

#### Introduction

- Much of social choice theory concerned with aggregating voters' preference ranking (over candidates) into a consensus ranking.
  - Started with Arrow's Theorem (1951)
  - Politics, meta web search, multi-criteria decision making, etc...

 Many aggregation methods: single transferable vote, Borda, Kemeny, ...

#### Introduction

Why need entire consensus ranking?

Why use a particular aggregation method (e.g. Kemeny)?

Decision criterion should directly influence how to aggregate rankings

We develop such an approach.

### Unavailable Candidate Model

(Motivating Example)

Hiring committee (Can hire one)



Job Candidates



Will they take job offer?

Candidate Uncertainty

Preferences

DECISION MAKER

Output Ranking



### Unavailable Candidate Model

(Recently, independently developed by Baldiga & Green)



Candidates  $C = \{c_1, ..., c_m\}$ 



Voters N =  $\{1..n\}$  with preference profile V =  $(v_1, ..., v_n)$  where  $v_i$  is a ranking



Probability P(S) only  $S \subseteq C$  are available.

Can't just select winning candidate!



Output a decision policy (aka choice function)

$$W:2^C \rightarrow C \cup \{\bot\}$$

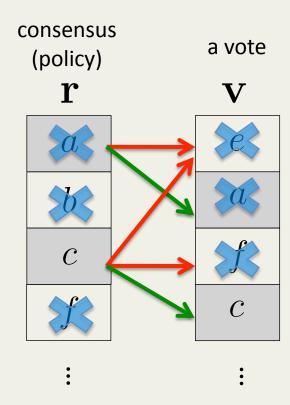
#### Unavailable Candidate Model

- Consider ranking policies: outputs top available candidate (aka rationalizable choice function)
- Other policy classes possible but rankings are compact, interpretable, easy to implement.
- Sometimes legally, or procedurally required
  - National Resident Matching Program, hospitals submit preference ranking (committee in hospital must reach *consensus*).
- Rationalizable approximation to optimal policy

## Minimizing Disagreement

Find consensus ranking minimizing expected #disagreements

Disagreement Example



# Comparison with Kendall-Tau (Kemeny)

#### **Defn (Kendall-Tau distance)**

$$\tau(r,v) = \sum_{\{i,j\}} 1[r,v \text{ disagree on } i,j]$$

#### **Defn (Kemeny consensus)**

Ranking 
$$r$$
 that minimizes  $K(r,V) = \sum_{\ell=1}^{n} \tau(r,v_{\ell})$ 

# Comparison with Kendall-Tau (Kemeny)

Kendall-tau (=#misordered pairs)

Misordered {a, e} has same "penalty" as misordered {c, f}

(policy) r a b c f

consensus

 $egin{array}{c} \mathbf{v} \\ e \\ \hline a \\ \hline f \\ \hline c \\ \hline \end{array}$ 

a vote

Our expected #dis.

Prob. of disagreement

"**r** outputs a,  $\mathbf{v}$  wants e"

is larger than

"r outputs c, v wants f"

Kendall-tau ignores *relative importance of misordering* (how far down in ranking). In most cases,
our cost focuses more on disagreements at top

# Minimizing Disagreement

• Given P, V, minimize expected #disagreements

$$r^* = \operatorname{argmin}_r \operatorname{E}_{S \sim P} \left[ \sum_{\ell=1}^n \mathbf{1}[r(S) \neq \operatorname{top}(v_\ell, S)] \right]$$
disagreement

- Simple P: prob. p any candidate is unavailable
- Can be shown expected #disagreements is

$$\sum_{\ell=1}^{n} \sum_{i=1}^{m} (1-p) p^{r(c_i)-1} \left(1-p^{v_{\ell}(c_i)-t(c_i,r,v_{\ell})-1}\right)$$

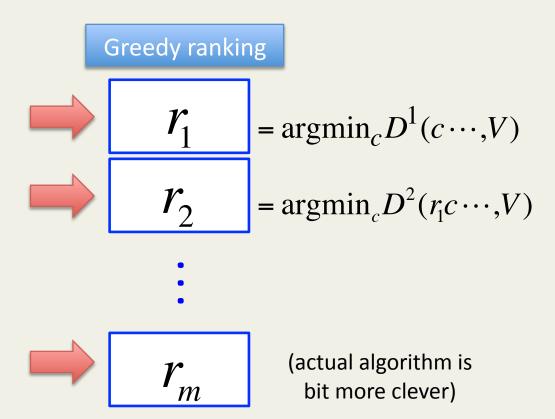
Candidates lower in the ranking contribute less to the expected #disagreements

## **Computational Optimization**

- NP-hard
  - Given p, V, and threshold  $t \ge 0$ , exists ranking r with expected #disagreements ≤ t?
- For optimal ranking
  - Can formulate as integer program
  - Large #vars and constraints (though polynomial)
  - CPLEX is slow to solve
- However, very good greedy algorithm

# **Greedy Algorithm**

• Let  $D^k(r, V)$  be expected #disagreements when  $k^{th}$  ranked candidate of r is top available



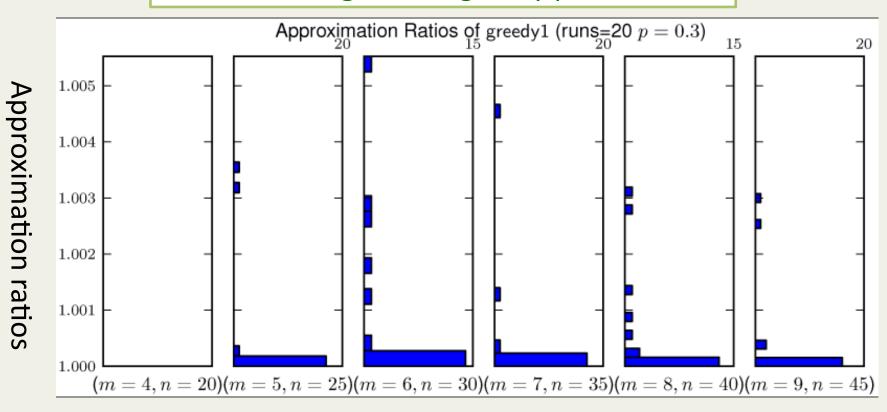
Approximation ratio at most  $1 + 2p/(1-p)^2$ 

Reminder *p* is prob. of unavailable

Bound likely loose, works very well experimentally!

## **Greedy Algorithm**

Rotated Histograms of greedy performance



Varying pairs (m = #candidates, n = #voters)

# PolyTime Approximation Scheme



Idea: get top positions of ranking "right"

Find top K candidates

$$\operatorname{argmin}_{r_1\cdots r_K} \sum_{\ell=1}^n \sum_{k=1}^K D^k(r_1\cdots r_K\cdots,V)$$

Approximation ratio at most  $1 + 2p^{K}/(1-p)^{2}$ 

Make less than  $1 + \varepsilon$  by

$$K = \left[ \log \frac{2}{\varepsilon (1-p)^2} / \log \frac{1}{p} \right]$$

Order remaining arbitrarily

Output ranking

*r*<sub>1</sub>

 $r_2$ 

•

 $r_{K}$ 

 $r_{K+1}$ 

•

 $r_m$ 

# Connection to Plurality Voting

 When p = 0 (all candidates available), then top candidate in an optimal ranking is one that receives the largest number of first place votes.

 A consequence of the definition of "disagreement"

## Connections to Kemeny

Focus of much work in computational social choice

- Max likelihood estimator of a distribution with a modal ranking (Young'95, Mallows'57)
  - Votes are I.I.D. samples of "objective" ranking
  - Motivation is *not* decision-theoretic
  - Aggregation is statistical inference

## Connections to Kemeny

#### Theorem

As p "approaches 1", the following holds

- 1. Any optimal ranking is also a Kemeny consensus
- 2. A Kemeny consensus  $K^*$  may not minimize expected #disagreements, however,
- 3. Any  $K^*$  has expected #disagreements at most factor of  $1 + \varepsilon$  worse than optimal ( $\varepsilon$  depends on p, m, n)

p = 0Plurality

Continuum of aggregation rules

 $p \rightarrow 1$ Kemeny

 Can be used to justify Kemeny as "decision policy under uncertainty" (in some cases)

#### Conclusion

#### **Take Home Message**

Decision criterion as foundation for rank aggregation

In our case, ranking to minimize expected disagreements under uncertain candidate availability

- Connections to Plurality, Kemeny
- •Decision-theoretic justification for Kemeny
- •Nice computational properties—good greedy algorithm, PTAS

#### **Future Work**

- More general class of distributions (e.g. tractable graphical models)
  - Computing expected disagreements
  - •Optimization: approximate, exact
- •Disagreement take into account strength of preferences
- •How good of an approximation is optimal ranking to optimal policy? *Loss of rationalizing choice functions.*
- •Incentive issues reporting preferences and probabilities
- Other decision-theoretic models of social choice

