

Multiselves equilibria: Are you thinking what I'm thinking?

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16th June 2025



My intro to game theory: *Scientific American* 1983

# Superrationality

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In **economics** and **game theory**, a participant is considered to have **superrationality** (or **renormalized rationality**) if they have **perfect rationality** (and thus maximize their **utility**) but assume that all other players are superrational too and that a superrational individual will always come up with the same strategy as any other superrational thinker when facing the same problem. Applying this definition, a superrational player who assumes they are playing against a superrational opponent in a **prisoner's dilemma** will cooperate while a rationally self-interested player would defect.

This [decision rule](#) is not a mainstream model in [game theory](#) and was suggested by [Douglas Hofstadter](#) in his article, series, and book *[Metamagical Themas](#)*<sup>[1]</sup> as an alternative type of rational [decision making](#) different from the widely accepted [game-theoretic](#) one. Hofstadter provided this definition: "Superrational thinkers, by recursive definition, include in their calculations the fact that they are in a group of superrational thinkers."<sup>[1]</sup>

# Douglas Hofstadter's Luring Lottery

Douglas Hofstadter writes: The Luring Lottery, proposed in my June column, created quite a stir. Let me remind you that it was open to anyone; all you had to do was send a postcard with a clearly specified positive integer on it stating how many entries you wanted to make. This integer was to be, in effect, your “weight” in the final drawing, so that if you wrote “100,” your name would be 100 times as likely to be drawn as that of someone who wrote “1.” The only catch was that the cash value of the prize was inversely proportional to the sum of all the weights received by June 30. Specifically, the prize to be awarded was  $\$1,000,000/W$ , where  $W$  is the sum of all the weights sent in.

The Luring Lottery was set up as an

# Douglas Hofstadter's Luring Lottery

them. What was the breakdown of entries? Part of it is given in this table:

1:	1,133
2:	31
3:	16
4:	8
5:	16
6:	0
7:	9
8:	1
9:	1
10:	49
100:	61
1,000:	46
1,000,000:	33
1,000,000,000:	11
602,300,000,000,000,000,000,000	
(Avogadro's number):	1
Googol ( $10^{100}$ ):	9
Googolplex ( $10^{\text{googol}}$ ):	14

Curiously, many if not most of the people who submitted just one entry patted themselves on the back for being “cooperators.” Stuff and nonsense! The real cooperators were those among the

this. Scores of readers strained their hardest to come up with inconceivably large numbers. Some filled their entire postcard with tiny 9's, others filled their card with rows of exclamation points, creating iterated factorials of gigantic size. A handful of people carried this game much further, recognizing that the optimal solution avoids all pattern (for the reason see Gregory J. Chaitin's article “Randomness and Mathematical Proof,” *Scientific American*, Vol. 232, No. 5, pages 47–52; May, 1975). It consists simply of a “dense pack” of definitions built on definitions, followed by a final line in which the “fanciest” of the definitions is applied to a relatively small number such as 2 or, better yet, 9.

I received, as I say, a few such entries. Some of them exploited such powerful concepts of mathematical logic and set theory that to evaluate which one was the largest became a serious problem, and it is not even clear that I, or for that matter anyone else, would be able to determine which is the largest integer submitted. I was strongly reminded of

# Multiselves in Learning in games, and elsewhere

Example: Fictitious play. Robinson (1951): if both players use FP in a 2-player zero-sum game, payoffs converge to the game's value.

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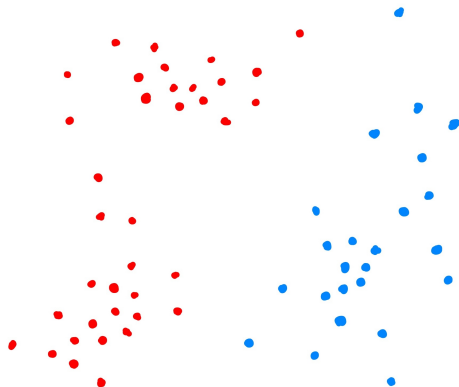
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Rendezvous problem (or dilemma): 2 robots want to end up in the same place.

# An obscure problem in ML theory

Introductory PAC learning problem:

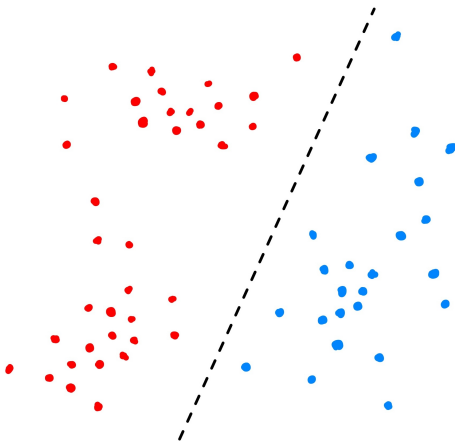


Want to classify linearly separable classes of points



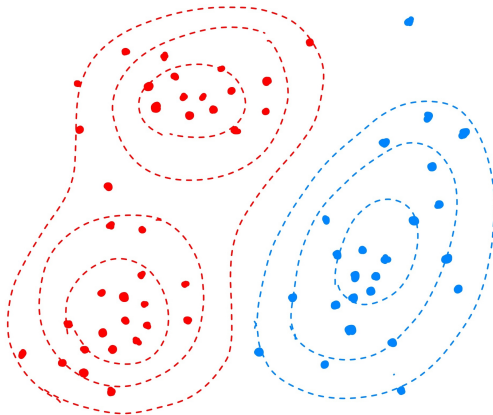
# An obscure problem in ML theory

Fact: any consistent linear separator works



# An obscure problem in ML theory

Fit distributions to data sets separately, use them to classify:



Some Discriminant-based PAC Algorithms, *JMLR* (2006)

# An obscure problem in ML theory

Design algorithm  $\mathcal{A}$  mapping point set to distribution such that:

If we run  $\mathcal{A}$  *twice*, once on the red points and once on the blue points, you get PAC learning when you combine the results in the obvious way.

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Problems:  $\mathcal{A}$  is complicated/artificial, and I never figured out how to learn propositional disjunctions...

It’s a “multi-selves” solution concept, of sorts

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Symmetric equilibrium: as a player, I take the view that my opponents are thinking what I'm thinking.

# Symmetric equilibria

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# Symmetric equilibria

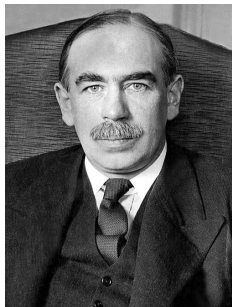
Objection: if you count on your opponents doing what you do, best to make no effort (incur no cost) and keep same probability to win the prize...

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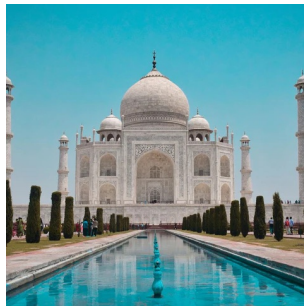
Is there a more meaningful scenario of counting on our opponents to do the same thing?

# A warm-up: the Keynesian beauty contest

“It is not a case of choosing those [faces] that, to the best of one’s judgment, are really the prettiest, nor even those that average opinion genuinely thinks the prettiest. We have reached the third degree where we devote our intelligences to anticipating what average opinion expects the average opinion to be. And there are some, I believe, who practice the fourth, fifth and higher degrees.” (Keynes, General Theory of Employment, Interest and Money, 1936).



# A warm-up: the Keynesian beauty contest



# Multiselves decision theory

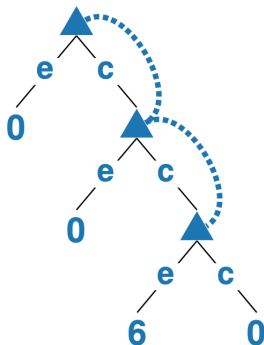
- Imperfect-Recall Games: Equilibrium Concepts and Their Complexity.  
*IJCAI 2024*
- The Computational Complexity of Single-Player Imperfect-Recall Games.  
*IJCAI 2023*

General set-up (examples to follow): extensive-form “game” with one player.

Decision tree is partitioned into information sets; within each info set all nodes have the same outdegree/labels.

# The absentminded driver

Sequence of exits off road, some exit other than the first is desired.  
How to proceed?



Absentmindedness: an info set may contain 2 nodes one of which follows from the other.

Note link to multiselves idea

# Sleeping Beauty

- 1 SB goes to sleep
- 2 Flip a fair coin, H: awaken SB once, T: awaken SB twice
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# Sleeping Beauty

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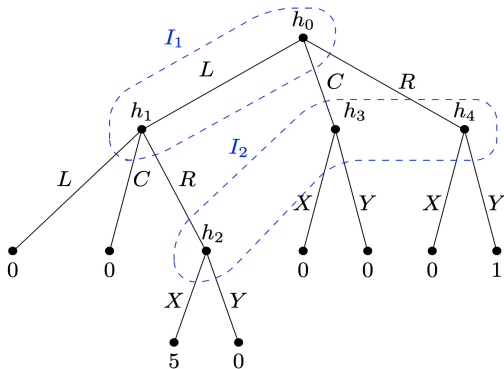
Answer 1:  $\frac{1}{2}$ : awaking provides no new information on the coin

Answer 2:  $\frac{1}{3}$ : if repeated, about  $\frac{1}{3}$  of wakings occur when heads is shown



# Causal vs. Evidential Decision Theory

The set-up: game tree with nodes  $\mathcal{H}$ , terminal nodes  $\mathcal{Z} \subseteq \mathcal{H}$ . Terminal nodes have real-valued *utilities*. There may be *chance nodes* that select one of their descendents with given probabilities. The other nodes are the player's *decision nodes*, partitioned into *info sets*.



Solution concept: for each info set  $I$ , a distribution over labels of actions that can be taken at nodes in  $I$ .

Given a game tree, utility maximisation is a problem of maximising a polynomial function subject to a collection of constraints.

NP-hard to compute an optimal strategy; results date back to Koller and Megiddo (1992).

$\exists\mathbb{R}$ -complete (Gimbert et al, 2020).

So, we look for solution concept that may be easier to find...

# Causal vs. Evidential Decision Theory

Two alternative solution concepts to 1-player extensive-form games with absentmindedness.

- Causal Decision Theory: Generalised Thirthing (GT)
- Evidential Decision Theory: Generalized Double Halving (GDH)

**Strategy:** maps each info set  $I$  to a distribution  $D_I$  over labels of its outgoing arcs.

Any strategy leads to an expected utility.

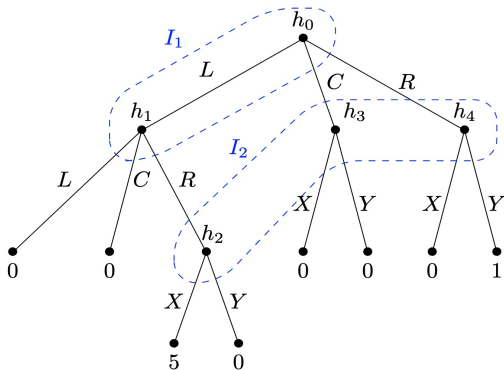
**Deviation:**

- (EDT, GDH): replace  $D_I$  with some  $D'_I$
- (CDT, GT): there's a distribution  $P_I$  over nodes in  $I$ . Replace  $D_I$  with  $D'_I$  at individual node sampled from  $P_I$ , assuming you leave  $D_I$  alone elsewhere.

**Equilibrium:** strategy with no profitable deviation.

(I think (CDT, GT) assumes SB always has opportunity to deviate, (EDT, GDH) assumes it's infrequent...)

# Example



Suppose player plays  $(R, X)$ . At info set  $I_1$ , consider deviation to  $L$ .  
 CDT: player will not deviate elsewhere in  $I_1$ , may win 5  
 EDT: deviation will lead to action history  $(L, L)$ .

Suppose player is in info set  $I_1$ , playing  $(C, X)$ .

No (EDT, GDH) deviation to a pure strategy, but to  $\frac{1}{2}L + \frac{1}{2}R$ .

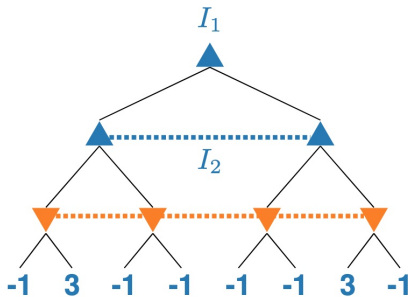
In games without absentmindedness, both solution concepts are the same.

Moving from global to local maximisation, takes to the class CLS... Also, (CDT,GT)-equilibrium is sort-of CLS-complete. In fact (CDT,GT)-equilibria are KKT points of the utility function. Containment requires a lower bound on positive visit frequencies to be easily obtainable. Hardness holds even for no chance nodes, tree depth 5, and one one info set.

CLS-completeness extends to (EDT,GDH) equilibria since it applies to games without absentmindedness.

Various NP-hardness results for search for equilibria having specific properties.

# Multi-player version (Tewolde et al, IJCAI'24)



Forgetful penalty shoot-out. There is no NE

## Multi-player

	Nash (D)	EDT (D)	CDT (S)
<b>exact</b>	$\exists\mathbb{R}$ -hard and in $\exists\mathbb{R}$ (Thms. 1 & 3)		—
<b>1/exp</b>	$\Sigma_2^P$ -complete (Thms. 2 and 4)		PPAD-complete (Thm. 6)
<b>1/poly</b>			

## Single-player

	Optimal (D)	EDT (S)	CDT (S)
<b>exact</b>	$\exists\mathbb{R}$ -complete [Gimbert <i>et al.</i> , 2020]	—	—
<b>1/exp</b>	NP-complete [Koller and Megiddo, 1992; Tewolde <i>et al.</i> , 2023]	PLS-complete (Thm. 5*)	CLS-complete [Tewolde <i>et al.</i> , 2023]
<b>1/poly</b>		P (Cor. 22*)	P (Cor. 17)

green: new results; (S): search problem; (D): decision problem

# Back to voting

Ongoing work with Tomasz Wąs.

Idea: in large-scale elections, we vote because we reckon others think like us. “social projection”

Model:  $n$  voters,  $m$  candidates. Each voter gets a signal about the candidates, drawn from a commonly-known distribution  $D$ .

Question: can we design distribution  $D$  that presents voters with a hard task? Variants:

- Voters are rewarded for voting for winner, à la KBC.
- ...or maybe, want to maximise welfare (seems challenging in the multi-winner setting)
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Thanks!