Graphical Games

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Graphical Models for Game Theory

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Multiplayer Games

Players i = 1...n

each with a finite set of **pure** strategies A_i

for simplicity we will assume that the action chosen by player $i a_i \in \{0,1\}$, (binary actions)

The payoffs to player *i* are given by a matrix M_i

This matrix is indexed by a joint action $\vec{a} \in \{0,1\}^n$

The value $M_i(\vec{a})$ (wlog $\in [0, 1]$) is the payoff for player *i* if players play the joint action \vec{a}

Normal Form (mixed strategies)

In our binary setting a mixed strategy for player *i* is given by the probability $p_i \in [0, 1]$ that the player will play 0

The expected payoff to player *i* from the joint mixed strategy \vec{p} is then defined as

$$M_i(\vec{p}) = E_{\vec{a} \sim \vec{p}}[M_i(\vec{a})]$$

Nash Equilibrium

 $\vec{p}[i:\bar{p}_i]$ denotes the joint mixed strategy which is the same as \vec{p} except that player i deviates to \bar{p}_i

Then \vec{p} is a NE for the game iff $\forall i, \bar{p}_i \in [0, 1]$

 $M_i(\vec{p}) \geq M_i(\vec{p}[i:\bar{p}_i])$

ϵ -Nash Equilibrium

 $\vec{p}[i:\bar{p}_i]$ denotes the joint mixed strategy which is the same as \vec{p} except that player i deviates to \bar{p}_i

Then \vec{p} is a ϵ -NE for the game iff $\forall i, \bar{p}_i \in [0, 1]$

 $M_i(\vec{p}) + \epsilon \geq M_i(\vec{p}[i:\bar{p}_i])$

Issues with Normal Form

Assuming *n* players and 2 actions, as we have here, leads to the need for : *n* matrices M_i (one for each player) each of size 2^n

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Furthermore tabular form fails to capture structure inherently present in the game

Structure in Games

- It is assumed that the payoff M_i for player *i* is a function of all the components a_j , (j = 1...n) in the joint action vector \vec{a}
- However the payoff for player i may be dependent only on the actions of a subset of players N(i)
- \longrightarrow conditional independence payoff assumption

Cond. Ind. Payoff



Graphical Games

An undirected graph G

- n vertices one for each player i
- ► N(i) is the neighborhood of player i \longrightarrow there is an edge $(i, j), \forall j \in N(i)$

Graphical Games Payoff Representation

Local payoff matrix for \overline{M}_i depends only on the actions taken by players in N(i)

 $M_i(\vec{a}) = \bar{M}_i(\vec{a}[N(i)])$

Representation Complexity

|N(i)| is the degree of local interaction for node *i*

The maximum k over the graph $k = \max_i |N(i)|$ defines the complexity of the representation $O(n2^k)$



Why this is cool

- Computational Specific topological properties can be used to yield effficient algorithms for finding Nash equilibria
- Structural
- Interdisciplinary

Why this is cool

- Computational
- Structural Provide a tool for examining whether the topology of G implies structural properties of the equilibria
- Interdisciplinary

Why this is cool

- Computational
- Structural
- Interdisciplinary Allow the use of powerful methods from different fields (e.g. machine learning , statistics)

Tree Graphical Games



Tree Games



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TreeNash Algorithm

Two pass algorithm.

- Downstream
- ► Upstream

TreeNash Algorithm

Two pass algorithm.

- Downstream Calculates cond. equilibria and passes "witness" lists down the tree
- Upstream

TreeNash Algorithm

Two pass algorithm.

- Downstream
- Upstream Selects "witness" lists going from the root to the leaves and calculates a NE

Each parent sends a table $T(V, U_i)$ such that $T(v, u_i) = 1$ iff there exists a NE in $(G^U, M_{V=v}^U)$ for which $U_i = u_i$



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If U_i is a leaf then $T(v, u_i) = 1$ iff u_i is a best response to v



T(w, v) = 1 iff v is a best response to w and $T(v, u_i) = 1, \forall i$



T(w, v) = 1 iff v is a best response to w and $T(v, u_i) = 1, \forall i \ \vec{u}$ is then added to the "witness" list of T(w, v)



At the root, the algorithm computes the table T(W) where T(w) = 1 iff w is a best response to \vec{v} and $T(w, v_i) = 1, \forall i$



TreeNash Upstream

The algorithm chooses a value w for which T(w) = 1, then passes this value plus the witness v to its parent (instructing it to "play" v)



TreeNash Upstream

V receives w, v and sends the witness \vec{u} of T(w, v) = 1 to its parents



TreeNash (A slight issue)

The actions (u, v, w...) are continuous variables \rightarrow Can T(w, v) be represented compactly?

Approximate TreeNash

Discretization of the action space

Player *i* can now only play action $q_i \in \{0, \tau, 2\tau, ..., 1\}$

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Theorem Approximate TreeNash computes a ϵ -NE for the game (G, M) in time polynomial in the representation of (G, M)

Exact TreeNash ?

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Yes! However its complexity is exponential in the number of vertices of G

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Computing an exact equilibrium in time polynomial in the size of the tree remains an open issue

The End