Translating Extensive Form Games to Open Games with Agency ACT2021

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Introduction

Game theory is the mathematical study of interaction among independent, self-interested agents.

- Essentials of Game Theory [LS08]

Open games as a framework for compositional game theory (string diagram) Classical game theory: normal form and extensive form (PD in both forms)

Introduction

In this work, we give a canonical translation from extensive form games to open games To do so, we introduce:

- 1. Open games with agency, an improved compositional framework for games,
- 2. An operator calculus for games, in particular new choice operators,
- 3. Inductive data types for extensive form games of (im)perfect information.

Game theory is the mathematical study of interaction among independent, self-interested agents.

- Essentials of Game Theory [LS08]

Examples: economic games, proof theory, machine learning, control theory, etc.

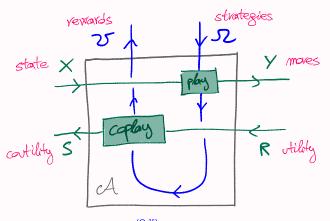
A game factors in two parts:

- 1. An arena, which models the interaction patterns in the game
- 2. Players, which intervene in the arena by making decisions at different points





An arena is an open system with three boundaries:



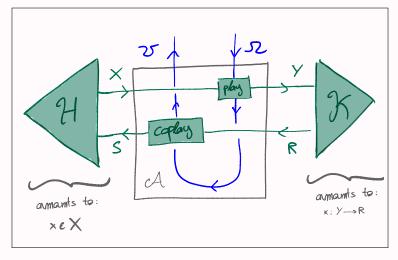
This is a parametrised lens $\mathcal{A}:(X,S)\stackrel{(\Omega,\mho)}{\longrightarrow}(Y,R)$ specified by two maps

$$\mathsf{play}: \mathbf{\Omega} \times X \to Y,$$

$$\mathsf{coplay}: \Omega{\times}X \times R \to \mho \times S$$

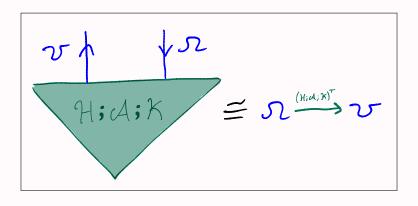


It can be closed by specifying an initial state and a utility function:





At the end of the day, the arena amounts to an evaluation of strategies with rewards:



Agents can then choose which strategies to play using a selection function:

$$\mathcal{E}: (\mathfrak{N} \longrightarrow \mathcal{V}) \longrightarrow \mathcal{P} \mathfrak{N}$$

Typically, Ω is finite, $\mho = \mathbb{R}$ and ε is argmax.

Interlude: selection functions

Let $\mathbb{S}(X,S):=(X\to S)\to \mathscr{P}X$. This is a functor on lenses:

$$\mathbb{S}(\alpha) : \mathbb{S}(X, S) \longrightarrow \mathbb{S}(Y, R)$$

$$\varepsilon \longmapsto \lambda k \cdot \{x \, ; \, \alpha \, | \, x \in \varepsilon(\alpha \, ; \, k)\}$$

where $\alpha:(X,S)\to (Y,R)$ and we implicitly used Lens $((1,1),(Y,R))\cong Y\to R$ and Lens $((X,S),(1,1))\cong X$:

(drawing)

This functor is lax monoidal (Nash product):

$$-\boxtimes -: \mathbb{S}(X,S) \times \mathbb{S}(Y,R) \longrightarrow \mathbb{S}(X \times Y, S \times R)$$
$$(\varepsilon,\eta) \longmapsto \lambda k \cdot \{(x,y) \,|\, x \in \varepsilon(k_1(x,y)) \text{ and } y \in \eta(k_2(x,y))\}$$

(drawing)

This yields a moncat Lens_S = $\int_{mon} S$.



The definition

Definition

An open game with agency G is a pair

$$\mathcal{A}: (X,S) \stackrel{(\Omega,\mho)}{\longrightarrow} (Y,R), \quad \varepsilon: \mathbb{S}(\Omega,\mho)$$

More succintly: a Lens_S-parametrised lens.

Agency \neq players: players arise from the presentation; if $\Omega = \prod_{p \in P} \Omega_p$, $\mho = \prod_{p \in P} \mho_p$ and $\varepsilon = \bigotimes_{p \in P} \varepsilon_p$ we speak of **open game with** (set of) **players** P.

Equilibria for $G = (A, \varepsilon)$ are given by:

$$eq_G(x, k) = \varepsilon(x \, \beta \, A \, \beta \, k).$$



Composing arenas

Arenas are the compositional heart of open games with agency:

Composing arenas

Notice: as of now, these operators can't be extended to selection functions in a canonical way!



Reparametrisation

Most importantly arenas form a (locally fibred) bicategory: one can **reparametrise** along a lens $\alpha: (\Omega', \mho') \to (\Omega, \mho)$.

This is crucial for introducing agency!

Regrouping

As an example, if $\mathcal A$ has set of players P and $r:P\to Q$ is a function, we can turn $\mathcal A$ into an arena with players Q by reparametrising along

$$\mathsf{regroup}_r: (\prod_{q \in Q} (r^*\Omega)_q, \prod_{q \in Q} (r^*\mho)_q) \longrightarrow (\prod_{p \in P} \Omega_p, \prod_{p \in P} \mho_p)$$

which only 'regroups' the factors according to r.

Example

If A_1 and A_2 have the same players P, $A_1 \, \S \, A_2$ has players P+P. So we regroup along $P+P \to P$ to get again an arena with players P.

Tying

The second most important application for reparametrisations is tying strategies at different point of the game. This is done by reparametrising along $(\Delta, \text{combine})$ (where combine $=+,\pi_2,\max,\ldots$).

Extensive form games & their translation

Extensive form

Extensive form is a classical formalism for games:

(nice pic)

A game is presented as a (rooted) tree, which should reflect the possible unfoldings of the game. Each node is labelled with a *player*. Branches represent possible *moves* and *payoff vectors* await at the leaves.

Two kinds:

- Perfect information: any player always knows exactly the state of the game (path from root to their nodes)
- 2. Imperfect information: otherwise

As customary in classical game theory, all players argmax their payoff.

PETs as inductive type

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We can easily represent PETs with an inductive type: (inductive type) where P: Set are players, R: Set are rewards. (tree and corresponding term)
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PETs

Correspondigly, we can define what strategies/moves/states/utilities are with inductive definitions: (def)

Translation

Finally, we can translate any PETree to an open game: (pic) where

- 1. $Dec([n], R^P)$ is the 'trivial' game where moves = strategies = [n] and coplay is just Δ .
- 2. regroup_p is regrouping along $\{p\} + P \rightarrow P$.



IETs

Imperfect information extensive form games are decorated with *information sets*: (pic)

Nodes in the same information set cannot be distinguished by players. Players do not share information sets, each player has their own partition of their nodes in information sets. Nodes in the same information set have the same moves.

IETs as inductive type

We can also represent IETs with an inductive type: (inductive type) where P: Set are players, R: Set are rewards, I: Set is a set of information labels, belongs: $I \to P$ assigns information sets to players, $n: I \to \mathbb{N}^+$ assigns uniformly moves to nodes of the same information set. (tree and corresponding term)

Conclusions & future work

In this work:

- We have seen how games naturally decompose in arenas and selection functions, with reparametrisations playing a key role
- 2. This allows to handle long-range correlations in players' behaviour
- 3. Hence we can easily map extensive form trees to open games with agency

What's next?

- 1. Is the translation 'functorial'? Possible domain cat described in [Str21].
- 2. Once the diagram is drawn, can we simplify it using topological moves? e.g. mapping 'simultaneity' to \otimes .
- 3. Infinite trees?
- 4. Can we treat SPE?
- 5. What kind of assignment is argmax: Arenas \rightarrow Games?
- 6. Can we make selections compositional?
- 7. Dependent types

Thanks for your attention!

Questions?

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