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Uncertainty in Coalition Formation

"Sequential Decision Making in Repeated Coalition Formation under Uncertainty" by Chalkiadakis and Boutilier "Bayesian Coalitional Games" by leong and Shoham

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> > May 17, 2010

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Introduction

- Given a coalition, how do we distribute the payoff according to some stability concepts ?
- Simplifying assumption: the generated value of the coalition and the potential of each member are common knowledge
- Issue: this is rarely the case in real-life scenarios
- How do we deal with uncertainty regarding other agents' capabilities ?
 - A reinforcement learning model, where agents update information about the others through repeated interaction
 - Generalize classical Coalitional Games Theory to settings with uncertainties

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Markov Decision Processes (MDP)

- ► The environment can be modelled as an MDP (S,A,R,P):
 - S the set of states
 - A the set of actions
 - R a reward function (R: S x A $\rightarrow \mathbb{R}$)
 - *P* a stochastic transition model (P: $S \times A \rightarrow S$)
- After each action, the agent "interprets" the response from the environment and changes accordingly his beliefs and his next possible moves.
- ▶ We define the **discounted future reward** as $\sum_{t=0}^{\infty} \gamma^t \cdot R(s, a)$ where $\gamma \in [0, 1]$ is a discount factor
- ► Our aim is to find a **policy** π: S → A, such that the discounted future reward is maximised
- A PO-MDP is a generalization of an MDP: the underlying state cannot be directly observed, so the agent maintains a probability distribution over the set of possible states

Exploration vs exploitation

- Each agent has two complementary objectives:
 - form efficient, profitable coalitions
 - gather as much reward as possible
- This means that agents should not seek to reduce uncertainty for *its own sake*, by employing crude exploration policies
- A better strategy is to explore only promising partners, while for the others some type uncertainty will still remain in the end
- It can be shown that Bayesian exploration finds an optimal balance between exploration and exploitation

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Bayesian Reinforcement Learning Framework

Bayesian Coalition Formation Model

- Given
 - N the set of agents
 - ► *T_i* a finite set of possible types for agent i
 - ► B_i(t_{-i}) the beliefs of agent i over the types of the other agents in the coalition
 - A_C the set of coalitional actions
 - O the set of outcomes
- The value of the coalition can be expressed as:
 - $V(C|t_{C}) = \max_{\alpha_{C} \in A_{C}} \sum_{o} Pr(o|\alpha, t_{C}) \cdot R(o) = \max_{\alpha_{C} \in A_{C}} Q(C, \alpha|t_{C})$
- However, each agent is uncertain about the types of its partners:

$$V_i(C|t_C) = \max_{\alpha_C \in A_C} \sum_{t_C \text{ in } T_C} B_i(t_C) \cdot Q_i(C, \alpha|t_C) = \max_{\alpha_C \in A_C} Q_i(C, \alpha)$$

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Example

- Assume we have the following game:
 - ➤ 3 major types: interface designer (ID), programmer (P), and systems engineer (SE)
 - ▶ 3 quality types: *bad*, *average*, and *expert*
 - 3 actions: bid for large/medium/small project
 - 3 outcomes: make large/medium/small profit
- The more members a coalition has, the more likely it is to be successful in getting higher profits if it tries to bid for large projects.
- However, it is not only about the size of the coalition, but also about how competent the team members are
- Given 5 agents:
 - a1 = bad P, a2 = average P, a3 = expert P, a4 = bad ID, and a5 = bad SE
- Then, the best coalition structure is:
 - ((a1), (a2, a3), (a4),(a5))

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Optimal repeated coalition formation

Bellman-like equations:

►
$$Q_i(C, \alpha, d_C, B_i) = \sum_o Pr(o|C, \alpha, B_i) \cdot [r_iR(o) + \gamma V_i(B_i^{o,\alpha})] = \sum_{t_c} B_i(t_c) \sum_o Pr(o|\alpha, t_c) \cdot [r_iR(o) + \gamma V_i(B_i^{o,\alpha})]$$

► $V_i(B_i) = \sum_{C|i \in C, d_c} Pr(C, \alpha, d_C|B_i) \cdot Q_i(C, \alpha, d_C, B_i)$

▶ where $r_i = \frac{d_i}{\sum_{j \in C} d_j}$ is the relative demand of agent *i*

After an action is taken, the agent observes the resulting state and updates its beliefs concerning its partners' types:

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Optimal repeated coalition formation (2)

- ► In typical Bellman equations: $V_i(s') = Q_i(s', a*)$, where $a* = argmax_a \sum_{i=1}^n Q_i(s', a)$
- However, this idea is not applicable in our case because the choices that dictate reward, namely, the coalition that is formed, are not in complete control of agent *i*
- ► Instead, *i* must predict, based on his beliefs, probability Pr(C, α, d_C|B_i) with which a specific coalitional agreement (C, α, d_C) will arise as a result of the negotiation

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Challenges

- The state space and action space grow exponentially with the number of agents
- An exact solution to the repeated coalition formation is generally infeasible
 - Computational approximations are needed

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Computational bottlenecks

- Computing $Pr(C, \alpha, d_C|B_i)$
 - Agents can only observe the outcome of their own coalition action only, and thus it is not possible for them to monitor how the beliefs of others are changing
- Solving the optimal exploration PO-MDP
 - The number of future states that need to be considered is too large



One-step Lookahead Algorithm (OSLA)

- Deals with only immediate successor belief states
- Approximates Pr(C, α, d_C|B_i) by viewing it as the probability of reaching an agreement after one negotiation step, rather than after a whole negotiation process
- Advantages:
 - Allows more flexibility in investigating the space of coalition structures, without forcing the agents to reach a stable structure at each stage before acting
 - Applies best when "real-time" performance is required
 - It can be shown that one-step methods converge to the Bayesian Core of the game (if that exists)

VPI Exploration method

- The main idea is to consider the gain achieved by learning the true value of some coalitional agreement σ = (C, α, d_C)
- Assume that by adopting σ we obtain exact evidence regarding the types of the agents in C (t^{*}_C)
- Based on t_C^* we define the true value of σ as:

•
$$q_{\sigma}^* = Q_i(C, \alpha, d_C | t_C^*) = r_i \cdot \sum_o Pr(o | t_C^*) \cdot R(o)$$



VPI Exploration method (2)

- Let σ₁ be the current best coalitional agreement, and σ₂ the second-best option
- Then, the gain achieved by learning the true value of σ is:

$$gain_{\sigma}(q^*_{\sigma}, t^*_{\mathcal{C}}) = \left\{ egin{array}{ll} q^*_{\sigma} - q_{\sigma_1} & ext{ if } \sigma
eq \sigma_1 ext{ and } q^*_{\sigma} > q_{\sigma_1} \ q_{\sigma_2} - q^*_{\sigma} & ext{ if } \sigma = \sigma_1 ext{ and } q^*_{\sigma} < q_{\sigma_2} \ 0 & ext{ otherwise} \end{array}
ight.$$

which is also known as the Value of Perfect Information

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VPI Exploration Method						

EVPI

- \blacktriangleright However, the agent does not know in advance what types will be revealed for σ
- What it actually considers is the Expected Value of Perfect Information:

•
$$EVPI(\sigma|B_i) = \sum_{t_c^*} gain(q_\sigma^*|t_c^*) \cdot B_i(t_c^*)$$

- EVPI gives the value of exploring σ, therefore agent should have a preference for maximizing:
 - $QV_i(\sigma|B_i) = Q_i(\sigma|B_i) + EVPI(\sigma|B_i)$



OSLA vs VPI

- Both compute the value of the agreements in a myopic manner
- VPI exploits the value of perfect information regarding the types, which is in contrast to OSLA which estimates the value of specific coalitional actions
 - VPI does not have to explicitly incorporate the prior hypothesis
 (B_i) in the calculation of the Q-values
 - VPI does not need to account for the probability of agreement when transitioning to future belief states
- None of them is tied to specific coalition formation processes, therefore can be used with any bargaining processes

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- Extension of Coalitional Game Theory (CGT) to include uncertainty about other agents
 - Payoff vector vs contract
 - Same solution concepts
 - Lots of notions needed to solve a BCG
- Meant to be a combination of Bayesian Theory and CGT to bring real life and CGT closer

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 $(N, \Omega, \mathbb{P}, (\mathcal{I}_j), (\succeq_j))$ with

• $N = \{1, 2, ..., n\}$ set of agents participating in the game

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- \succeq_j agent j's preference over distributions of payoffs

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BCG definition interpretation

 $(N, \Omega, \mathbb{P}, (\mathcal{I}_j), (\succeq_j))$ with

Ω represents uncertainty: multiple possible worlds for possible agent types

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- Ω represents uncertainty: multiple possible worlds for possible agent types
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- ► *I_j* is similar to Coalition Structure: partition over possible worlds

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- ► ≿_j represents preference over distributions of payoffs, (more or less) i.e contracts
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▶ A contract maps possible worlds to payoff vectors: $\mathbf{c}^S : \Omega \mapsto \mathbb{R}^S$

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Coalition	ω_1	ω_2
{1, 2}	(2.5, 3.0)	(6.0, 2.5)
{3}	0.5	7.0

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- Captures the uncertainty in numbers
- Let us focus on grand contracts, c^N

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- Imagine a contract with high payoff in lots of worlds except one, and that world has a high prior probability



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- Possibly different solution for all three timings



► Three timings, three forms of blocking, three forms of the core

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- Three timings, three forms of blocking, three forms of the core
- "Given a BCG (N, Ω, ℙ, (I_j), (≥_j)) and a grand contract c^N, a coalition S ex-ante blocks c^N if there exists an S-contract c^S such that c^S_S ≻_S c^N_S."



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- Ex-interim blocking is more difficult, explained later



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- Ex-interim blocking is more difficult, explained later
- "A grand contract c^N is in the (ex-ante, ex-interim, ex-post) core of a BCG if no coalition S ⊆ N (ex-ante, ex-interim, ex-post) blocks c^N."



Why is ex-interim blocking more difficult? Example:

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- Why is ex-interim blocking more difficult? Example:
 - Two agents $\{1, 2\}$ and two worlds $\{\omega_1, \omega_2\}$, true world ω_1



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 - Information partitions $\mathcal{I}_1 = \{\{\omega_1, \omega_2\}\}\$ and $\mathcal{I}_2 = \{\{\omega_1\}, \{\omega_2\}\}\$, and grand contract c^N



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- Information partitions $\mathcal{I}_1 = \{\{\omega_1, \omega_2\}\}$ and $\mathcal{I}_2 = \{\{\omega_1\}, \{\omega_2\}\}$, and grand contract c^N
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 - Agent 2 knows that ω_1 is the true world! And will prefer c^S
 - Agent 1 sees him preferring that, and can deduct that ω₁ is the true world too
- ► Agents can *observe* and *learn* from other agents

- Why is ex-interim blocking more difficult? Example:
 - Two agents {1, 2} and two worlds $\{\omega_1, \omega_2\}$, true world ω_1
 - Information partitions $\mathcal{I}_1 = \{\{\omega_1, \omega_2\}\}\$ and $\mathcal{I}_2 = \{\{\omega_1\}, \{\omega_2\}\}\$, and grand contract c^N
 - ► Preferences $c_2^S | \mathcal{I}_2(\omega_1) \succ_2 c_2^N | \mathcal{I}_2(\omega_1)$ and $c_2^N | \mathcal{I}_2(\omega_2) \succeq_2 c_2^S | \mathcal{I}_2(\omega_2)$
 - Agent 2 knows that ω_1 is the true world! And will prefer c^S
 - Agent 1 sees him preferring that, and can deduct that ω₁ is the true world too
- Agents can observe and learn from other agents
- ► Hard to formalize, so we won't do that :)



 iterated elimination of dominated information sets: eliminate worlds that are not 'attractive': worlds that are not preferred by any agents



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- Result: set of worlds Ω*(c^N, c^S) that is attractive to all agents engaged in contract c^S



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- "Agents only consider contracts for worlds that other agents, given their information partitions, would find attractive"

Outline

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RCF vs BCG

Conclusion

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Comparison between Repeated Coalition Forming and Bayesian Coalitional Games

Motivation exactly the same

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 - ► Final outcome: Best/fairest coalition!

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- Two possible solutions: Repeated Coalition Forming and Bayesian Coalitional Games
- Completely different models and ways of finding the best coalition
- Many other possible ways of dealing with uncertainty in coalition formation

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Thank you!!

Questions?

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