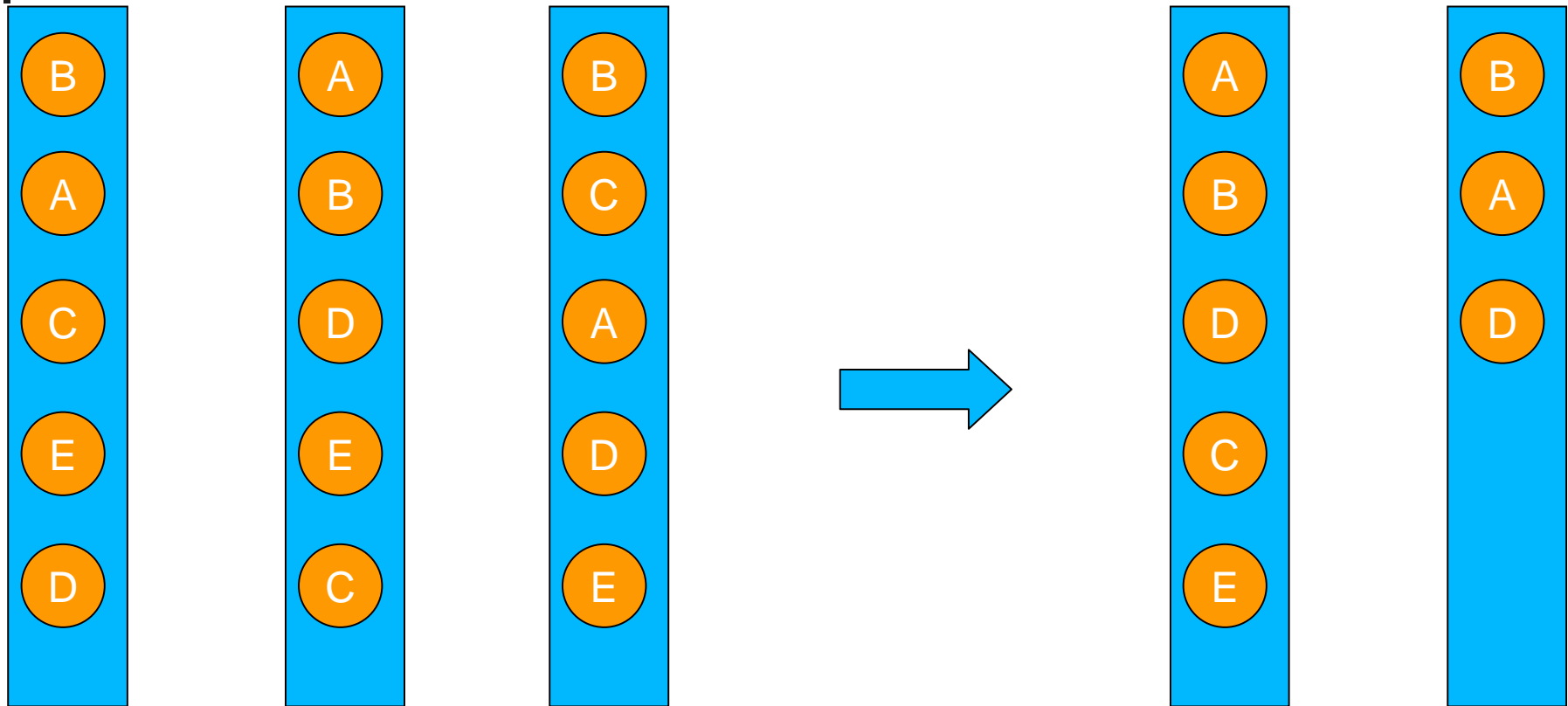
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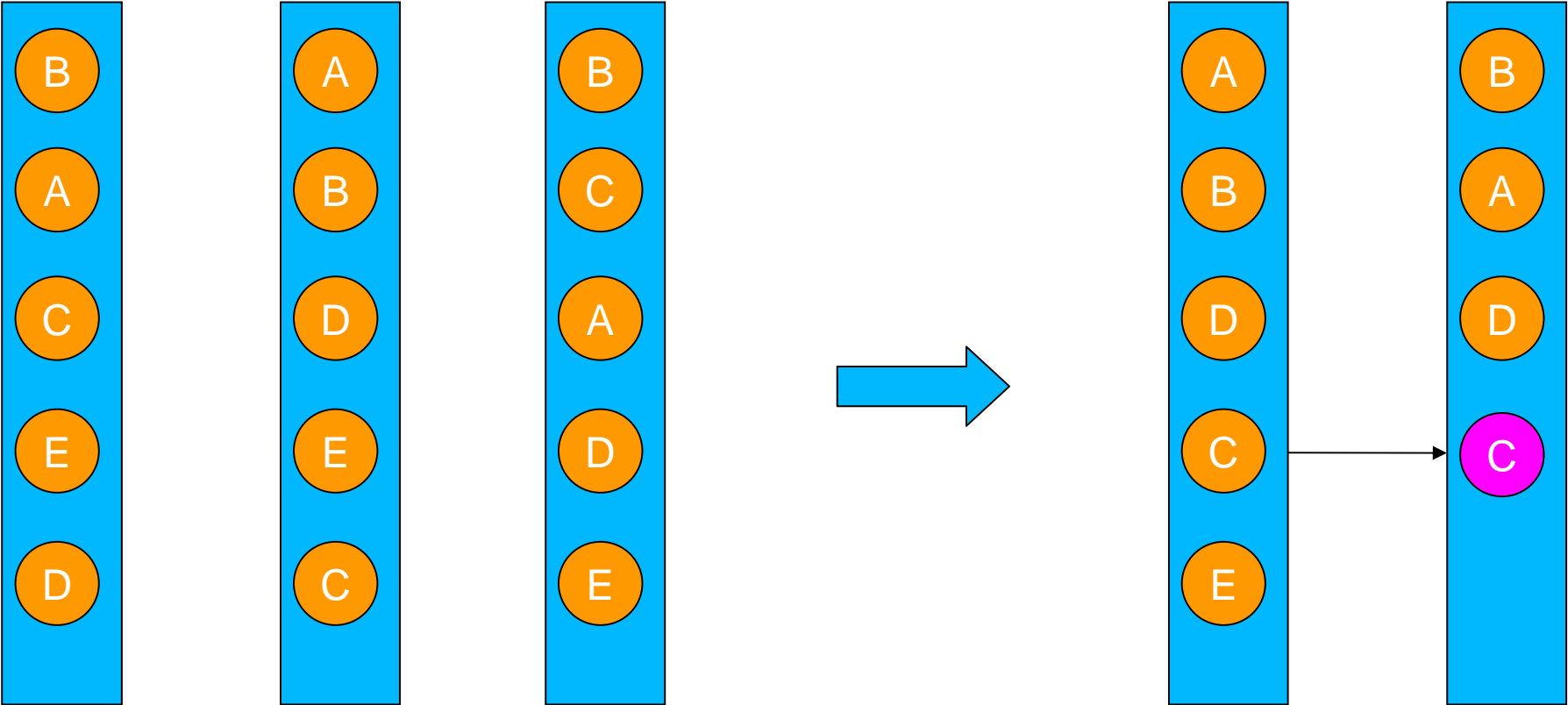
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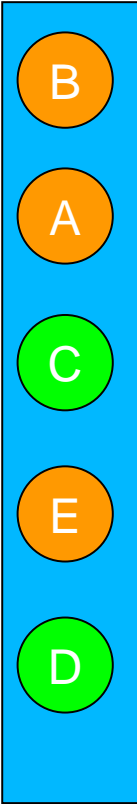
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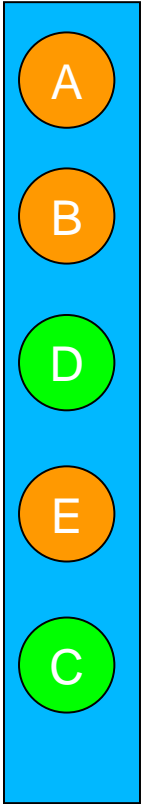
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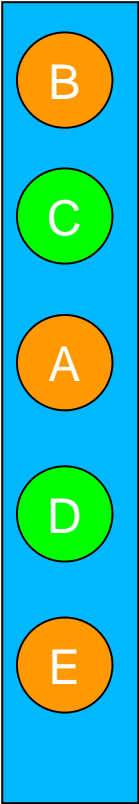
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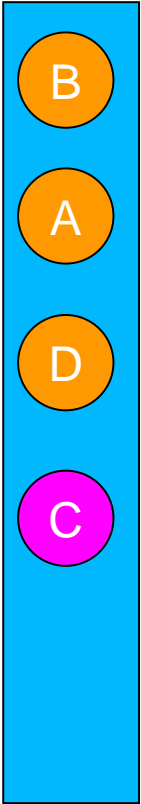
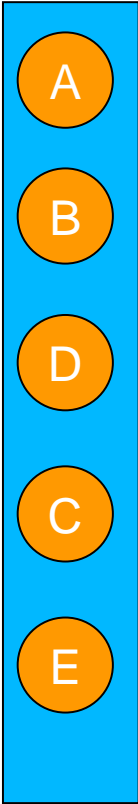
C>D



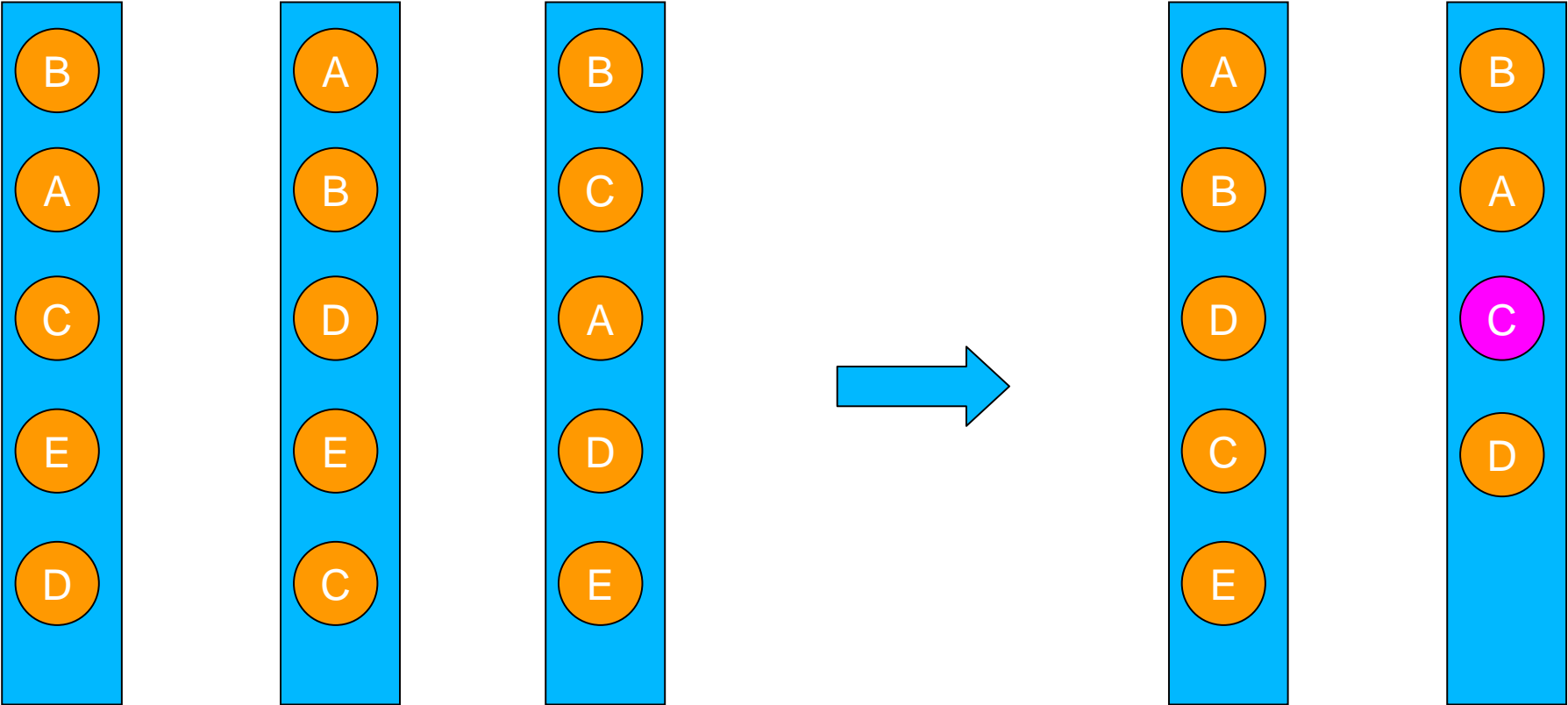
D>C



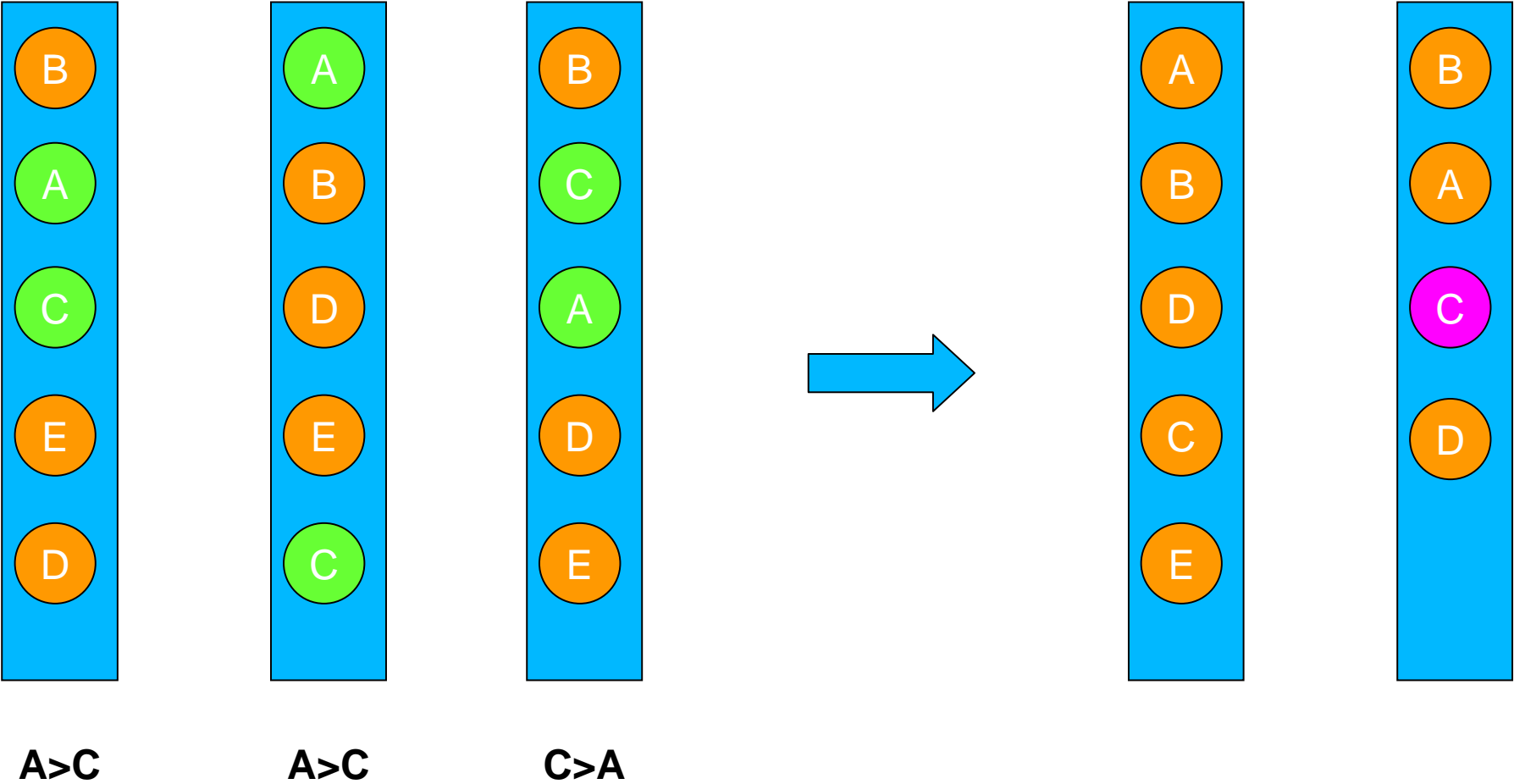
C>D



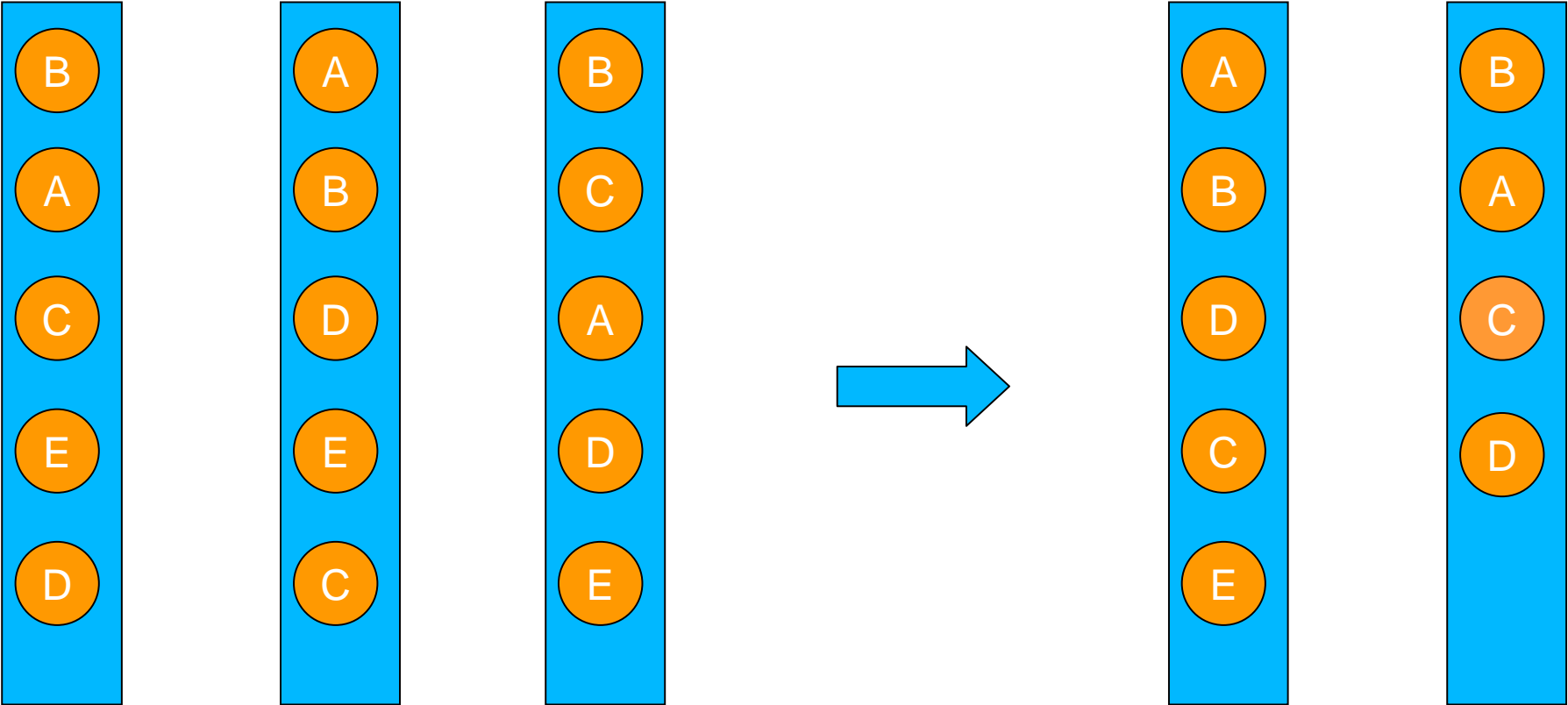
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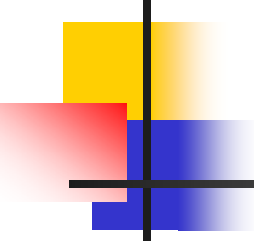
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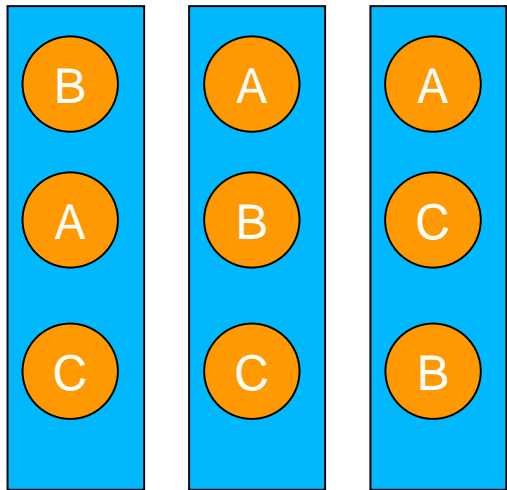
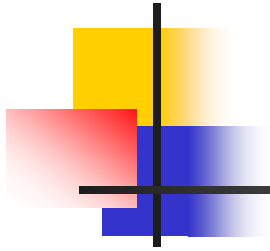
- **Markov chain**

- A set of states $S=\{1,2,\dots,n\}$
- An $n \times n$ matrix M
- Begins with an initial state x
- At each step the system moves from state i to state j with probability M_{ij}

Second result: Markov chain methods

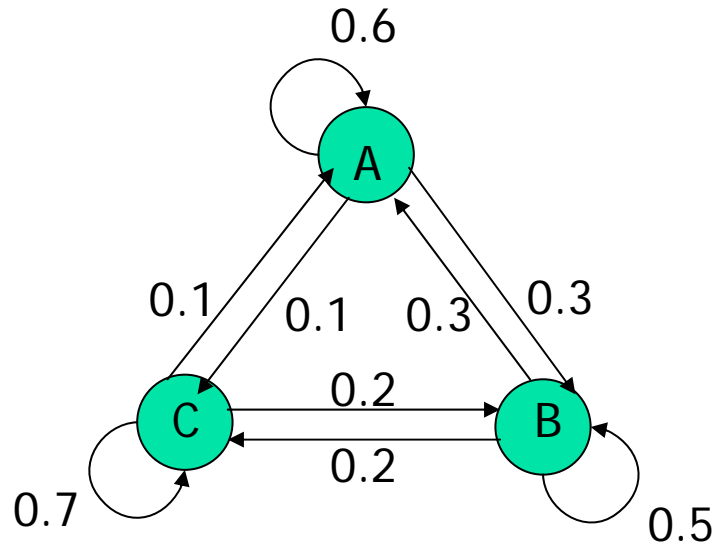
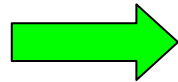
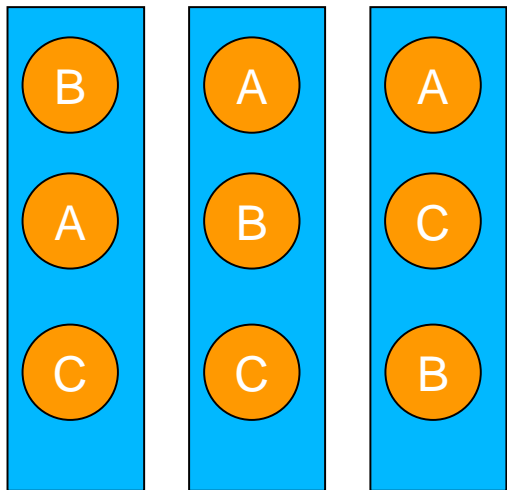
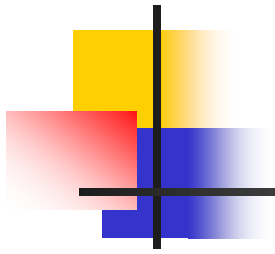
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- **Under some nice condition, system eventually reaches a fixed point irrespective of the initial state x**

Second result: Markov chain methods



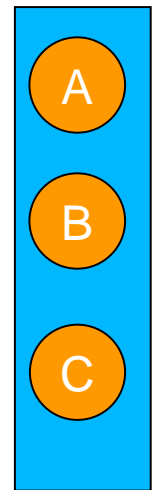
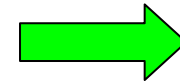
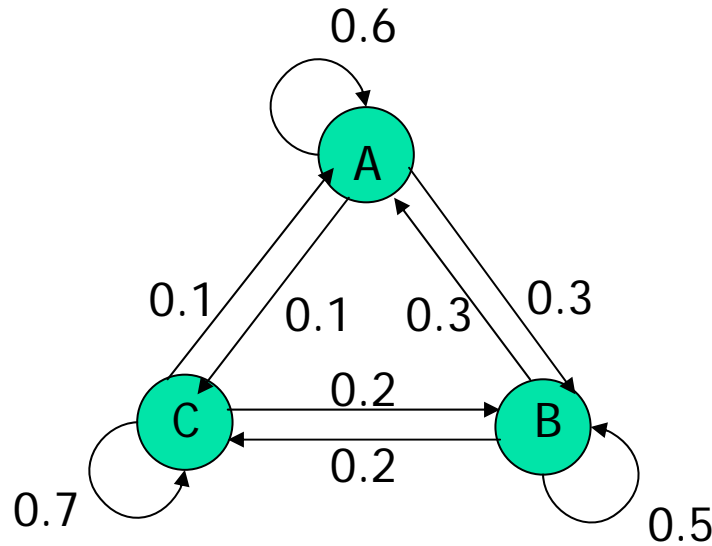
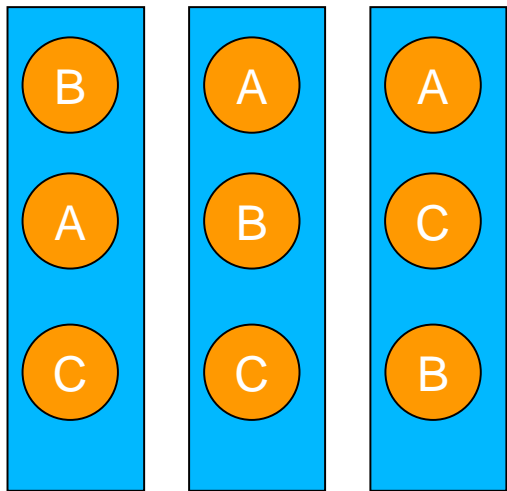
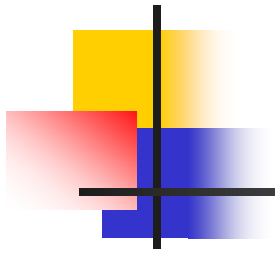
Original rankings

Second result: Markov chain methods



Original rankings

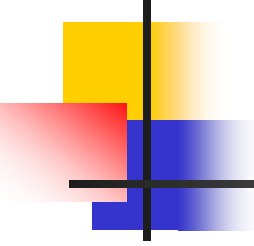
Second result: Markov chain methods



Original rankings

Aggregated ranking

Second result: Markov chain methods

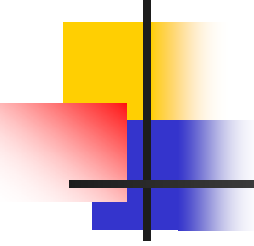
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 - MC_1 : The next state is chosen uniformly from the multiset of all pages that were ranked higher than or equal to P by some search engine that ranked P
 - Please refer to the paper for the rest ...
 - MC_2
 - MC_3
 - MC_4

Second result: Markov chain methods



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- Multi-criteria search
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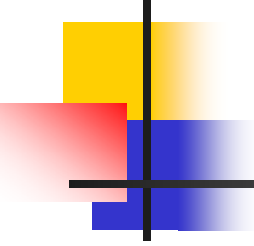
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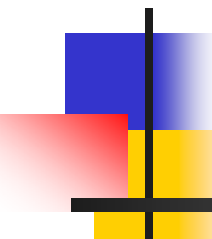
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- Experiment on meta-search using several keywords: “affirmative action”, alcoholism, sushi, ...

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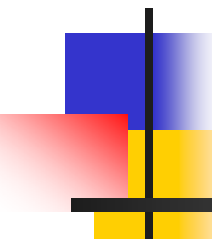
Experiments



	K		IF		SF	
	- LK	+ LK	- LK	+ LK	- LK	+ LK
Borda	0.221	0.214	0.353	0.345	0.440	0.438
SFO	0.112	0.111	0.168	0.167	0.137	0.137
MC ₁	0.133	0.130	0.216	0.213	0.292	0.291
MC ₂	0.131	0.128	0.213	0.210	0.287	0.286
MC ₃	0.116	0.114	0.186	0.183	0.239	0.239
MC ₄	0.105	0.104	0.151	0.149	0.181	0.181

Table 2: Performance of various rank aggregation methods for meta-search. “K” is Kendall distance, “IF” is induced footrule distance, and “SF” is scaled footrule distance. “- LK” and “+ LK”, respectively, denote without and with Local Kemenization.

Experiments



	K		IF		SF	
	- LK	+ LK	- LK	+ LK	- LK	+ LK
Borda	0.221	0.214	0.353	0.345	0.440	0.438
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MC ₄	0.105	0.104	0.151	0.149	0.181	0.181

SFO and MC₄ outperform the other 4 algorithms

MC₄ performs better than SFO most of the time

Experiments



- **Experiment on spam reduction using queries:
Feng shui, organic vegetable, gardening**

Experiments

url	AV	AW	GG	HB	LY	NL	SFO	MC ₄
www.lucky-bamboo.com	4	43			41		144	63
www.cambriumcrystals.com		9	51		5		31	59
www.luckycat.com	11	14	26		13		49	36
www.davesorganics.com	84	19	1		17		77	93
www.frozen.ch		9		63	11		49	121
www.eonseed.com		18		6	16		23	66
www.augusthome.com	26	16		27	12	16	57	54
www.taunton.com		25			21		78	67
www.egroups.com		34			29		108	101

Experiments

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Experiments

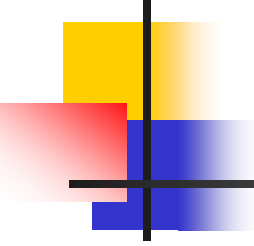
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Local Kemenization works!

Outline

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- What is rank aggregation problem
 - Motivation
 - Challenges
 - Preliminaries
 - First result: spam resistance in meta-search
 - Second result: Markov chain methods
 - Applications
 - Experiments
 - Conclusion

Conclusion



- **Proposed several rank aggregation techniques using Markov chain**
- Established the value of Extended Condorcet Criterion (ECC)
 - Spam resistance
- Future work
 - Obtain a qualitative understanding of why Markov chain methods perform well

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- **Established the value of Extended Condorcet Criterion (ECC)**
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- **Future work**
 - **Obtain a qualitative understanding of why Markov chain methods perform well**

Questions???

