

# Introduction to Computational Social Choice

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## Social Choice Theory

SCT studies collective decision making: how should we aggregate the preferences of the members of a group to obtain a “social preference”?

$\triangle \succ_1 \circ \succ_1 \square$

$\square \succ_2 \triangle \succ_2 \circ$

$\circ \succ_3 \square \succ_3 \triangle$

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SCT is traditionally studied in Economics and Political Science, but now also by “us”: *Computational Social Choice*.

## Talk Outline

- Computational Social Choice: *research area* and *community*
- Examples for typical research questions (mostly Amsterdam)
- Conclusions and how to find out more

## Computational Social Choice

*Social choice theory* studies mechanisms for collective decision making, such as voting procedures or protocols for fair division.

- Precursors: Condorcet, Borda (18th century) and others
- serious scientific discipline since 1950s
- Classics: Black, Arrow, May, Sen, Gibbard, Satterthwaite, . . .

*Computational social choice* adds a computational perspective to this, and also explores the use of concepts from social choice in computing.

- “classical” papers: ~1990 (Bartholdi et al.)
- active research area with regular contributions since ~2002
- name “COMSOC” and biannual workshop since 2006

Y. Chevaleyre, U. Endriss, J. Lang, and N. Maudet. *A Short Introduction to Computational Social Choice*. Proc. SOFSEM-2007.

## The COMSOC Research Community

- International Workshop on Computational Social Choice:
  - 1st edition: COMSOC-2006 in Amsterdam, December 2006  
48 paper submissions and 80 participants (14 countries)
  - 2nd edition; COMSOC-2008 in Liverpool, September 2008  
55 paper submissions and ~80 participants (~20 countries)
  - 3rd edition: COMSOC-2010 in Düsseldorf, September 2010  
Paper submission deadline: 15 May 2010
- Special issues in international journals:
  - *Mathematical Logic Quarterly*, vol. 55, no. 4, 2009
  - *Journal of Autonomous Agents and Multiagent Systems*, 2010
- Journals and conferences in AI, MAS, TCS, Logic, Econ, ...
- COMSOC website: <http://www.illc.uva.nl/~ulle/COMSOC/>  
(workshop proceedings, related events, mailing list, etc.)

## Example from Voting

Suppose the *plurality rule* is used to decide an election: the candidate receiving the highest number of votes wins.

Assume the preferences of the people in, say, Florida are as follows:

49%: Bush  $\succ$  Gore  $\succ$  Nader

20%: Gore  $\succ$  Nader  $\succ$  Bush

20%: Gore  $\succ$  Bush  $\succ$  Nader

11%: Nader  $\succ$  Gore  $\succ$  Bush

So even if nobody is cheating, Bush will win this election. But:

- In a *pairwise contest*, Gore would have defeated anyone.
- It would have been in the interest of the Nader supporters to *manipulate*, i.e., to misrepresent their preferences.

Is there a better voting procedure that avoids these problems?

## Some Voting Procedures

- *Plurality*: elect the candidate ranked first most often
- *Borda*: each voter gives  $m-1$  points to the candidate they rank first,  $m-2$  to the candidate they rank second, etc., and the candidate with the most points wins
- *Copeland*: award 1 point to a candidate for each pairwise majority contest won and  $\frac{1}{2}$  points for each draw, and elect the candidate with the most points
- *Single Transferable Vote (STV)*: keep eliminating the plurality loser until someone has an absolute majority
- *Approval*: voters can approve of as many candidates as they wish, and the candidate with the most approvals wins

## Complexity as a Barrier against Manipulation

By the *Gibbard-Satterthwaite Theorem*, any voting rule for choosing between  $\geq 3$  candidates can be manipulated (unless it is dictatorial).

Idea: So it's always *possible* to manipulate, but maybe it's *difficult*!

Tools from *complexity theory* can be used to make this idea precise.

- For the *plurality rule* this does *not* work: if I know all other ballots and want  $X$  to win, it is *easy* to compute my best strategy.
- But for *single transferable vote* it does work. Bartholdi and Orlin showed that manipulation of STV is *NP-complete*.

Recent work in COMSOC has expanded on this idea:

- NP is a worst-case notion. What about average complexity?
- Also: complexity of winner determination, control, bribery . . .

J.J. Bartholdi III and J.B. Orlin. Single Transferable Vote Resists Strategic Voting. *Social Choice and Welfare*, 8(4):341–354, 1991.

## Preferences and Ballots

Two common assumptions in voting theory:

- Voters have *preferences* that are *total orders* over candidates.
- Voters vote by submitting a structure just like their preferences, truthfully or not (*ballots* and preferences have *the same* structure).

We may want to drop these assumptions, because:

- For lack of information or processing resources, voters may be *unable to rank* all candidates (in their mind or on the ballot sheet).
- To reduce *complexity of communication*, we may want to design voting rules that work with ballots of bounded size.
- For *approval voting*, ballots cannot be encoded using total orders.

## Beyond Classical Voting Theory

In recent work we have proposed a model where:

- *preferences* and *ballots* can be different structures; and
- a notion of *sincerity* replaces the standard notion of *truthfulness* (because the ballot language may *not allow* you to be truthful).
- Now you can get positive results for certain combinations:
  - Under *approval voting* with standard preferences, you can never benefit from not voting sincerely.
  - If you have *dichotomous preferences*, you can never benefit from not voting sincerely for a wide range of voting procedures.
  - Voting sincerely and effectively is *computationally tractable* in above scenarios.

U. Endriss, M.S. Pini, F. Rossi, and K.B. Venable. *Preference Aggregation over Restricted Ballot Languages: Sincerity and Strategy-Proofness*. Proc. IJCAI-2009.

## Arrow's Impossibility Theorem

It seems reasonable to require a *social welfare function* (SWF), mapping profiles of individual preference orderings to a social preference ordering, to satisfy the following axioms:

- *Unanimity* (UN): if every individual prefers alternative  $x$  over alternative  $y$ , then so should society
- *Independence of Irrelevant Alternatives* (IIA): social preference of  $x$  over  $y$  should only depend on individual pref's over  $x$  and  $y$
- *Non-Dictatorship* (ND): no single individual should be able to impose a social preference ordering

**Theorem 1 (Arrow, 1951)** For *three* or more alternatives, there exists *no SWF* that satisfies all of (UN), (IIA) and (ND).

K.J. Arrow. *Social Choice and Individual Values*. 2nd edition, Wiley, 1963.

## Formal Verification of Arrow's Theorem

Logic has long been used to *formally specify* computer systems, facilitating formal or even *automatic verification* of various properties. Can we apply this methodology also to *social choice* mechanisms?

Tang and Lin (2009) show that the “*base case*” of Arrow's Theorem with 2 agents and 3 alternatives can be fully modelled in *propositional logic*:

- Automated theorem provers can verify  $\text{ARROW}(2, 3)$  to be correct in  $< 1$  second — that's  $(3!)^{3! \times 3!} \approx 10^{28}$  SWFs to check
- Opens up opportunities for quick sanity checks of hypotheses regarding new possibility and impossibility theorems.

Our own work using *first-order logic* tries to go beyond such base cases.

P. Tang and F. Lin. Computer-aided Proofs of Arrow's and other Impossibility Theorems. *Artificial Intelligence*, 173(11):1041–1053, 2009

U. Grandi and U. Endriss. *First-Order Logic Formalisation of Arrow's Theorem*. Proc. LORI-2009.

## Social Choice in Combinatorial Domains

Many social choice problems have a *combinatorial structure*:

- Elect a *committee* of  $k$  members from amongst  $n$  candidates.
- Find a fair *allocation* of  $n$  indivisible goods to agents.

Seemingly small problems generate huge numbers of alternatives:

- Number of 3-member committees from 10 candidates:  $\binom{10}{3} = 120$   
(i.e.,  $120! \approx 6.7 \times 10^{198}$  possible rankings)
- Allocating 10 goods to 5 agents:  $5^{10} = 9765625$  allocations and  $2^{10} = 1024$  bundles for each agent to think about

Conclusion: We need good *languages* for representing preferences!

Y. Chevaleyre, U. Endriss, J. Lang, and N. Maudet. Preference Handling in Combinatorial Domains: From AI to Social Choice. *AI Magazine*, Winter 2008.

## Weighted Goals

A compact representation language for modelling utility functions (cardinal preferences) over products of binary domains —

Notation: finite set of propositional letters  $PS$ ; propositional language  $\mathcal{L}_{PS}$  over  $PS$  to describe requirements, e.g.:

$$p, \quad \neg p, \quad p \wedge q, \quad p \vee q$$

A *goalbase* is a set  $G = \{(\varphi_i, \alpha_i)\}_i$  of pairs, each consisting of a (consistent) propositional formula  $\varphi_i \in \mathcal{L}_{PS}$  and a real number  $\alpha_i$ . The utility function  $u_G$  generated by  $G$  is defined by

$$u_G(M) = \sum \{\alpha_i \mid (\varphi_i, \alpha_i) \in G \text{ and } M \models \varphi_i\}$$

for all models  $M \in 2^{PS}$ .  $G$  is called the *generator* of  $u_G$ .

Different syntactic restrictions give different representation languages.

## Some Results

Examples from our research on weighted goal languages:

- *Expressivity*: If all formulas and weights are positive, then we can express all monotonic utility function, and only those.
- *Succinctness*: Conjunctions of literals can express the same functions as general formulas, but do so strictly less succinctly.
- *Complexity*: Finding the most preferred model is NP-hard in general, but in  $O(n \log n)$  if all formulas are literals.
- Applications: *combinatorial auctions* and *expressive voting*

J. Uckelman. *More than the Sum of its Parts: Compact Preference Representation over Combinatorial Domains*. PhD thesis, ILLC, University of Amsterdam, 2009.

J. Uckelman, Y. Chevaleyre, U. Endriss, and J. Lang. Representing Utility Functions via Weighted Goals. *Mathematical Logic Quarterly*, 55(4):341–361, 2009.

## Judgment Aggregation

Preferences are not the only structures that we may wish to aggregate. JA studies the aggregation of judgments on related propositions.

	$p$	$p \rightarrow q$	$q$
Agent 1:	Yes	Yes	Yes
Agent 2:	No	Yes	No
Agent 3:	Yes	No	No
Majority:	Yes	Yes	No

While each individual set of judgments is logically consistent, the collective judgment produced by the majority rule is not.

Research issues: impossibility theorems; characterisation of admissible agendas; proposals for “good” aggregation procedures; ...

C. List and C. Puppe. Judgment Aggregation: A Survey. *Handbook of Rational and Social Choice*. Oxford University Press, 2009.

## Complexity of Judgment Aggregation

What about computational considerations in JA?

In recent work we address the following questions:

- *Safety of the Agenda*: Given an agenda  $\Phi$  (set of propositions), can we guarantee that any aggregation procedure belonging to a given class of procedures (characterised via some axioms) will never “produce a paradox”?
- What is the computational complexity of deciding SoA? (turns out to be  $\Pi_2^p$ -complete for all interesting axioms)

U. Endriss, U. Grandi, and D. Porello. *Complexity of Judgment Aggregation*. Working Paper, ILLC, University of Amsterdam, 2009.

## Other Topics

Social choice theory is not just about voting and preferences. Aggregation also plays a role in other domains, e.g.:

- Multiagent Resource Allocation and Fair Division
- Mechanism Design
- Stable Matchings

## Computational Social Choice

We have seen several examples for work in COMSOC.

Research can be broadly classified along two dimensions —

The kind of *social choice problem* studied, e.g.:

- aggregating individual judgements into a collective verdict
- electing a winner given individual preferences over candidates
- fairly dividing a cake given individual tastes

The kind *computational technique* employed, e.g.:

- algorithm design to implement complex mechanisms
- complexity theory to understand limitations
- logical modelling to fully formalise intuitions
- knowledge representation techniques to compactly model problems
- deployment in a multiagent system

## Last Slide

- COMSOC is an exciting area of research bringing together ideas from mathematical economics (particularly social choice theory) and computer science (including logic, AI, MAS, TCS).
- COMSOC website: <http://www.illc.uva.nl/~ulle/COMSOC/>
- You are welcome to attend the COMSOC Seminar at the ILLC (typically, every 2-3 Fridays at 4pm)
- There's a Workshop on Preferences at the ILLC *this* Friday at 2pm (speakers: Jérôme Lang, Francesca Rossi, Mike Wooldridge et al.)
- Papers are on my website (including the surveys below).

Y. Chevaleyre, U. Endriss, J. Lang, and N. Maudet. *A Short Introduction to Computational Social Choice*. Proc. SOFSEM-2007.

Y. Chevaleyre, U. Endriss, J. Lang, and N. Maudet. Preference Handling in Combinatorial Domains: From AI to Social Choice. *AI Magazine*, Winter 2008.