



Collective Intelligence

A Multiagent Systems Perspective

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Collective Intelligence

- In theory: cooperation, competition, and coordination produce some **emergent** super-individual capabilities that transcend those of the individual
- In practice (currently): crowdsourcing, folksonomies, social computation, human-based computation
- **Emergence:** “you throw the pieces of a watch on a table and suddenly, there’s a watch!”
- Tempting: “emergent macro-scale phenomena will do our work for us”. But how do you **design** for intelligence to emerge? (And what do you do when it does?)

Agents

- Ambiguous, overused, overhyped term - let's just assume it means something like "AI with an integration focus"
- Single-agent systems: integration of capabilities in one embodied, situated entity operating on behalf of a user
- Multiagent systems: integration of several agents (esp with different goals) in the same environment
- The "different goals" bit is the **genuinely** new element

Data

- Petascale data processing is the name of the game today, massive application benefits
- Data is not “intelligent” in itself, it is human interpretation that makes it useful
- Analysing data relies on human expert, this part often not explicit in systems design
- Effectively, the intelligent agents in the “data-driven” world of computing are still human experts

Design

- Human (and machine) intelligence as reflected in data says nothing about what to do with it
- Describing purposeful computation and aiding it with automated support is missing
- Top-down perspective of specification and reasoning vs. bottom-up perspective of data-driven computation
- Agents provide right technology for this, but largely disconnected from “data world”

Incentives

- The missing link between data and behaviour toward it
– the key to design for collective intelligence
- Economics based models currently offer the only rigorous framework (game theory etc)
- These models don't account for many types of human behaviour (esp. in constantly changing Web world)
- Agents: modelling systems with potential conflict of interest, simulating human behaviour

Semantics

- **Meaning = Cognition + Incentive**
 - Human intelligence of individuals interprets information with a purpose, and produces knowledge
- Evolution of collective intelligence mediated by configuration of different participants' incentives and interpretations
- Emergent collective knowledge reflected in evolution of semantics of symbols used
- “Evolving Symbol Systems Hypothesis”: Incentive-aware rules for manipulating semantics sufficient to achieve collective intelligence

Challenges

- Multi-level modelling and evolution of representations, linking “deep” to “shallow” reasoning
- Reasoning about incentives and behaviour, but engaging with humans to learn about these
- Lack of automation support for users and application designers: tapping into new data is useful, but not enough

(My own) Examples

- Computing solutions to interaction problems from high-level problem specifications
- Mining opponent and context models from executions of agent interaction protocols
- Collaborative and competitive methods for distributed machine learning
- Developing executable declarative languages for implementing social reasoning in agents