

Large games and population protocols

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- ▶ Much of the work on games I talk about here is joint with Soumya Paul, currently at Univ. Luxembourg.
- ▶ I do not know much about population protocols but am hoping to learn.

Summary

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 - ▶ Players are anonymous, interaction is simple. Pure strategy Nash equilibria exist for a large class of games.
- ▶ Population protocols:
 - ▶ Systems with a large number of identical finite state automata.
 - ▶ Interaction is simple, outcome based on states of interacting automata.
 - ▶ Compute exactly the **semi-linear** predicates.
- ▶ Are there interesting connections between the two ? I do not know, but suspect so.

Joining clubs

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- ▶ *The Santa Fe bar problem*: The payoff depends on how many others act as I do.
- ▶ Network congestion problems.

Expectations of a population

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- ▶ What would be a **logical** basis for expecting others to act in a particular way ?
- ▶ This is hard for a one-shot game, but in repeated play, or in games of long duration, rationale based on observation can significantly affect game dynamics.

Main issues

Framework: outcomes determined by choice distributions.

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- ▶ This can lead to interesting stability issues.
- ▶ In turn, this can affect players' strategizing.

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- ▶ The game has no pure strategy Nash equilibrium, but a mixed strategy Nash equilibrium (NE).
- ▶ One property of this NE is that it is not *information proof*: once you are informed of the other player's move, you have an incentive to switch (from the mixed strategy to the pure mismatch strategy).

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n -player mismatch game.

- ▶ Simultaneously, each of n players of type A and n players of type B have to choose between u or d .
- ▶ The payoff to every player of type A equals the proportion of players of type B that her choice matches.
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- ▶ When $n = 2$, this is the earlier game. So clearly, it inherits some of the trouble.

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- ▶ Suppose that every player, either type, chooses one of the two randomly with equal probability.
- ▶ Then within each group the proportions of the two selected choices are likely to be close to one half, and no player would be able to gain much by switching.
- ▶ There is a high probability for the events of no possible improvement greater than some given epsilon holding *simultaneously* for all players.

Majority mismatch game

n -player mismatch game with a twist.

- ▶ The payoff to every player of type A is 1 if her choice matches that of at least one half of the choices of type B , and 0 otherwise.
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- ▶ If n is odd, no matter what strategies are played, at every known outcome at least one half of the players will have a strong incentive to unilaterally revise their choices.
- ▶ There is no information proof equilibrium because of the discontinuity in the payoff function.

Many questions

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- ▶ What are good models for large games ?

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- ▶ What are good models for large games ?
- ▶ What are the implications for social algorithms ?

Any good news ?

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In some respects, large games are easier to reason about than small ones.

- ▶ Behaviour for large n can smooth out many individual irregularities.
- ▶ Many problems related to mutual intersubjectivity and surprise moves disappear.
- ▶ When the number of players is large but the number of player **types** is small, we can sometimes reduce the analysis to small games.

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- ▶ A function $f_i : \mathbf{Y} \rightarrow \mathbb{Q}$ for every player i .

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- ▶ Player i gets a limit average payoff:

$$p_i(\rho) = \lim_{m \rightarrow \infty} \inf \frac{1}{m} \sum_{j=1}^m f_i(y_j).$$

Main questions

We study player types specified by formulas that code up beliefs of players about others. We will discuss the logic and its formulas later.

- ▶ **Main question:** Given an initial type distribution of players, which types are eventually stable ?

Type based analysis

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- ▶ Can we use the structure of these transducers to do this reduction ?

Products of transducers

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- ▶ We show in the case of **deterministic transducers**, that the product of a type with itself is isomorphic to the type.
- ▶ Thus a population of 1000 players with only two types needs to be represented only by pairs of states and not 1000-tuples.
- ▶ However, there is no free lunch: an exponential price has to be paid for determinization. But we characterize when this can be worthwhile.

Transducer reduction

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- ▶ **Theorem:** Suppose that we have n players, k choices and t types. Let $p = \max_i |R_i|$, where the i th type formula induces a nondeterministic FST with state space R_i . Then the type based analysis is more efficient when

$$\frac{n}{t} > 0.693 \cdot k \cdot \pi(p)$$

where $\pi(p)$ is the number of primes less than or equal to p .

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- ▶ A map $F : A_{p,q} \rightarrow R^d$ is said to be direction-preserving if for any $r_1, r_2 \in A_{p,q}$ with $|r_1 - r_2|_\infty \leq 1$, we have, for all $i, 1 \leq i \leq d$: $(F_i(r_1) - r_1^i)(F_i(r_2) - r_2^i) \geq 0$.

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- ▶ Note that the fixed point computation happens in a discrete space (where we do not have Brouwer - Kakutani fixed point theorems). So we use a different technique due to Chen.

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- ▶ Let $\sigma \in \Sigma_i$. We say σ is a best response to a set T of player types, if for every profile π such that $\mathcal{T}(\pi^{-i}) = T$, $u_i(\sigma; \pi^{-i}) \geq u_i(\pi)$.

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- ▶ A profile π is in local equilibrium if for all i , $\pi(i)$ is the best response to $\mathcal{T}(\pi^{-i})$.

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- ▶ Stability in this notion is sensitive to the way projections of strategies to types is defined.
- ▶ The projection function is uniform in the definition above. In general, it would be indexed by players, or better, by types again !
- ▶ We can show that local equilibrium is a new notion, in the sense that we can define games that have local but no global equilibria, or the other way.

Two kinds of stability

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Local equilibria predict stable play in the dynamics of strategy improvement. But this assumes visibility to be static.

- ▶ In large games, visibility is dynamic as well.
- ▶ This results in a dynamic game form.
- ▶ Note that the two dynamics are recursive in each other.
- ▶ We describe the game form dynamics by **neighbourhoods**.

Last words on large games

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- ▶ Game theorists have mainly studied utility functions and learning, interaction / communication models are very simplistic.
- ▶ Our results:
 - ▶ Soumya Paul and R. Ramanujam, “Dynamics of choice restriction in large games”, *Journal of Game Theory Review*, vol 15, no. 4, 156-184, 2013.
 - ▶ Soumya Paul and R. Ramanujam, “Subgames within large games and the heuristic of imitation”, *Studia Logica*, vol 102, no. 2, 361-388, 2014.

Population protocols

Population protocols were introduced by Angluin, D., Aspnes, J., Diamadi, Z., Fischer, M.J., Peralta, R.: Computation in networks of passively mobile finite-state sensors. Distributed Computing 18(4), 2006.

The defining features of the basic model are:

- ▶ Finite-state agents and uniformity.
- ▶ Computation by direct interaction, and unpredictable interaction patterns.
- ▶ Distributed inputs and outputs.
- ▶ Convergence rather than termination.

The basic model

An n -agent PP is a tuple $P = (Q, \delta, \iota, \omega)$ over (Σ, Γ) where Σ is the input alphabet, Γ is the output alphabet, $\iota : \Sigma \rightarrow Q$, $\omega : Q \rightarrow \Gamma$ and $\delta \subseteq Q^4$.

- ▶ The initial configuration is determined by the inputs via ι .
- ▶ δ describes pairwise interaction and thus configuration change.
- ▶ Via ω all automata constantly produce output.
- ▶ Fairness assumption: if C appears infinitely often in a computation and $C \rightarrow C'$ then C' appears infinitely often in it.
- ▶ A protocol computes a function f that maps multisets of elements of Σ to Γ if, for every such multiset I and every fair execution that starts from the initial configuration corresponding to I , the output value of every agent eventually stabilizes to $f(I)$.

The OR protocol

The aim of the protocol is to output the 'or' of all input bits.

- ▶ $\Sigma = \Gamma = Q = \{0, 1\}$ and the input and output maps are the identity functions.
- ▶ The only interaction in δ is $(0, 1) \rightarrow (1, 1)$.
- ▶ If all agents have input 0, no agent will ever be in state 1.
- ▶ If some agent has input 1 the number of agents with state 1 cannot decrease and fairness ensures that it will eventually increase to n .

The dancers protocol

The agents are dancers, and each dancer is (exclusively) a leader or a follower. The problem is to determine whether there are more leaders than followers.

- ▶ $\Gamma = \{0, 1\}$. We set $\Sigma = \{L, F\}$ and $Q = \{L, F, 0, 1\}$.
- ▶ The input map is the identity; the output maps L and 1 to 1 , F and 0 to 0 .
- ▶ δ has: $(L, F) \rightarrow (0, 0)$, $(L, 0) \rightarrow (L, 1)$, $(F, 1) \rightarrow (F, 0)$ and $(0, 1) \rightarrow (0, 0)$.
- ▶ In case of a tie, the last rule ensures that the output stabilizes to 0 .

Convergence

It is not obvious that this protocol converges.

- ▶ Consider the sequence of configurations:

$$(L, L, F), (0, L, 0), (1, L, 0), (0, L, 0), (0, L, 1), (0, L, 0)$$

- ▶ Repeating the last four transitions yields a non-converging execution, but it is not **fair**.

Some exercises

The notion of fairness is subtle: it is distinct from the condition that each pair of agents must interact infinitely often. E.g. consider $(L, L, L)^\omega$ where all interactions take place between the first two agents: it is fair.

- ▶ Show the dancers protocol converges in every fair execution.
- ▶ Design a protocol to determine whether more than $2/3^{\text{rds}}$ of the dancers are leaders.
- ▶ Design a protocol to determine whether more than $2/3^{\text{rds}}$ of the dancers play the same role.

Modular arithmetic

Suppose each agent is given an input from $\Sigma = \{0, 1, 2, 3\}$. Consider the problem of computing the sum of the inputs, modulo 4.

- ▶ The protocol gathers the sum (modulo 4) into a single agent. Once an agent has given its value to another agent, its value becomes null, and it obtains its output value from the eventually unique agent with a non-null value.

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- ▶ $Q = \{0, 1, 2, 3, n_0, n_1, n_2, n_3\}$, where n_v stands for null value with output v .
- ▶ δ has $(v_1, v_2) \rightarrow (v_1 + v_2, n_{v_1+v_2})$ (addition modulo 4) and $(v_1, n_{v_2}) \rightarrow (v_1, n_{v_1})$.

Computability

We can represent multisets over Σ by vectors. For instance (a, b, a, b, b) over $\Sigma = \{a, b, c\}$ by $(2, 3, 0)$. Thus we can speak of input vectors (x_1, \dots, x_d) in \mathcal{N}^d where $d = |\Sigma|$.

- ▶ **Threshold** predicates are of the form $\sum_{i=1}^d c_i x_i < a$, and **remainder** predicates are: $\sum_{i=1}^d c_i x_i = a \pmod{b}$.
- ▶ Angluin et al (easily) show that population protocols can compute these and their boolean combinations.
- ▶ Surprisingly, the converse also holds: these are the **only** predicates that a population protocol can compute.

The theorem

Theorem (Angluin et al): A predicate is computable in the basic population protocol model if and only if it is **semilinear**.

- ▶ The proof is quite involved, the main tool is Higman's Lemma. There are three main steps to the proof.

The steps

The three main steps:

- ▶ Show that any predicate stably computed by a population protocol is a finite union of monoids: sets of the form $\{(b + k_1 a_1 + k_2 a_2 + \dots) \mid k_i \in \mathcal{N} \text{ for all } i\}$, where the number of terms may be infinite.

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- ▶ Show that when detecting if a configuration x is output-stable, it suffices to consider its truncated version:

$$\tau_k(x_1, \dots, x_d) = (\min(x_1, k), \dots, \min(x_d, k))$$

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- ▶ Finally we can reduce the problem to a form of coverability.

Variations on the model

Many variants of the basic model have been studied in the literature.

- ▶ One-way interaction: we have sender and receiver agents. This leads to immediate and delayed observation models, and queued transmission models.
- ▶ Delayed observation models can detect multiplicity of input symbols (upto a threshold) and essentially only such predicates.
- ▶ Immediate observation models can count the number of agents with a particular input symbol (upto a threshold).
- ▶ Queued models have the same power as the basic model.
- ▶ Interaction graphs, in general, lead to Turing computability.
- ▶ Many papers study **random** interaction models.

Games

We consider two player games in normal form.

The two players are I (initiator) and R (responder).

Let $S(I), S(R)$ denote the (finite) sets of strategies of the players.

$BR_I : S(R) \rightarrow S(I)$ is the **best response** map for player I ; BR_R similarly.

Assume that $S(I) = S(R) = S$, for now, and let Δ be a fixed integer constant.

From games to protocols

To each such game we can associate a population protocol as follows:

- ▶ $Q = S$.
- ▶ $(q_1, q_2, q'_1, q'_2) \in \delta$ iff:
 - ▶ $q'_1 = q_1$ if $u_I(q_1, q_2) \geq \Delta$; $q'_1 = x \in BR_I(q_2)$, otherwise.
 - ▶ $q'_2 = q_2$ if $u_R(q_1, q_2) \geq \Delta$; $q'_2 = x \in BR_R(q_1)$, otherwise.
- ▶ We vary input and output functions and Δ to get a class of protocols associated with the game.

Pavlovian protocol

Call a population protocol **Pavlovian** if it can be obtained from a game by the rules above.

- ▶ **Proposition:** The class of predicates computable by Pavlovian population protocols is closed under negation.
- ▶ The proof proceeds by a kind of determinization: by constructing 'equivalent' games that have unique best response, which makes the rules above deterministic.
- ▶ It is not clear that predicates computable by Pavlovian population protocols are closed under conjunction or disjunction.

Product protocols

Consider k two-player games, and define (in the natural way) the associated protocol by the k -fold product of the rules above. We call them **Multi-Pavlovian** protocols.

- ▶ **Theorem:** The predicates defined by Multi-Pavlovian protocols are exactly the semi-linear ones. Thus every population protocol corresponds to a finite product of 2-player normal form games.
- ▶ There are surprises when we restrict ourselves to symmetric games. (A population protocol is symmetric if whenever $(q_1, q_2, q_3, q_4) \in \delta$ then $(q_2, q_1, q_4, q_3) \in \delta$ as well.)

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- ▶ Evolutionary game theory is a well-studied subject approach to population dynamics, modelling simple interactions on a large scale.
- ▶ Population protocols provide a very interesting model of computation that is very similar.
- ▶ This may be one way to **scale up** systems of automata and their interactions the study of which has been too rigid.
- ▶ The model of games and automata has been used well in the context of systems with a fixed number of players / components. Moving to large distributed systems, the models of large games and population protocols seem promising and worthy of study.

Discussion time

Thank you.

Questions, comments, suggestions welcome; also, please write to jam@imsc.res.in.