

When the Universe is Too Big

Bounding Consideration Probabilities

for Plackett–Luce Rankings

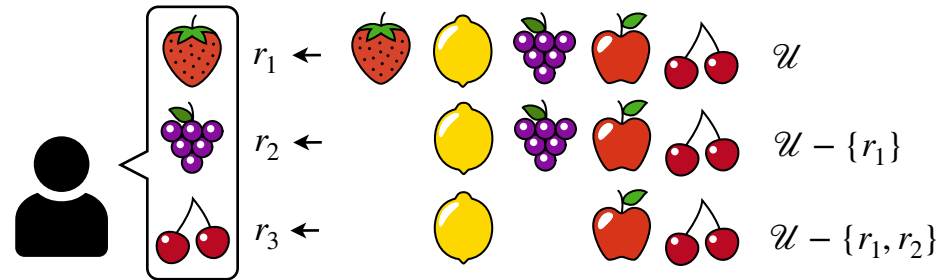
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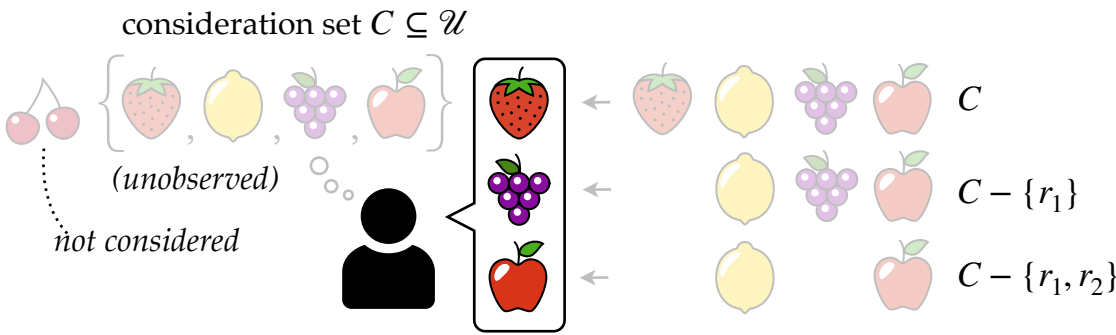


Background and Model

Choice-based ranking: top- k ranking r from universe \mathcal{U}



Consider-then-choose ranking: select a consideration set $C \subseteq \mathcal{U}$ to consider, then rank k elements from C .



Consideration Set Model

$$\Pr_C(C) = 0 \quad \text{if } |C| < k$$
$$\Pr_C(C) \propto \left(\prod_{i \in C} p_i \right) \prod_{j \in \mathcal{U} - C} (1 - p_j)$$

Selection Model:
Plackett–Luce [1, 2]

$$\Pr_{PL}(r | C) = \prod_{i=1}^k \frac{\exp(u_{r_i})}{\sum_{j \in \mathcal{U} - \{r_1, \dots, r_{i-1}\}} \exp(u_j)}$$

Plackett–Luce with Consideration (PL+C)

1. People consider at least k items
2. Each item i is considered independently w.p. p_i
3. Given C , items are ranked by Plackett–Luce based on utility u_i

$$\Pr_{PL+C}(r) = \sum_{C \subseteq \mathcal{U}} \Pr_C(C) \Pr_{PL}(r | C)$$

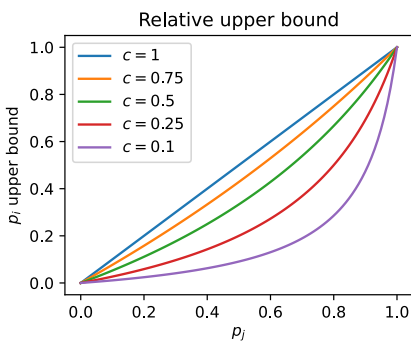
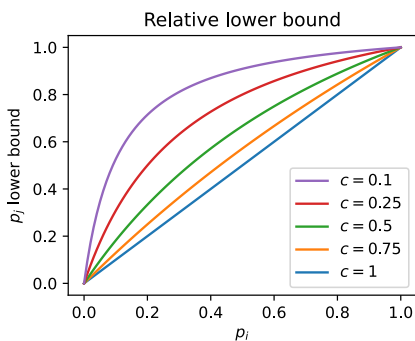
Q: Can we tell from rankings what items were considered?
A: No. ... and yet, we can still derive:

1. **relative bounds** on consideration probabilities, given known utilities.
2. **absolute bounds** on consideration probabilities, given known utilities and a lower bound on expected number of items considered.
3. **algorithms** to tighten our absolute bounds using our relative bounds.
4. **experimental results** based on psychological study data on Americans' perceptions of states' relative contribution to U.S. history [3].

Theorem 1: We can't learn consideration probabilities (in general). PL+C consideration probabilities aren't identifiable, even with known utilities.

Theorem 2: We can use "ranking flips" to infer relative consideration.

If $u_i > u_j$ but i is ranked in the top- ℓ $c \leq 1$ times as often, then $\frac{p_i}{1-p_i} \leq c \cdot \frac{p_j}{1-p_j}$. Equivalently, $p_i \leq \frac{cp_j}{1-p_j+cp_j}$ and $p_j \geq \frac{p_i}{c-cp_i+p_i}$.



References

- [1] Plackett. The analysis of permutations. J R Stat Soc Ser C Appl Stat, 1975.
- [2] Luce. Individual Choice behavior: A theoretical analysis, Wiley 1959.
- [3] Putnam, Ross, Soter, and Roediger. Collective Narcissism: Americans Exaggerate the Role of Their Home State in Appraising U.S. History Psychological Science, 2018.

Theory, Continued

If $\sum_{i \in \mathcal{U}} p_i \geq \alpha k$ for $\alpha > 1$... $\Pr[i \text{ appears in top-}k \text{ rankings}]$

Theorem 3. ... then $\Pr_{PL+C}(\mathcal{R}_{i \leq k}) \cdot \left[1 - (\alpha e^{1-\alpha})^k \right] \leq p_i$.

$1 / \Pr[i \text{ ranked first}]$ (ignoring consideration) $\Pr[i \text{ ranked first}]$ correction (Chernoff)

Theorem 4. ... then $p_i \leq \frac{\sum_{j \in \mathcal{U}} \exp(u_j)}{\exp(u_i)} \cdot \left[\Pr_{PL+C}(\mathcal{R}_{i=1}) + \frac{k(\alpha e^{1-\alpha})^k}{1 - (\alpha e^{1-\alpha})^k} \right]$.

Thm. 3 (absolute lower bound): i must have been considered whenever it's ranked + correction for conditioning.
Thm. 4 (absolute upper bound): if i was ranked first less than expected, it can't have been considered + correction for conditioning.

Efficient bound propagation algorithm

Given: utilities u_i , top- ℓ ranking probabilities $\Pr_{PL+C}(\mathcal{R}_{i \leq \ell})$, $\alpha > 1$ s.t. $\sum_{i \in \mathcal{U}} p_i \geq \alpha k$ ($\geq \alpha k$ items considered on average).

1. Initialize upper/lower bounds according to Theorem 3/4
2. Construct DAG G of all item reversals (utility vs top- ℓ ranking probability)
3. Propagate bounds per Thm. 2 along topological order of G

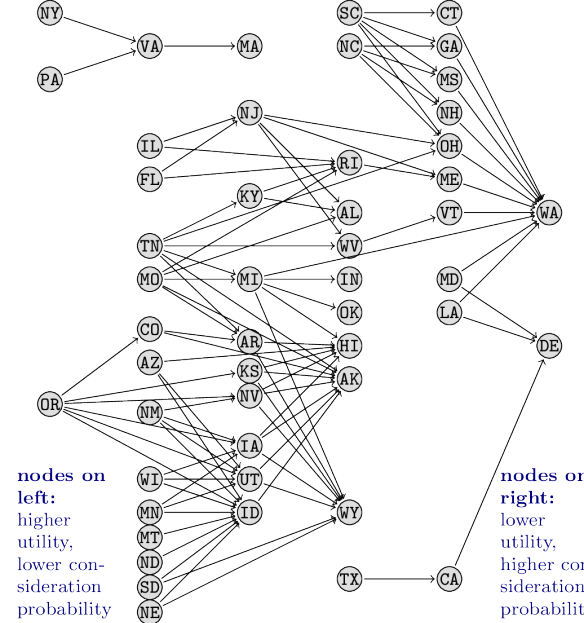
Experiments

Survey of Americans' perceptions of U.S. History [3]:

~3K participants.

Top-3 Q: What three states contributed most to U.S. history?
ranking with consideration

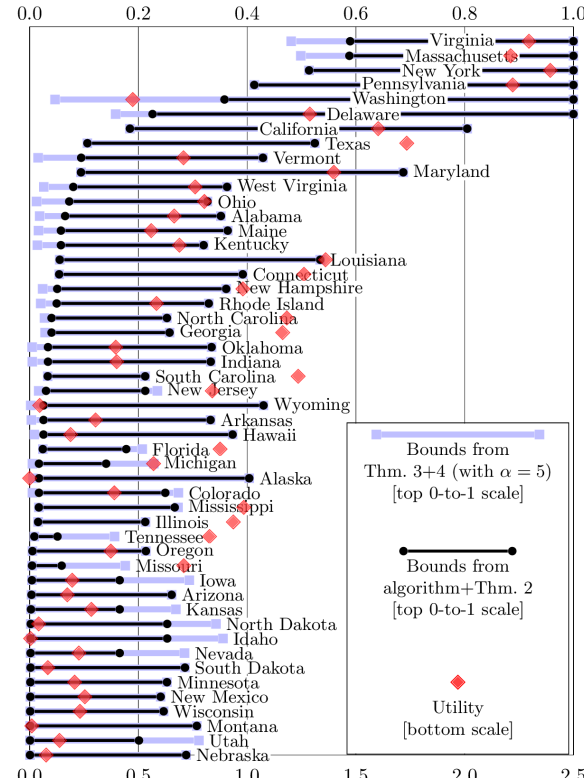
Random-10 Q: What percentage of U.S. history did [10 random states] contribute?
ranking without consideration



learn utilities from Random-10

apply propagation algorithm to Top-3 with $\alpha = 5$ (assume 15+ states considered on average)

bounds on how often each state was considered



Overview

Theoretical Results