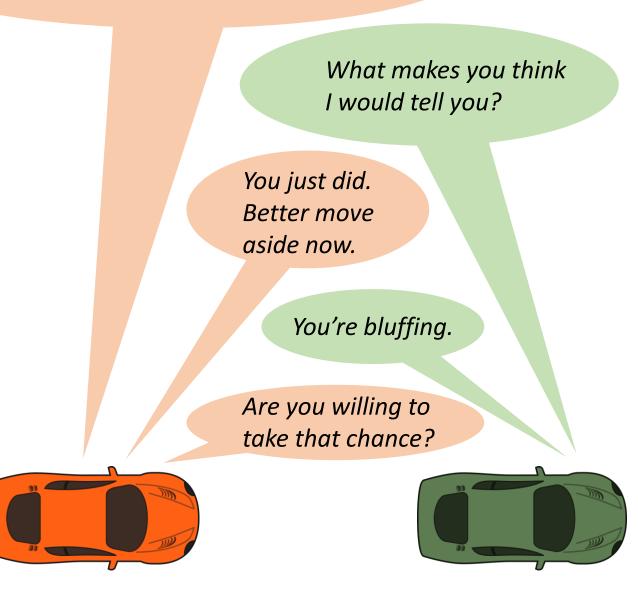
If I tailgate you, will your occupant take back control and pull over?

Al Agents May Cooperate Better If They Don't Resemble Us

Vincent Conitzer

Early blue sky paper: <u>Designing Preferences, Beliefs, and Identities</u> <u>for Artificial Intelligence.</u> In Proceedings of the Thirty-Third AAAI Conference on Artificial Intelligence (AAAI-19).

Also see Cooperative AI community <u>https://www.cooperativeai.com/</u> and our new lab at CMU! <u>http://www.cs.cmu.edu/~focal/</u>



Outline

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		ly: The Genetics of	f Animal Design (Paperback	() Pric	e at a Glance
by	y Peter A. Lawrence			Lis	
	Return to product inf	ormation		Price	: from \$35.54
	Always pay through Amazon.com's Shopping Cart or 1-Click. Learn more about Safe Online Shopping and our safe buying guarantee.				from \$1,730,045.91
				Have one	to sell? Sell yours here
All New	(2 from \$1,730,045.91)	Used (15 from \$35.54)			
Show ^O New	O Anime offers only	(0)		Sorted t	Price + Shipping
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rice + Shipping \$1,730,045.9	Condition	Seller: prof Seller Rating (8,193 total r In Stock, Shi Domestic shi	inath a AAAAA: 93% positive over the past ratings) ips from NJ, United States.	12 months	Add to Cart Or Sign in to turn on 1-Click
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\$1,730,045.9 \$1,730,045.9 \$3.99 shipping \$2,198,177.9	Condition	Seller: prof Seller Rating (8,193 total r In Stock. Shi Domestic shi Brand new, F Seller: bord Seller Rating (125,891 tota In Stock. Shi	inath : ******* 93% positive over the past ratings) ips from NJ, United States. pping rates and return policy. Perfect condition, Satisfaction Guaranteed. deebook : ******* 93% positive over the past	12 months.	Add to Cart or Sign in to turn on 1-Click ordering.

From The Atlantic, "Want to See How Crazy a Bot-Run Market Can Be?" By <u>James Fallows</u> April 23, 2011 OLIVIA SOLON BUSINESS 04.27.2011 03:35 PM

SIGN IN SUBSCRIBE

How A Book About Flies Came To Be Priced \$24 Million On Amazon

Two booksellers using Amazon's algorithmic pricing to ensure they were generating marginally more revenue than their main competitor ended up pushing the price of a book on evolutionary biology — Peter Lawrence's The Making of a Fly — to \$23,698,655.93. [partner id="wireduk"]The book, which was published in 1992, is out of print but is commonly [...]

Two booksellers using Amazon's algorithmic pricing to ensure they were generating marginally more revenue than their main competitor ended up pushing the price of a book on evolutionary biology -- Peter Lawrence's <u>The Making of a Fly</u> -- to \$23,698,655.93.

[partner id="wireduk"]The book, which was published in 1992, is out of print but is commonly used as a reference text by <u>fly experts</u>. A post doc student working in Michael Eisen's lab at UC Berkeley first discovered the pricing glitch when looking to buy a copy. As <u>documented on Eisen's blog</u>, it was discovered that Amazon had 17 copies for sale -- 15 used from \$35.54 and two new from \$1,730,045.91 (one from seller <u>profnath</u> at that price and a second from <u>bordeebook</u> at \$2,198,177.95).

This was assumed to be a mistake, but when Eisen returned to the page the next day, he noticed the price had gone up, with both conies on offer for around \$2.8 million. By the end of the day, profnath had raised its price again to \$3,536,674.57. He worked out that once a day, profnath set its price to be 0.9983 times the price of the copy offered by bordeebook (keen to undercut its competitor), meanwhile the prices of bordeebook were rising at 1.270589 times the price offered by profnath.

WATCH

Maleficent: Re-creating Fully Digital Characters



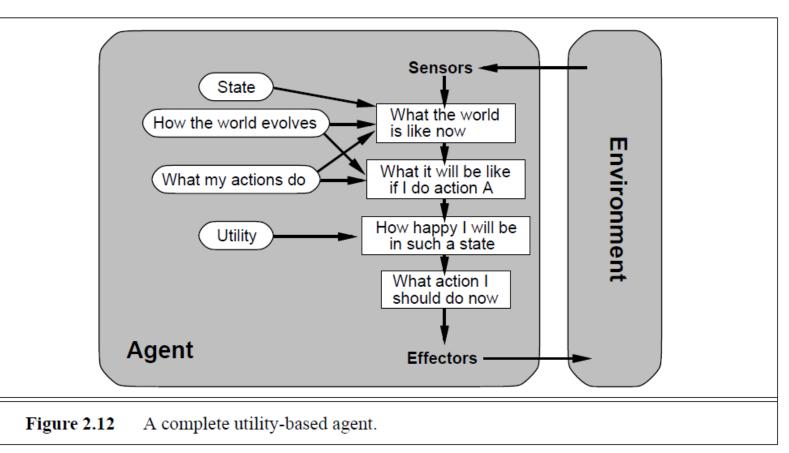


The May 6, 2010, flash crash, ^{[1][2][3]} also known as the crash of 2:45 or simply the flash crash, was a United States trilliondollar^[4] stock market crash, which started at 2:32 p.m. EDT and lasted for approximately 36 minutes.^{[5]:1} Between 2:45:13 and 2:45:27, HFTs traded over 27,000 contracts, which accounted for about 49 percent of the total trading volume, while buying only about 200 additional contracts net.

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Russell and Norvig's "AI: A Modern Approach"







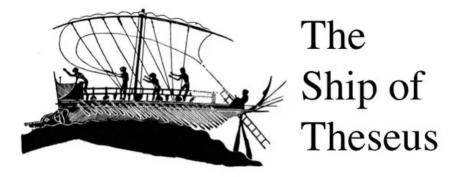
Stuart Russell

Peter Norvig

"... we will insist on an objective performance measure imposed by some authority. In other words, we as outside observers establish a standard of what it means to be successful in an environment and use it to measure the performance of agents."

What should we want? What makes an individual?

- Questions studied in philosophy
 - What is the "good life"?
 - Ship of Theseus: does an object that has had all its parts replaced remain the same object?
- Al gives a new perspective



Personal Identity

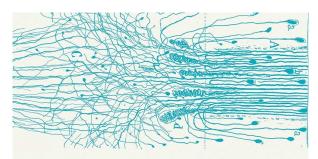
What ensures my survival over time?

The Bodily Criterion
The Brain Criterion
The Psychological Criterion John Locke



image from <u>https://www.quora.com/What-solutions-are-</u> there-for-the-Ship-of-Theseus-problem

Al Alignment

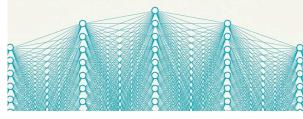


THE ALIGNMENT PROBLEM

Machine Learning and Human Values

BRIAN CHRISTIAN

Best-Selling Author, Algorithms to Live By





Fifth AAAI /ACM Conference on Artificial Intelligence, Ethics, and Society Oxford August 1-3, 2022





Institute for Ethics in Al

Oxford leading the way in Al ethics

Stanford University

One Hundred Year Study on Artificial

Intelligence (AI100)

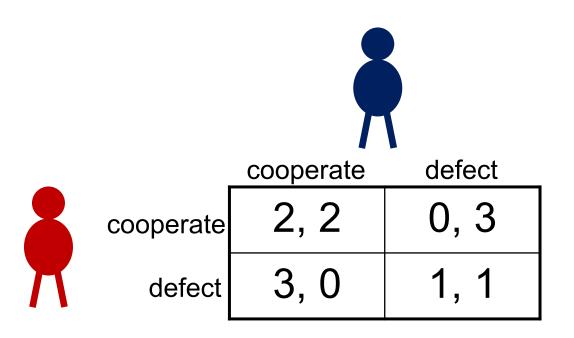
Even almost perfectly aligned agents can perform horribly in equilibrium

- Two agents each provide part of a service, each chooses quality q_i
- Overall quality determined by min_i q_i
- Agents care primarily about overall quality, but also have a slight incentive to be the lower one

	100	90	80	70	60	50	40	30	20	10	0	
100	111, 111	90, 112	80, 102	70, 92	60, 82	50, 72	40, 62	30, 52	20, 42	10, 32	0, 22	
90	112, 90	101, 101	80, 102	70, 92	60, 82	50, 72	40, 62	30, 52	20, 42	10, 32	0, 22	
80	102, 80	102, 80	91, 91	70, 92	60, 82	50, 72	40, 62	30, 52	20, 42	10, 32	0, 22	
70	92, 70	92, 70	92, 70	81, 81	60, 82	50, 72	40, 62	30, 52	20, 42	10, 32	0, 22	
60	82, 60	82, 60	82, 60	82, 60	71, 71	50, 72	40, 62	30, 52	20, 42	10, 32	0, 22	
50	72, 50	72, 50	72, 50	72, 50	72, 50	61, 61	40, 62	30, 52	20, 42	10, 32	0, 22	
40	62, 40	62, 40	62, 40	62, 40	62, 40	62, 40	51, 51	30, 52	20, 42	10, 32	0, 22	
30	52, 30	52, 30	52, 30	52, 30	52, 30	52, 30	52, 30	41, 41	20, 42	10, 32	0, 22	
20	42, 20	42, 20	42, 20	42, 20	42, 20	42, 20	42, 20	42, 20	31, 31	10, 32	0, 22	
10	32, 10	32, 10	32, 10	32, 10	32, 10	32, 10	32, 10	32, 10	32, 10	21, 21	0, 22	
0	22, 0	22, 0	22, 0	22, 0	22, 0	22, 0	22, 0	22, 0	22, 0	22, 0	11, 11	

(Cf. Traveler's Dilemma)

Prisoner's Dilemma



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Science Q ALL NEWS > HUMAN ALTRUISM TRACES BACK TO THE ORIGINS OF HUMANITY NEWS | BRAIN & BEHAVIOR Human altruism traces back to the origins

of humanity

Study probes why humans are more cooperative than other animals

27 AUG 2014 · BY MICHAEL BALTER



BIOLOGY PSYCHOLOGY Cooperation Is What Makes Us Human

Where we part ways with our ape cousins.

BY KAT MCGOWAN ILLUSTRATIONS BY JOHN HENDRIX APRIL 29, 2013



BROWSE BY SUBJECT FREE TRIAL

Philos Trans R Soc Lond B Biol Sci. 2010 Sep	PMCID: PMC2936178
12; 365(1553): 2663–2674.	PMID: 20679110
doi: 10.1098/rstb.2010.0157	

How is human cooperation different?

Alicia P. Melis^{1,*} and Dirk Semmann^{2,*}

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This article has been cited by other articles in PMC.

ABSTRACT

Philos Trans R

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Biol Sci

Go to: 🖂

Although cooperation is a widespread phenomenon in nature, human cooperation exceeds that of all other species with regard to the scale and range of cooperative activities. Here we review and

Why We're So Nice: We're Wired to Cooperate

By Natalie Angier

July 23, 2002

FOREIGN AFFAIRS When the System Fails

When the System Fails

COVID-19 and the Costs of Global Dysfunction

By Stewart Patrick July/August 2020



) ad

he chaotic global response to the coronavirus pandemic has tested the faith of even the most ardent internationalists. Most nations, including the world's most powerful, have turned inward, adopting travel bans, implementing export controls, hoarding or obscuring



Googl

Why International Cooperation is Failing

How the Clash of Capitalisms Undermines the Regulation of Finance

Settings and

Thomas Kalinowski

- Provides a new alternative to liberal and realist mainstream theories of International Political Economy
- Extends research in Comparative and International Political Economy beyond eurocentrism and nation state focus to studies of East Asian and euro capitalism
- Provides a new methodological approach to International Studies by combining International

WHY COOPERATION FAILED IN 1914

By STEPHEN VAN EVERA*

T HE essays in this volume explore how three sets of factors affect the degree of cooperation or non-cooperation between states. The first set comprises the "structures of payoffs" that states receive in return for adopting cooperative or noncooperative policies; payoff structures are signified by the rewards and penalties accruing to each state from mutual cooperation (CC); cooperation by one state and "defection" by another (CD and DC); and mutual defection (DD). The second set comprises the "strategic setting" of the international "game"—that is, the rules and conditions under which international relations are conducted. Two aspects of the strategic setting are considered: the size of the "shadow of the future," and the ability of the players to "recognize" past cooperators and defectors, and to distinguish between them.¹ The third set is the number of players in the game, and the influence these

The Global Climate Talks Ended In Disappointment

One activist group pronounced the conclusions a "pile of shite" and dumped manure outside the meeting hall.



Posted on December 15, 2019, at 10:29 a.m. ET





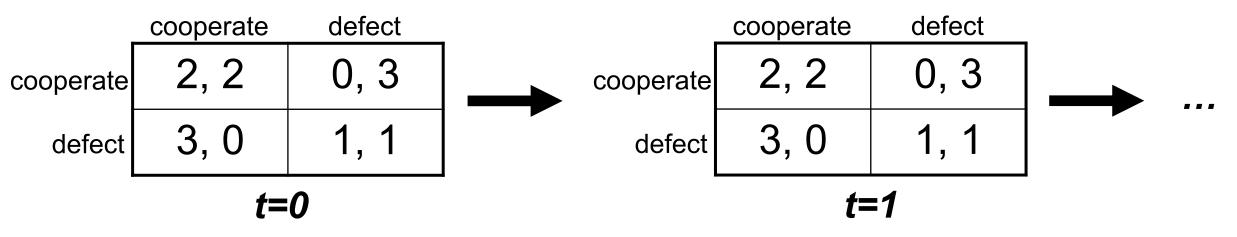
Some (highly interdisciplinary) discussion points: Should we make AI more human-like?

- Should we make our agents have prosocial inclinations? Ethics?
 - Genuine solution vs. wishful thinking?
 - What about norms and rules?
- Do certain human cognitive limitations limit tragedies? Should/can we replicate that in AI agents?
 - Traveler's dilemma
 - Any fool can tell the truth, but it requires a man of some sense to know how to lie well. -- Samuel Butler
- Might AI do better on cooperation than humans? On its own? With some deliberate design decisions?

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Infinitely Repeated Prisoner's Dilemma



- **Grim trigger** strategy: cooperate as long as everyone cooperates; after that, defect forever. (Equilibrium, if players are somewhat patient.)
- *Folk theorem*: with sufficiently patient players, can always sustain cooperation this way, in any game.
- Folk theorem can be used to efficiently compute equilibria (in infinitely repeated games with sufficiently patient players) [Littman & Stone DSS 2005, Andersen & C., AAAI'13]

Repeated games on social networks [Moon & C., IJCAI'15]

- Common assumption: an agent's behavior is instantly observable to all other agents (instant punishment)
- What if there is a delay in knowledge propagation due to network structure?

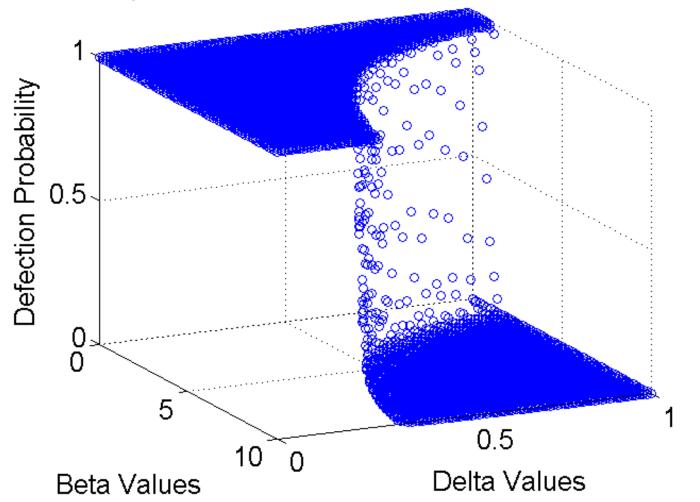


Catherine Moon



• Algorithm for finding (unique) maximal set of cooperating agents

Experiments on random graphs: **Phase transition** between complete cooperation and complete defection



Random graph models: Erdős–Rényi (ER) Barabási–Albert preferentialattachment (PA)

Beta = cooperation benefit, delta = discount factor

Disarmament Game [Deng & C., AAAI'17, '18]



Yuan Deng

	GR	KR	ST
R	10, -5	0, -4	0, -4
PF	4, 1	-10, 3	-0.5, -0.5
PD	-6, 8	-10, 3	-0.5, -0.5

Disarmament Game

	GR	KR	ST		
R	10, -5	0, -4	0, -4		
PF	4, 1	-10, 3	-0.5, -0.5		
PD	-6, 8	J, 3	-0.5, -0.5		
Pure Nash equilibria Pure Stackelberg equilibria (no matter who takes the lead)					

Disarmament Game

	GR	KR	ST	
R	10, -5	0, -4	0, -4	
PF	4, 1	-10, 3	-0.5, -0.5	
PD	-6, 8	-10, 3	-0.5, -0.5	
Desired Outcome Pareto better than the Nash equilibrium outcome				

	GR	KR	ST
R	10, -5	0, -4	0, -4
PF	4, 1	-10, 3	-0.5, -0.5
PD	-6, 8	-10, 3	-0.5, -0.5

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	GR	ST
R	10, -5	0, -4
PF	4, 1	-0.5, -0.5
PD	-6, 8	-0.5, -0.5

Incentivize Row to commit in the next round

	GR	ST
R	10, -5	0, -4
PF	4, 1	-0.5, -0.5
PD	-6, 8	-0.5, -0.5

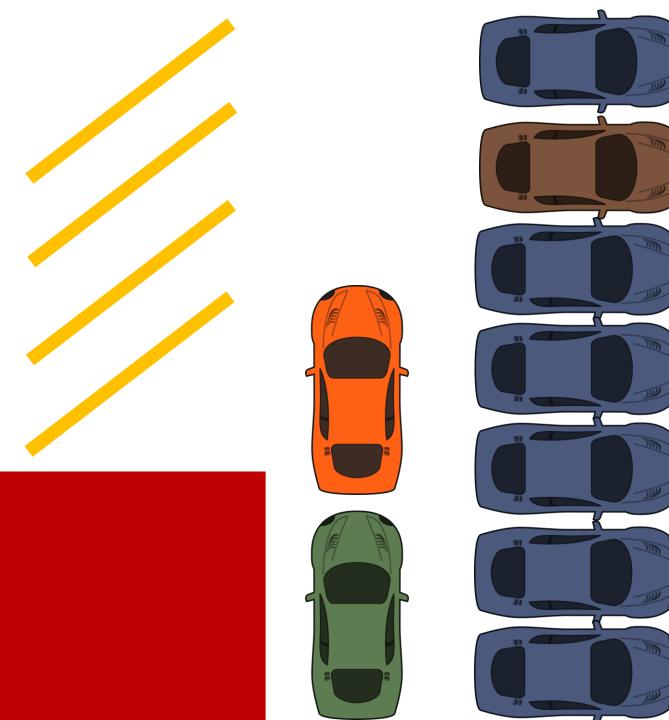
	GR	ST
PF	4, 1	-0.5, -0.5
PD	-6, 8	-0.5, -0.5

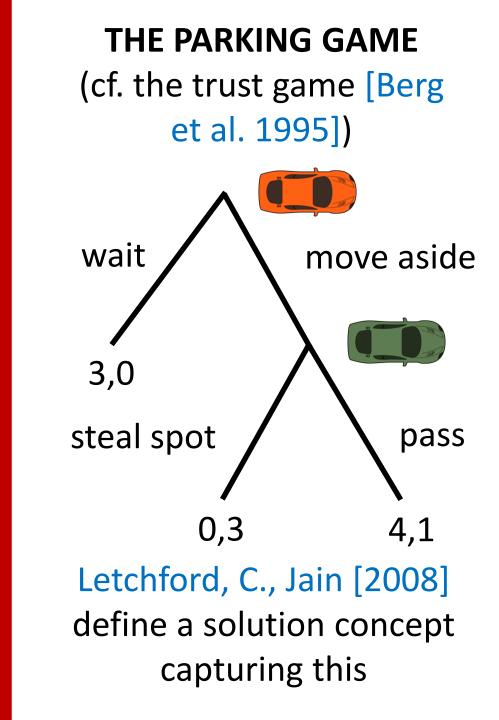
	GR	ST
PF	4, 1	-0.5, -0.5
PD	-6, 8	-0.5, -0.5

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	GR	KR	ST
R	10, -5	0, -4	0, -4
PF	4, 1	-10, 3	-0.5, -0.5
PD	-6, 8	-10, 3	-0.5, -0.5

Fact: The desired outcome cannot be achieved if Row commits first In general, it is an NP-hard problem to determine whether an outcome can be reached without creating incentive to deviate from disarmament

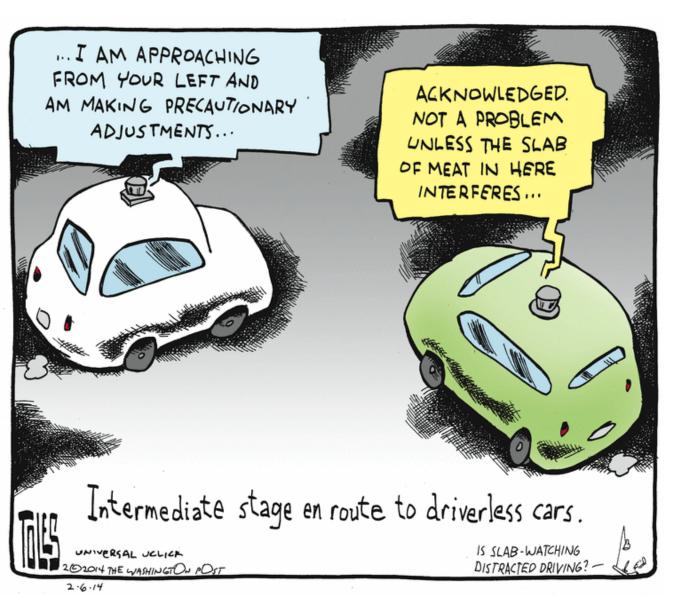




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Example: network of self-driving cars



- Should this be thought of as one agent or many agents?
- Should they have different preferences -- e.g., act on behalf of owner/occupant?
 - May increase adoption [Bonnefon, Shariff, and Rahwan 2016]
- Should they have different *beliefs* (e.g., not transfer certain types of data; erase local data upon ownership transfer; ...)?

Splitting things up in different ways

t = 2

t = 2

t = 1

t = 1



shared objective but no data sharing (for privacy)

all data is shared but cars act on behalf of owner

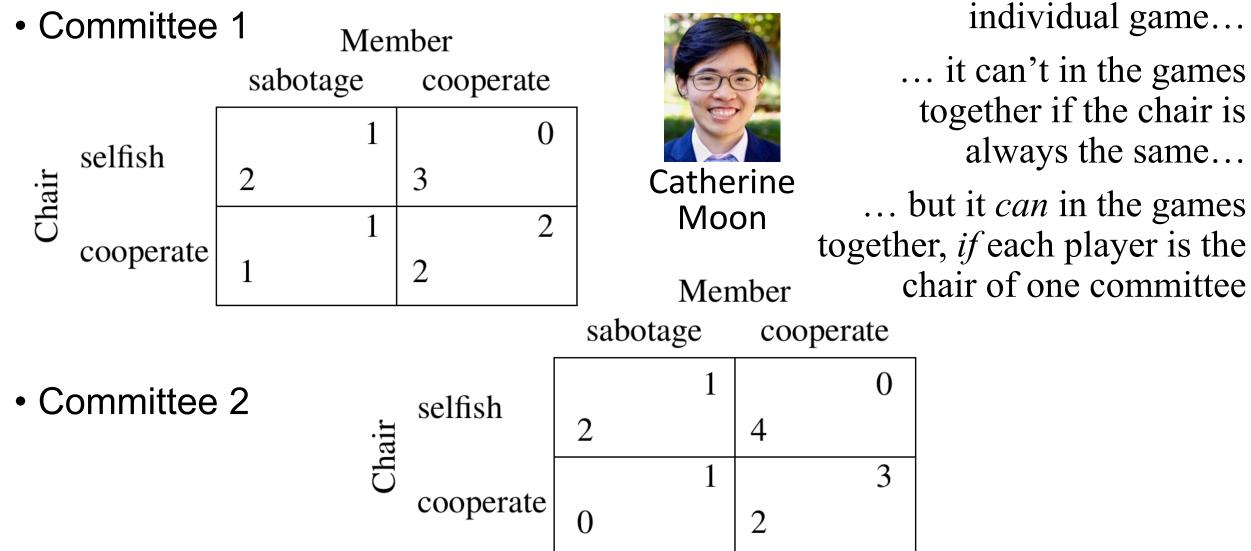
shared objective over time but data erasure upon sale (for privacy)

data is kept around but car acts on behalf of current owner

(cooperate, cooperate) Role assignment [Moon & C., IJCAI'16] cannot be sustained even

Two individuals for roles in two committees

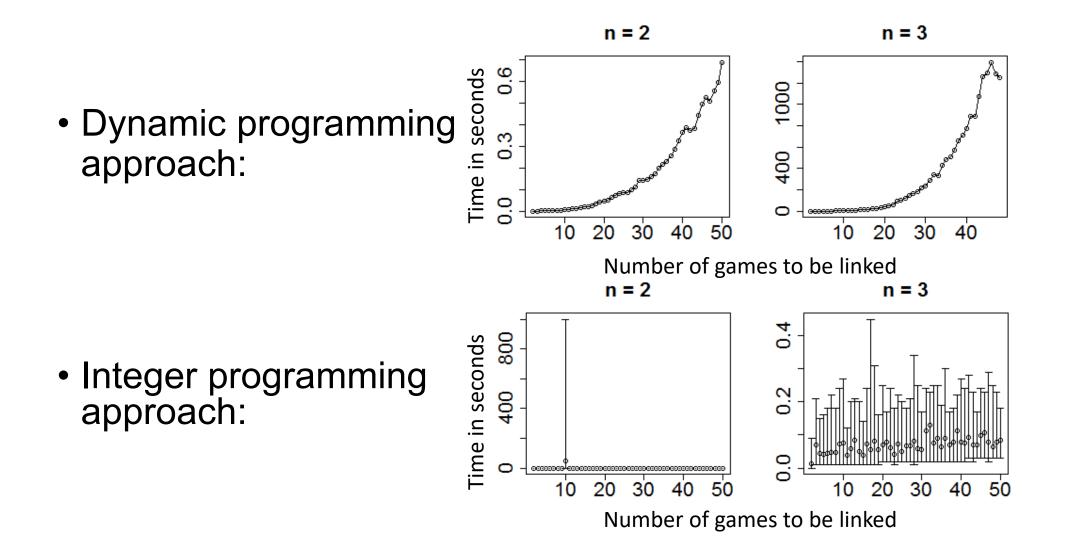
Committee 1

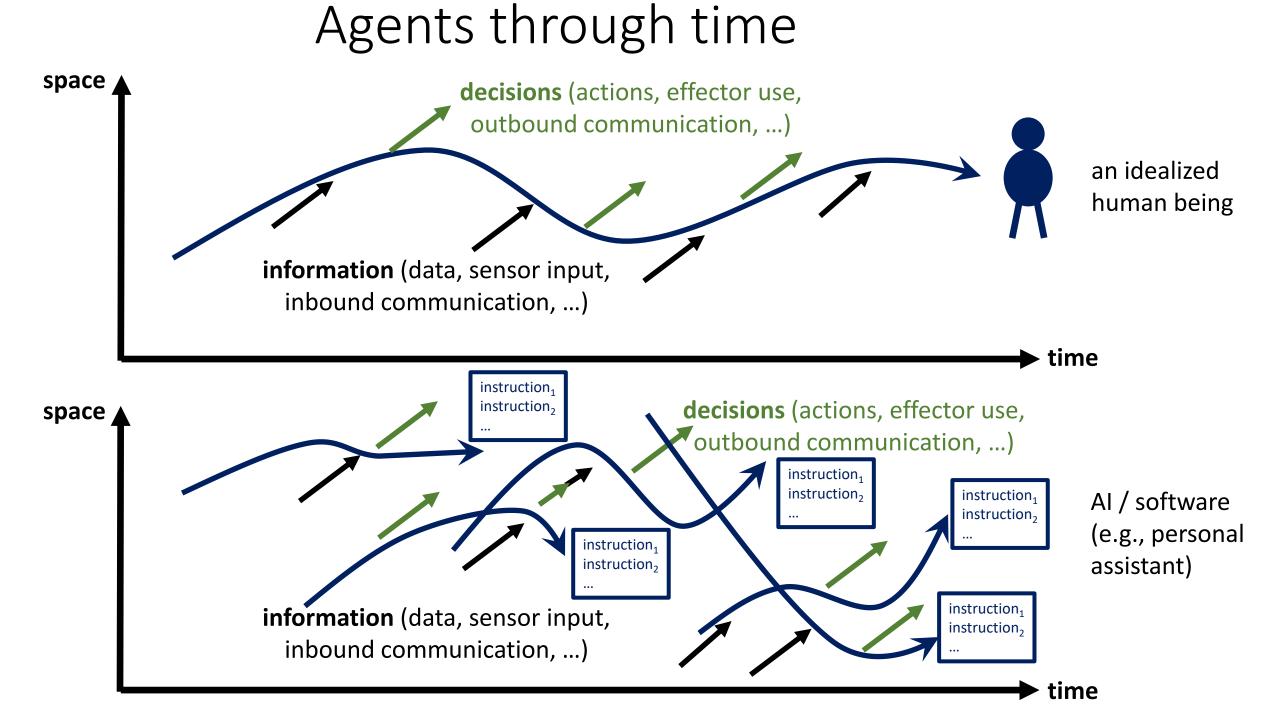


with repetition in each

Computation for optimal role assignment

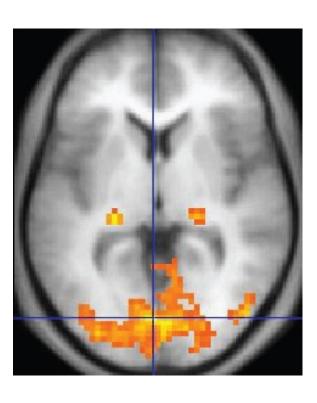
Problem is NP-hard



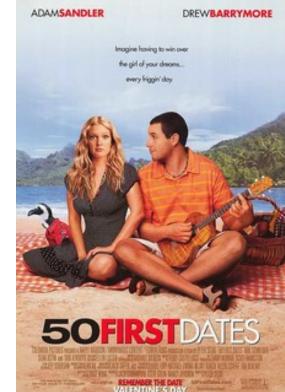


What should you do if...

- ... you knew others could read your code?
- ... you knew you were facing someone running the same code?
- ... you knew you had been in the same situation before but can't possibly remember what you did?

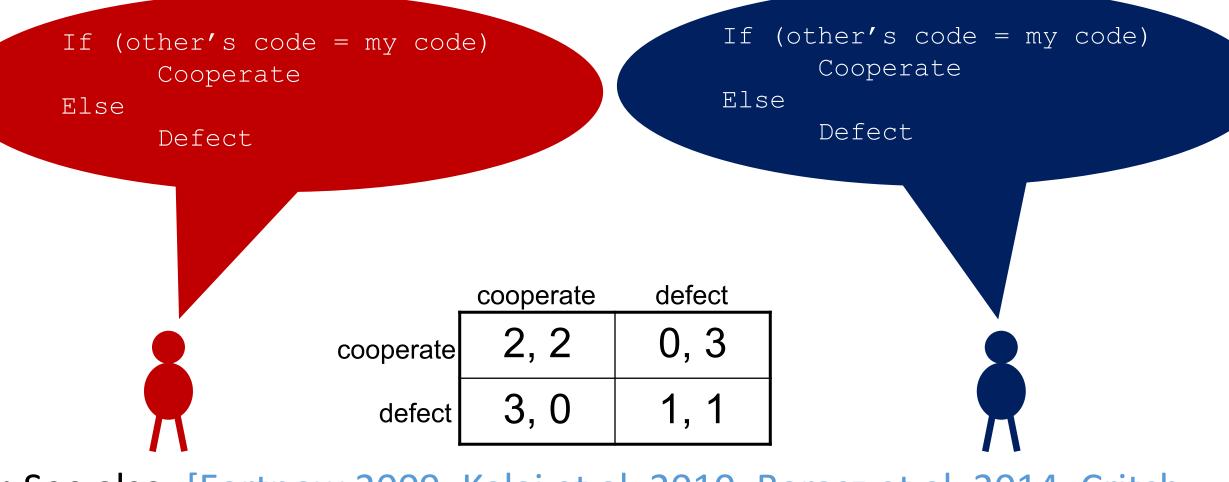






Program equilibrium [Tennenholz 2004]

• Make your own code legible to the other player's program!



• See also: [Fortnow 2009, Kalai et al. 2010, Barasz et al. 2014, Critch 2016, Oesterheld 2018, ...]

Robust program equilibrium [Oesterheld 2018]

• Can we make the equilibrium less fragile?

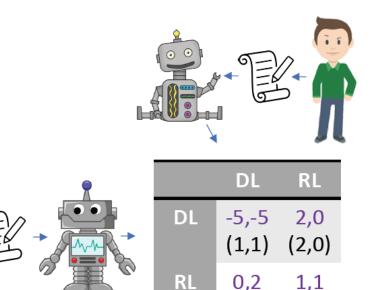


With probability ɛ ... Cooperate Else Do what the other program does against this program cooperate defect 2, 2 0, 3 cooperate 3, 0 1, 1 defect

Safe Pareto improvements for delegated game playing [AAMAS'21], with



Caspar Oesterheld

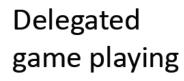


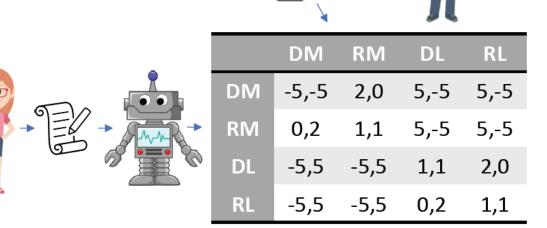
• Each player's contract says: Play this alternative game if the other player adopts an analogous contract.

(0,2)

(1,1)

- The games are essentially isomorphic.
 - DM ~ DL
 - RM ~ RL
- Safe Pareto improvement on the original game: outcome of new game is better for both players with certainty.



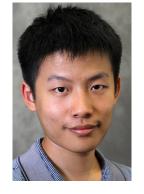


 Representatives are competent at playing games and the original players trust the representatives.

=> Default: aligned delegation

- DL,RL are strictly dominated and therefore never played
- Equilibrium selection problem
 - => Pareto-suboptimal outcome (DM,DM) might occur

Disarmament revisited: Committing to your first few lines of code



. . .

defect

0, 3

1, 1

Yuan Deng

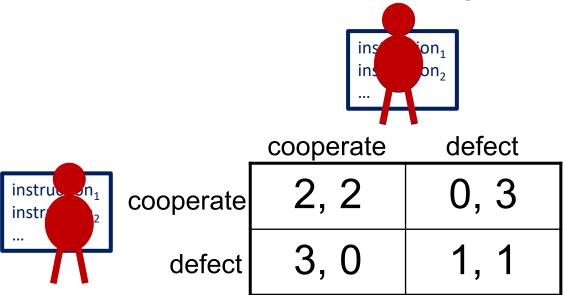
1. With probability 40%, cooperate 3. With probability 40%, cooperate defect 3, 0

• • •

With probability
 40%, cooperate
 4. With probability
 40%, cooperate

- E.g., if Blue refuses to add line 2, then Red defects with probability .6, resulting in at most .4*3 + .6*1 = 1.8 for Blue
- "Folk theorem" [Deng & C., AAAI'17, '18] that cooperation can always be achieved this way!

Prisoner's Dilemma against (possibly) a copy



- What if you play against your twin that you always agree with?
- What if you play against your twin that you almost always agree with?

related to working paper [Oesterheld, Demski, C.]

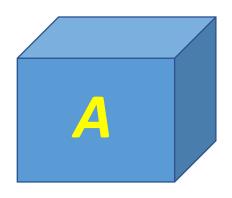


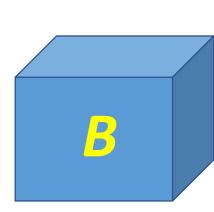


Caspar Oesterheld Abram Demski

Newcomb's Demon

- Demon earlier put positive amount of money in each of two boxes
- Your choice now: (I) get contents of Box B, or (II) get content of **both** boxes (!)
- Twist: demon first predicted what you would do, is uncannily accurate
- If demon predicted you'd take just B, there's \$1,000,000 in B (and \$1,000 in A)
- Otherwise, there's \$1,000 in each
- What would **you** do?

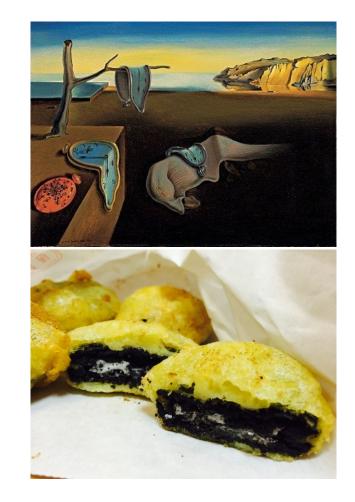


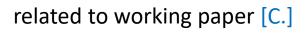




The lockdown dilemma

- Lockdown is **monotonous**: you forget what happened before, you forget what day it is
- Suppose you know lockdown lasts two days (unrealistic)
- Every morning, you can decide to eat an unhealthy cookie! (or not)
- Eating a cookie will give you +1 utility immediately, but then -3 later the *next* day
- But, carpe diem: you only care about today
- Should you eat the cookie right now?

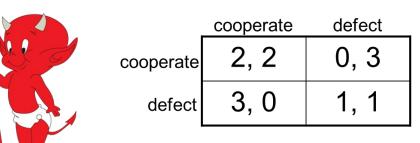




Your own choice is **evidence**...

- ... for what the demon put in the boxes
- ... for whether your twin defects
- ... for whether you eat the cookie on the other day

- Evidential Decision Theory (EDT): When considering how to make a decision, consider how happy you expect to be conditional on taking each option and choose an option that maximizes that
- Causal Decision Theory (CDT): Your decision should focus on what you causally affect

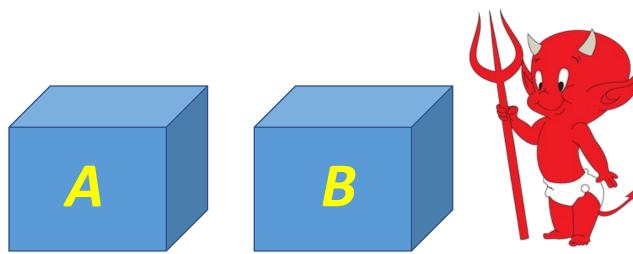




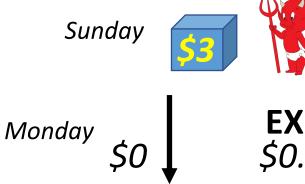
Turning causal decision theorists into money pumps

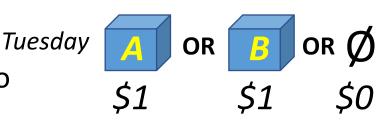
[Oesterheld and C., Phil. Quarterly]

- Adversarial Offer:
- Demon (really, any good predictor) put \$3 into each box it predicted you would not choose
- Each box costs \$1 to open; can open at most one
- Demon 75% accurate (you have no access to randomization)
- CDT will choose one box, knowing that it will regret doing so
- Can add earlier opt-out step where the demon promises not to make the adversarial offer later, if you pay the demon \$0.20 now









Imperfect recall

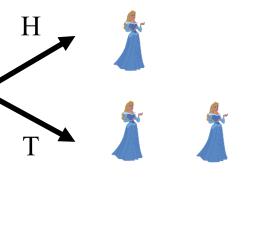
- An AI system can deliberately forget or recall
- Imperfect recall already used in poker-playing AI
 - [Waugh et al., 2009; Lanctot et al., 2012; Kroer and Sandholm, 2016]
- But things get weird....





The Sleeping Beauty problem [Elga'00]

- There is a participant in a study (call her Sleeping Beauty)
- On Sunday, she is given drugs to fall asleep
- A coin is tossed (H or T)
- If H, she is awoken on Monday, then made to sleep again
- If T, she is awoken Monday, made to sleep again, then **again** awoken on Tuesday
- Due to drugs she cannot remember what day it is or whether she has already been awoken once, but she remembers all the rules
- Imagine you are SB and you've just been awoken. What is your (subjective) probability that the coin came up H?



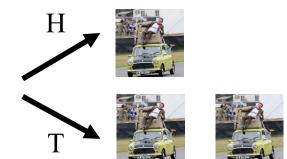
Sunday Monday Tuesday

don't do this at home / without IRB approval...

Modern version

- Low-level autonomy cars with AI that intervenes when driver makes major error
- Does not keep record of such event
- Two types of drivers: Good (1 major error), Bad (2 major errors)
- Upon intervening, what probability should the AI system assign to the driver being good?

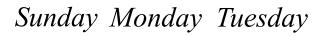
Sunday Monday Tuesday

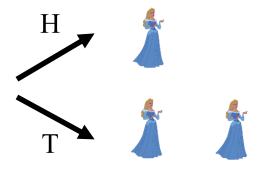




Taking advantage of a Halfer [Hitchcock'04]

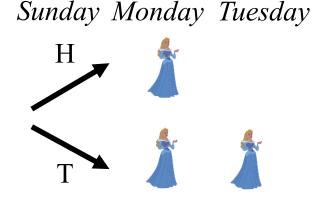
- Offer Beauty the following bet *whenever she awakens*:
 - If the coin landed Heads, Beauty receives 11
 - If it landed Tails, Beauty pays 10
- Argument: Halfer will accept, Thirder won't
- If it's Heads, Halfer Beauty will get +11
- If it's Tails, Halfer Beauty will get -20
- Can combine with another bet to make Halfer Beauty end up with a sure loss (a Dutch book)





Evidential decision theory

- Idea: when considering how to make a decision, should consider what it would tell you about the world if you made that decision
- EDT Halfer: "With prob. ½, it's Heads; if I accept, I will end up with 11. With prob. ½, it's Tails; if I accept, then I expect to accept the other day as well and end up with -20. I shouldn't accept."
- As opposed to more traditional causal decision theory (CDT)
- CDT Halfer: "With prob. ½, it's Heads; if I accept, it will pay off 11. With prob. ½, it's Tails; if I accept, it will pay off -10. Whatever I do on the other day I can't affect right now. I should accept."
- EDT Thirder can also be Dutch booked
- CDT Thirder and EDT Halfer cannot
 - [Draper & Pust '08; Briggs '10]
- EDTers arguably can in more general setting
 - [C., Synthese'15]
 - ... though we've argued against CDT in other work [Oesterheld & C, Phil. Quarterly'21]



Dutch book against EDT [C. 2015]

• Modified version of Sleeping Beauty where she wakes up in rooms of various colors

	WG $(1/4)$	WO $(1/4)$	BO $(1/4)$	BG $(1/4)$
Monday	white	white	black	black
Tuesday	grey	black	white	grey

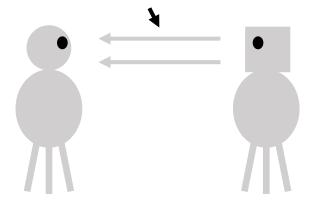
Fig. 3 Sequences of coin tosses and corresponding room colors, as well as their probabilities, in the WBG Sleeping Beauty variant.

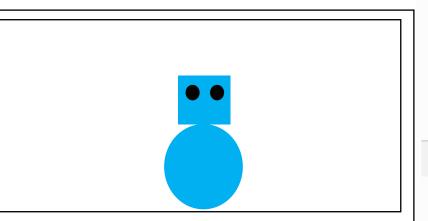
	WG $(1/4)$	WO $(1/4)$	BO $(1/4)$	BG $(1/4)$
Sunday	bet 1: 22	bet 1: -20	bet 1: -20	bet 1: 22
Monday	bet 2: -24	bet 2: 9	bet 2: 9	bet 2: -24
Tuesday	no bet	bet $2:9$	bet 2: 9	no bet
total gain from accepting all bets	-2	-2	-2	-2

Fig. 4 The table shows which bet is offered when, as well as the net gain from accepting the bet in the corresponding possible world, for the Dutch book presented in this paper.

Philosophy of "being present" somewhere, sometime

simulated light (no direct correspondence to light in our world)





1: world with creatures 2: displayed perspective simulated on a computer of one of the creatures

- To get from 1 to 2, need *additional* code to:
 - A. determine *in which real-world colors* to display perception *See also: [Hare 2007-2010, Valberg 2007, Hellie 2013, Merlo 2016, ...]*
 - B. which agent's perspective to display
- Is 2 more like our own conscious experience than 1? If so, are there *further facts* about presence, perhaps beyond physics as we currently understand it?

Erkenntnis

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A Puzzle about Further Facts

Authors	Authors and affiliations
Vincent Conitzer 🖂	
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Abstract

In metaphysics, there are a number of distinct but related questions about the existence of "further facts"—facts that are contingent relative to the physical structure of the universe. These include further facts about qualia, personal identity, and time. In this article I provide a sequence of examples involving computer simulations, ranging from one in which the protagonist can clearly conclude such further facts exist to one that describes our own condition. This raises the question of where along the sequence (if at all) the protagonist stops being able to soundly conclude that further facts exist.

Keywords

Metaphysics Philosophy of mind Epistemology

Absentminded Driver Problem [Piccione and Rubinstein, 1997]

- Driver on monotonous highway wants to take second exit, but exits are indistinguishable and driver is forgetful
- Deterministic (behavioral) strategies are not *stable*
- Optimal randomized strategy: exit with probability p where p maximizes $4p(1-p) + (1-p)^2 = -3p^2 + 2p + 1$, so $p^* = 1/3$
- What about "from the inside"? P&R analysis: Let b be the belief/credence that we're at X, and p the probability that we exit. Maximize with respect to p: (1-b)(4p+1(1-p)) + b(4p(1-p) + 1(1-p)²) = -3bp² + (3-b)p + 1, so p* = (3-b) / (6b) = 1/(2b) 1/6
- But if p = 1/3, then b = 3/5, which would give p* = 5/6 1/6 = 2/3?
 So also not stable?
- Resembles EDT reasoning... But not really halfing... Shouldn't b depend on p...

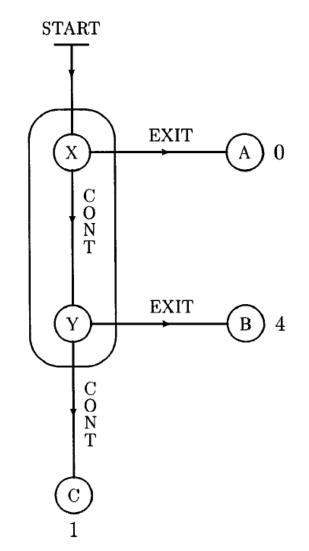


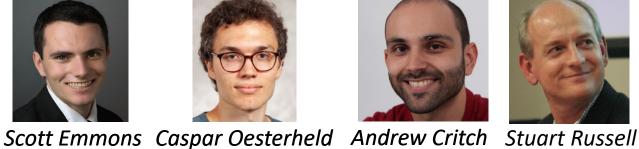
FIG. 1. The absent-minded driver problem.

Image from Aumann, Hart, Perry 1997

A different analysis [Aumann, Hart, Perry, 1997]

- AHP reason more along thirder / CDT lines:
- Imagine we normally expect to play p = 1/3. Should we deviate this time only?
- If we exit now, get (3/5)*0 + (2/5)*4 = 8/5
- If we continue now, get $(3/5)^*((1/3)^*4+(2/3)^*1) + (2/5)^*1$ = 8/5
- So indifferent and willing to randomize (equilibrium)
- Questions
- Joint work with:







START EXIT Х -0 \mathbf{C} 0 Ν Т EXIT B 4 0 Ν Т

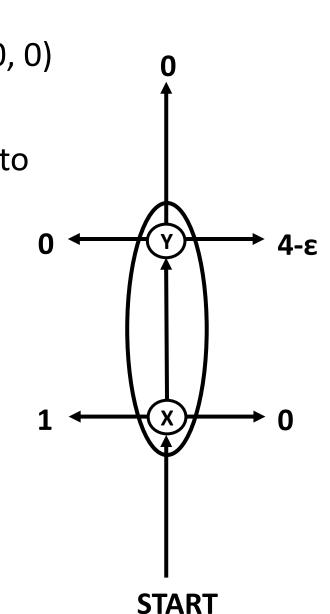
- Does this always work? Yes! (See also Taylor [2016])
- Does some version of EDT work with some version of belief formation?

FIG. 1. The absent-minded driver problem.

Image from Aumann, Hart, Perry 1997

A challenging example for the evidential decision theorist

- Optimal strategy to commit to is to just go left: $(p_1, p_s, p_r) = (1, 0, 0)$
- If you're at an intersection, what does EDT say you should do?
- When considering $(p_{|}, p_{s}, p_{r}) = (1, 0, 0)$, you presumably expect to be at X and get 1 (really just need: no more than 1)
- When considering (p₁, p_s, p_r) = (0, ½, ½), then say b is your subjective probability of being at Y
 - **Assume:** *b* > 0
 - Assume: *b* is not a function of ε
- So, expected utility: $b^{*}/_{2}^{*}(4-\epsilon) + (1-b)^{*}/_{4}^{*}(4-\epsilon) = 1+b-/_{4}\epsilon-/_{4}b\epsilon$
- For sufficiently small ϵ this is greater than 1
- Hence EDT suggests (0, ½, ½) over (1, 0, 0)!
- ... right? ... right?



A way for EDT to get the right answer (+SSA)

START

- Consider probabilities of whole trajectories, plus where you are, under strategy (0, ½, ½), in a halfing sort of way
- $P(XY(4-\epsilon), @X) = P(XY(4-\epsilon)) * P(@X|XY(4-\epsilon)) = \frac{1}{4} * \frac{1}{2}$
- $P(XY(4-\epsilon), @Y) = P(XY(4-\epsilon)) * P(@Y|XY(4-\epsilon)) = \frac{1}{4} * \frac{1}{2}$
- Any other trajectory with positive probability gives payoff 0
- So expected utility is 2 * ¼ * ½ * (4-ε) = 1- ε/4, which is worse than 1, so EDT gets the right answer
- What just happened?
- Under this way of reasoning, if you tell me that I'm at X, it's more likely that I'm on trajectory X(0) than on one of the XY ones
- P(XY(4-ε), @X) = ¼ * ½ ; P(XY(0), @X) = ¼ * ½ ; P(X(0), @X) = ½ * 1
- So $P(X(0) | @X) = \frac{1}{2} / (\frac{1}{2} + \frac{1}{4}) = \frac{2}{3} (not 1/2)$
- Previous slide had **hidden assumption**: where I am carries no information about my **future** coin tosses

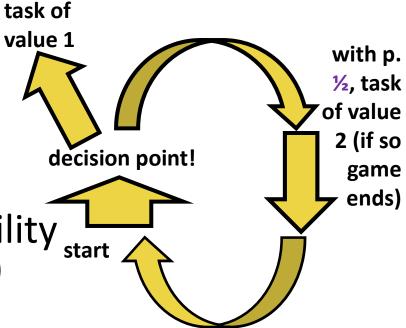
Making decisions with imperfect recall [cf. absentminded driver problem: PR97, AHP97]

- Optimal strategy without recall: go Right with probability 5/8. (Outside view.) Follow that.
- You arrive at decision point. What is the probability that you're there for the first time? (Inside view.)
- Thirder: in expectation 1 first awakening, and

 (1/2)(5/8)(16/25) / (1-(5/8)(16/25)) = 1/3 later
 awakenings, so probability of first time = 1/(4/3) = ³/₄
- Going Left gives 1 and going Right gives (1/2)(3/4)(2) + ((1/2)(3/4)+(1/4))(16/25)(3/8) / (1-(5/8)(16/25)) = 1
- Theorem. This is always true!
- ... but can have other equilibria



Scott Emmons Caspar Oesterheld Andrew Critch Stuart Russell



discount factor (probability that game continues): y = 16/25 = .64







Fraction of time replicator dynamic finds best solution

A N	2	3	4	5	A N	2	3	4	5
2	0.93	0.81	0.68	0.65	2	0.58	0.45	0.40	0.33
3	0.81	0.70	0.58	0.46	3	0.57	0.35	0.29	0.27
4	0.76	0.58	0.36	0.34	4	0.53	0.37	0.28	0.25
5	0.69	0.43	0.36	0.30	5	0.51	0.33	0.33	0.24

(a) RandomGame

(b) CoordinationGame

N = #players (or #nodes) A = #actions per player (or per node)

Functional Decision Theory

[Soares and Levinstein 2017; Yudkowsky and Soares 2017]

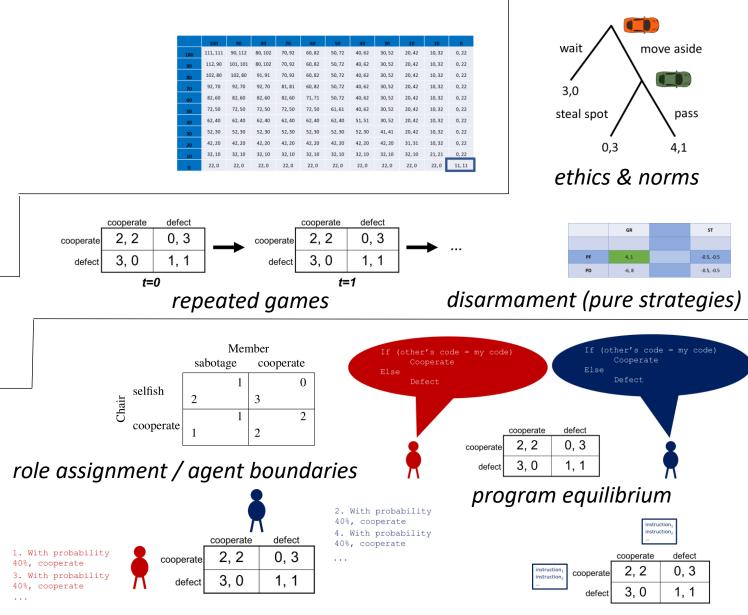
- One interpretation: act as you would have precommitted to act
- Avoids my EDT Dutch book (I think)
- ... still one-boxes in Newcomb's problem
- ... even one-boxes in Newcomb's problem with transparent boxes
- An odd example: Demon that will send you \$1,000 if it believes you would otherwise destroy everything (worth -\$1,000,000 to everyone)

Don't do it!

 FDT says you should destroy everything, even if you only find out that you are playing this game after the entity has already decided not to give you the money (too-late extortion?)

Summary of approach

- Game-theoretic failures to cooperate can happen even with almost perfectly aligned agents
- Some ways of getting to cooperation make sense for humans as well...
- ... but there are others that seem more natural for (advanced) AI agents
- Let's not unnecessarily limit our toolkit!



disarmament (mixed strategies)

philosophical foundations (evidential decision theory, selflocating belief, ...)

Outline

- Tragedies of algorithmic interaction examples and worries
- Rethinking the design of intelligent agents
 - (Intelligence + value alignment) still allows game-theoretic tragedies
- Should AI systems cooperate like humans do?
- Techniques for achieving cooperation that (also) fit humans
- Techniques for achieving cooperation that don't fit humans
- Open questions and call to action

Many open questions

- What are the foundations of game theory for highly advanced AI?
- How should an agent play with other agents with overlapping code? With visible code?
- How should an agent play when it may be being simulated? When it can't remember the past?
- What design decisions can improve cooperation?
 - How realistic are they? How do we make them more so?
 - How robust are they? How do we make them more so?
- What is the role of learning?
 - Can we design learning algorithms that converge to good equilibria?
 - In contexts of logical uncertainty?

THANK YOU FOR YOUR ATTENTION!