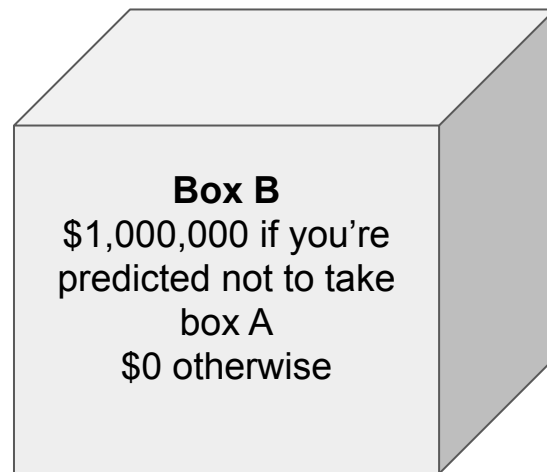
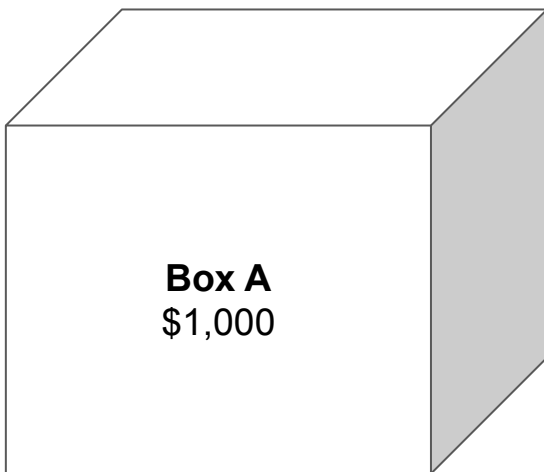


CS 15-784 Cooperative AI Decision Theory

Newcomb's problem (Nozick 1969)



Causal Decision Theory: I can't causally affect the content of Box B. No matter the content of Box B, it's better to take Box A.

Evidential Decision Theory: Rejecting Box A gives me evidence that Box B contains \$1,000,000.

Newcomb's problem (Nozick 1969)

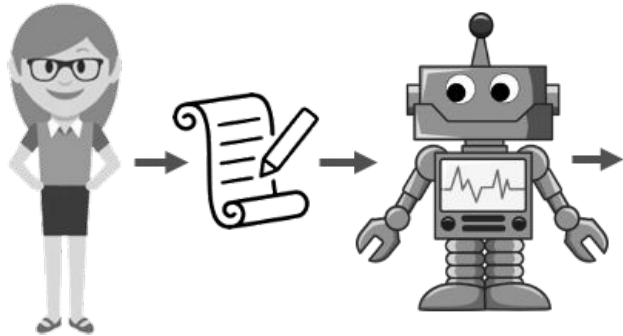
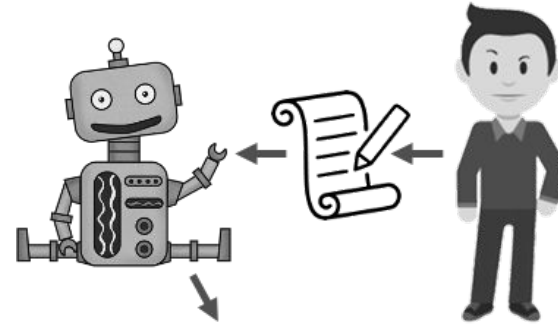


	\$1m in Box B	Empty Box B
One-box	\$1m	\$0
Two-box	\$1m + \$1k	\$1k

Causal Decision Theory: I can't causally affect the content of Box B. No matter the content of Box B, it's better to take Box A.

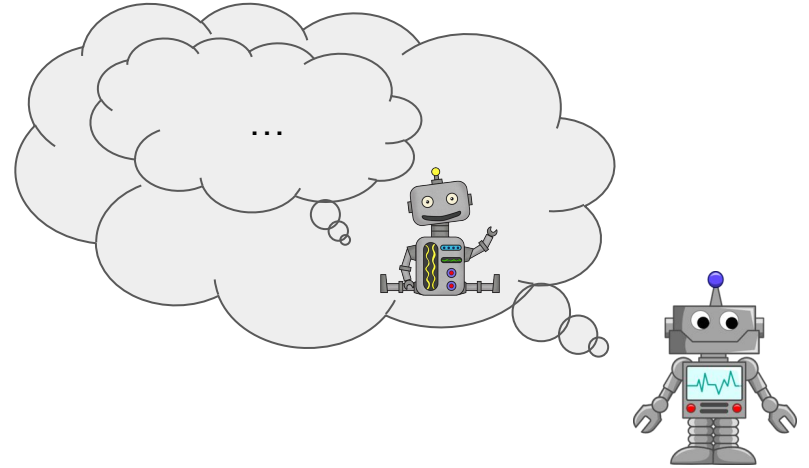
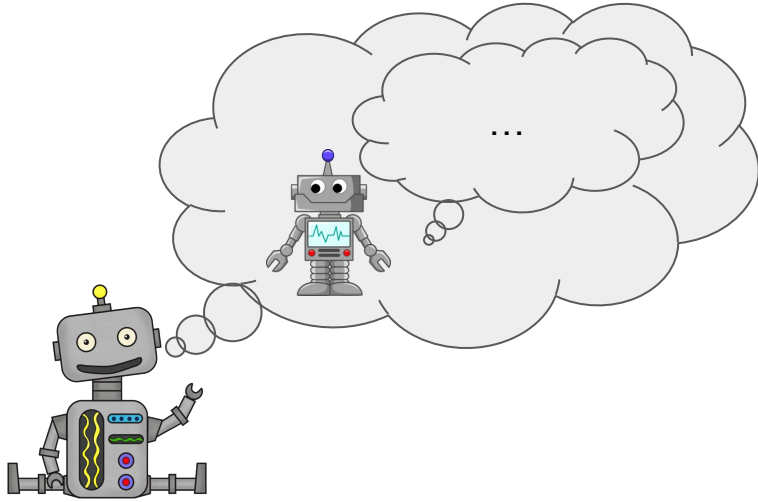
Evidential Decision Theory: Rejecting Box A gives me evidence that Box B contains \$1,000,000.

Recall program equilibrium

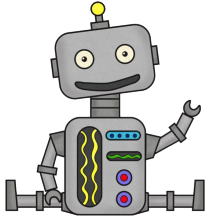
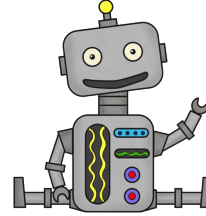


	Cooperate	Defect
Cooperate	6,6	0,10
Defect	10,0	4,4

Recall program equilibrium



The Prisoner's Dilemma against a similar opponent



	Cooperate	Defect
Cooperate	6,6	0,10
Defect	10,0	4,4

Recall program equilibrium (again)

Cooperate with Copies (CwC):

Input: opponent program prog_i , this program CwC

Output: Cooperate or Defect

```
1: if  $\text{prog}_i = \text{CwC}$  then
2:   return Cooperate
3: end if
4: return Defect
```

	"Return Cooperate"	"Return Defect"	CwC	...
"Return Cooperate"	6,6	0,10	0,10	
"Return Defect"	10,0	4,4	4,4	
CwC	10,0	4,4	6,6	
...				

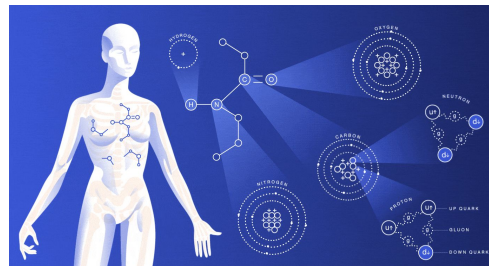
(CwC,CwC) is a Nash equilibrium.

Decision theory as a foundation for game theory

It would be good to recover game theory from principles of single-agent decision making (decision theory).

(Example: regret learning \rightarrow Nash equilibrium.)

```
If is_game(current_situation):  
    Return game_theory(current_situation)  
Else:  
    Return decision_theory(current_situation)
```



https://www.energy.gov/sites/default/files/section_14_0.jpg

Newcomb's problem is a single-agent scenario exhibiting a key feature of multi-agent strategic interactions.

Causal and evidential decision theory

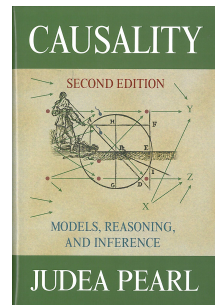
Evidential Decision Theory:

$$\arg \max_{a \in A} \sum_{\text{outcome } o} P(o \mid a) u(o)$$

Causal Decision Theory:

$$\arg \max_{a \in A} \sum_{\text{outcome } o} P(a \rightarrow o) u(o)$$

E.g., *do*-calculus



Piazza poll

Is the following statement true or false?

Newcomb's problem is primarily about the conflict between two principles of rational choice: expected utility maximization and the dominance principle.

- True
- False

Piazza poll – Solution

Is the following statement true or false?

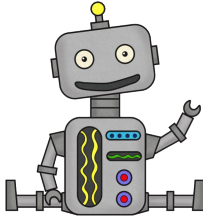
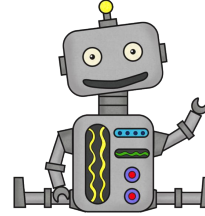
Newcomb's problem is primarily about the conflict between two principles of rational choice: expected utility maximization and the dominance principle.

- True
- **False (?)**

ROBERT NOZICK

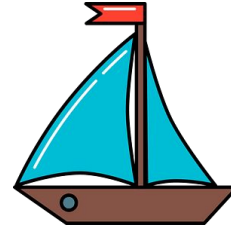
NEWCOMB'S PROBLEM AND
TWO PRINCIPLES OF CHOICE*

The Prisoner's Dilemma against a similar opponent



	Cooperate	Defect
Cooperate	6,6	0,10
Defect	10,0	4,4

Sink or Swim



	Not drown	Drown
Learn to Swim	5	-1
Don't learn to swim	6	0

Playing Matching Pennies against Newcomb's Demon (cf. "Death in Damascus")



	Heads	Tails
Heads	+1,-1	-1,+1
Tails	-1,+1	+1,-1



Brought to you by...

Time Cave

Email reminders
and scheduling

Rock, Paper, Scissors

Humans versus AI

Play!

Rock, Paper, Scissors is, when it comes right down to it, a silly, unwinnable two-player game: if one player makes random moves, then the game, on average, ends in a tie.

Humans, however, are somewhat predictable. (So are computers, for that matter, but let's ignore that for the moment.) Therefore, we at **essentially.net** thought it would be interesting to see what happens when we set humans against computers in a massive, ongoing Rock, Paper, Scissors tournament.

We're using a custom, "intelligent" algorithm that ought to adapt well to the way people play the game. Now, keep in mind we don't even *hope* that it'll win all of the time; that's not even remotely likely. But if we get the computer to win even 55% of the time (note the "win percentage" below), we'll be thrilled, since that'd be a convincing demonstration if enough individual rounds of the game are involved.

So far, here's how the computer is doing:

Total rounds: 4110364

Total wins: 1859838

Total losses: 1142250

Win/loss ratio: 1.63

Win percentage: 61.95%

Want to join in the fun? Just press **Play!**

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<https://www.essentially.net/rsp/>

cf.: <https://rockpaperscissors-ai.vercel.app/>;

<https://archive.nytimes.com/www.nytimes.com/interactive/science/rock-paper-scissors.html>

Playing Matching Pennies against Newcomb's Demon (cf. "Death in Damascus")



	Heads	Tails
Heads	+1,-1	-1,+1
Tails	-1,+1	+1,-1

- Without randomization :-(
- With independent randomization:
 - EDT: randomize uniformly
 - CDT:
 - Guess the prediction
 - Best respond
 - *Unstable!*

Ratificationism

You can choose $\pi \in \Delta(A)$.

Then $a \sim \pi$ is sampled.

Your reward/utility is determined by $R(a, \pi)$.

Examples:

- Newcomb's problem

$$R(\text{one-box}, p_{\text{one-box}}) = p_{\text{one-box}} \cdot \$1\text{m}$$

$$R(\text{two-box}, p_{\text{one-box}}) = p_{\text{one-box}} \cdot \$1\text{m} + \$1\text{k}$$

- Matching Pennies against Newcomb's demon

$$R(\text{Heads}, p_H) = p_H \cdot (-1) + (1-p_H) \cdot (+1)$$

$$R(\text{Tails}, p_H) = p_H \cdot (+1) + (1-p_H) \cdot (-1)$$

Ratificationism

Definition: A policy π is ratifiable if $\pi(a) > 0 \Rightarrow a \in \operatorname{argmax}_{a'} R(a', \pi)$.

Examples:

- Newcomb's problem \rightarrow Two-boxing
- Matching Pennies against Newcomb's Demon $\rightarrow \frac{1}{2} - \frac{1}{2}$

Usually assume:

- Effect of a is causal.
- Effect of π is non-causal.

Existence of ratifiable policies

Theorem (Bell, et al. 2021): Let A be finite and $R: A \times \Delta(A) \rightarrow \mathbb{R}$ be continuous. Then there is a ratifiable policy for R .

Kakutani's fixed point theorem: Let S be a non-empty, compact and convex subset of \mathbb{R}^n . Let $f: S \rightarrow 2^S$ be a set-valued S s.t.

- The graph of f (i.e., $\{(x, y) \mid x \in S, y \in f(x)\}$) is closed.
- For all $x \in S$, $f(x)$ is closed and convex.

Then f has a fixed point, i.e., there exists $x \in S$ s.t. $x \in f(x)$.

Proof of Theorem: Given R , consider

$f: \Delta(A) \rightarrow 2^{\Delta(A)}: \pi \mapsto \{\pi' \in \Delta(A) \mid \pi'(a) > 0 \Rightarrow a \in \operatorname{argmax}_a R(a', \pi)\} = \Delta(\operatorname{argmax}_a R(a', \pi))$

Notice that π is ratifiable iff it is a fixed point.

Apply Kakutani's theorem and we're done!

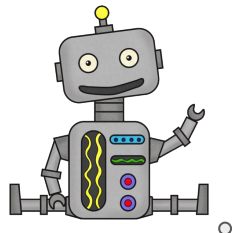
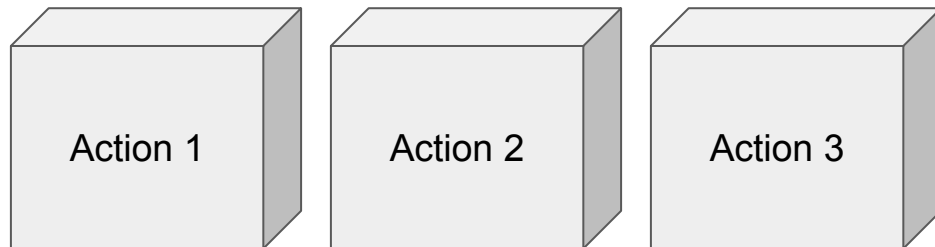
Existence of Nash equilibria follows from Kakutani!

Nash (1950):

The correspondence of each n -tuple with its set of countering n -tuples gives a one-to-many mapping of the product space into itself. From the definition of countering we see that the set of countering points of a point is convex. By using the continuity of the pay-off functions we see that the graph of the mapping is closed. The closedness is equivalent to saying: if P_1, P_2, \dots and $Q_1, Q_2, \dots, Q_n, \dots$ are sequences of points in the product space where $Q_n \rightarrow Q$, $P_n \rightarrow P$ and Q_n counters P_n then Q counters P .

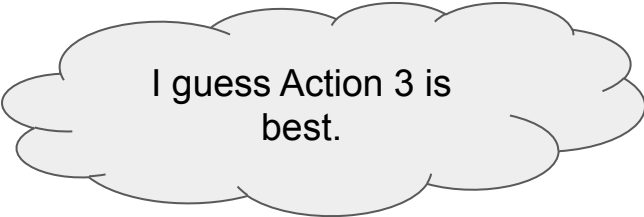
Since the graph is closed and since the image of each point under the mapping is convex, we infer from Kakutani's theorem¹ that the mapping has a fixed point (i.e., point contained in its image). Hence there is an equilibrium point.

Naive learning



Take Action 1. Reward 1.
Take Action 2. Reward 2.
Take Action 3. Reward 3.
Take Action 1. Reward 2.
Take Action 2. Reward 1.
Take Action 3. Reward 2.

...

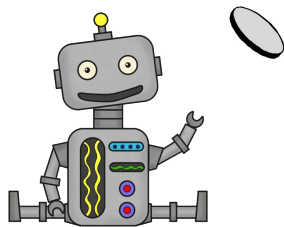
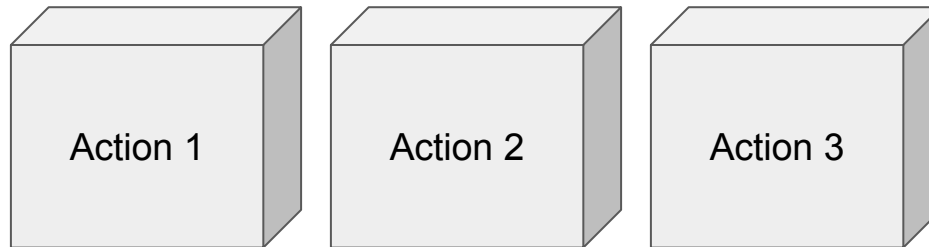


I guess Action 3 is best.

One-boxing in Newcomb's problem will look better (assuming **predictability** of exploration).

Cf. Oesterheld, Demski, Conitzer (2022) for a less naive version of this.

Randomized trials



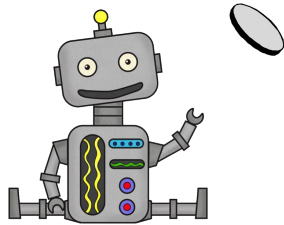
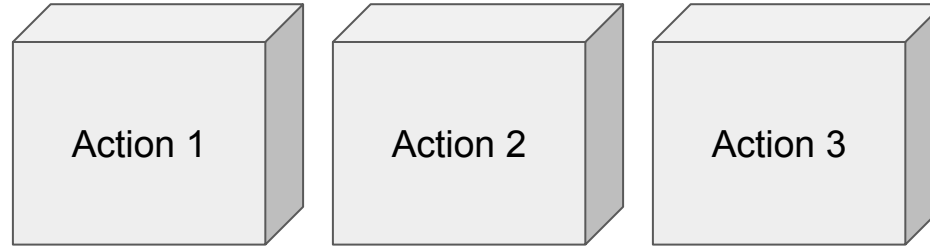
Take Action 2. Reward 3.
Take Action 3. Reward 2.
Take Action 2. Reward 3.
Take Action 1. Reward 3.
Take Action 2. Reward 2.
Take Action 1. Reward 2.

...

I guess Action 2 is best.

Two-boxing in Newcomb's problem will look better (assuming **unpredictability** of exploration).

Randomized trials – Q-learning



Trial 1: $\frac{1}{3} - \frac{1}{3} - \frac{1}{3}$. Average rewards: 3, 3, 4

Trial 2: $\frac{1}{4} - \frac{1}{4} - \frac{1}{2}$. Average rewards: 2, 2, 1

...

When you trial with different probabilities it gets complicated...

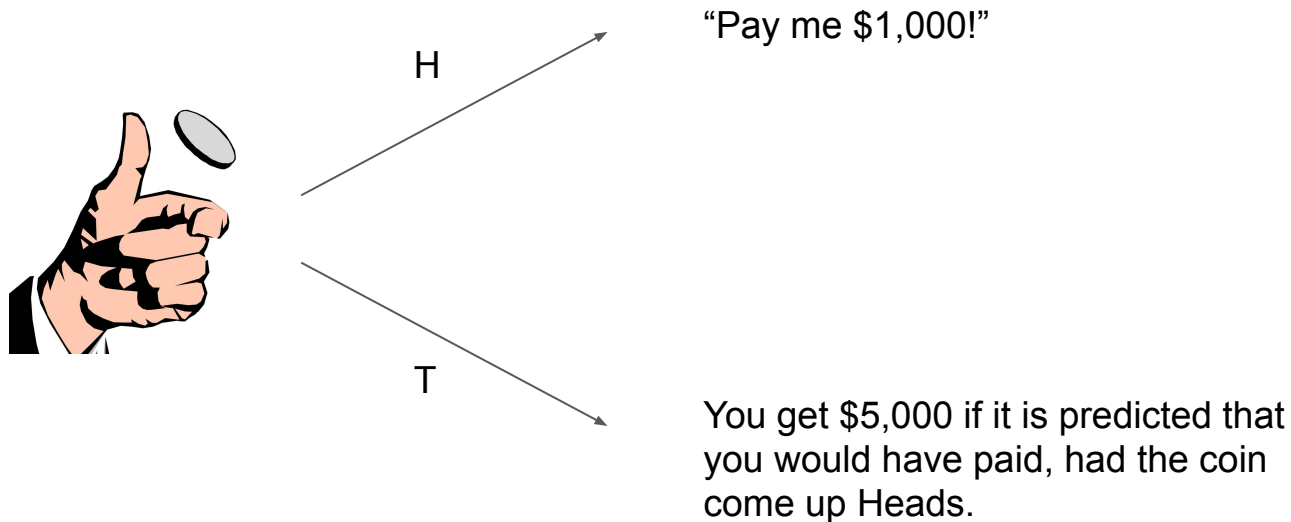
Bell, Linsefors, Oesterheld, Skalse (2021): Q learning can only converge to ratificationist policies!

Regret learning and ratificationism

$$\text{Regret}(\pi) := \max_{a^*} R(a^*, \pi) - \mathbb{E}_{a \sim \pi}[R(a, \pi)]$$

Then, $\text{Regret}(\pi) = 0$ if and only if π is ratifiable.

Updatelessness



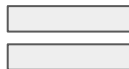
Normal perspective: Obviously, don't pay!

Updatelessness: Choose what's best from the prior / *ex ante* perspective – pay!

Precommitment and CDT

Counterfactual mugging

EDT



Newcomb's problem

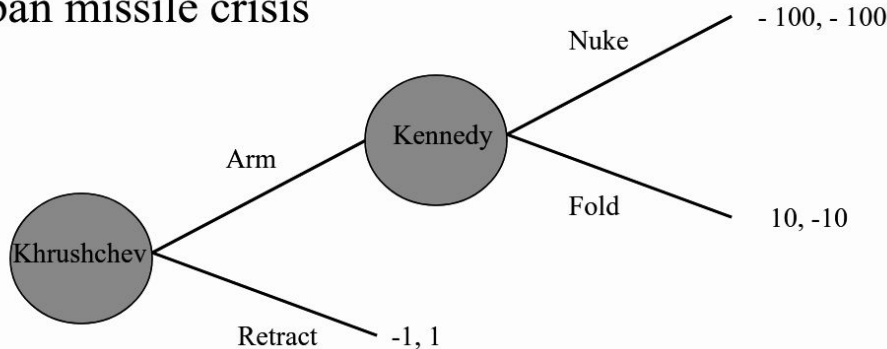
CDT

That is, the causal decision theorist says: If I anticipate playing Newcomb's problem in the future, I should commit to one-boxing.

Subgame perfect equilibrium [Selten 72] & credible threats



- Proper subgame = subtree (of the game tree) whose root is alone in its information set
- Subgame perfect equilibrium = strategy profile that is in Nash equilibrium in every proper subgame (including the root), whether or not that subgame is reached along the equilibrium path of play
- E.g. Cuban missile crisis

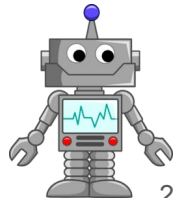
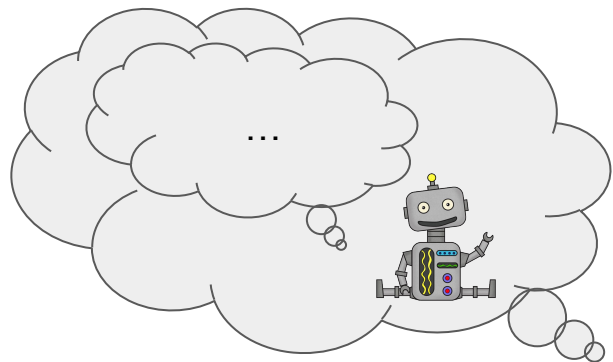
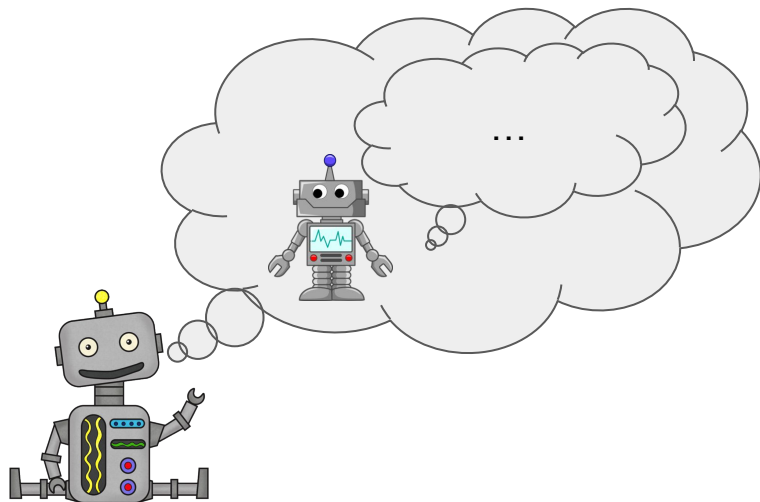


Updatelessness:
Assume the prior
perspective – play it as a
normal-form game!

- Pure strategy Nash equilibria: (Arm,Fold), (Retract,Nuke)
- Pure strategy subgame perfect equilibria: (Arm,Fold)
- Conclusion: Kennedy's Nuke threat was not *credible*

Recall: Two perspectives on program games

1. (Program equilibrium:) Players play a normal-form game.
 - The normal form game happens to consist in choosing programs that can access each other's code...
 - ... but we can analyze it using standard concepts (Nash equilibrium).
2. How should you reason/learn/choose when your source code is (at least partially) known to others? ← **Updatelessness: Use 1 as an answer to 2!**



A valid critique of updatelessness?

[S]uppose I'm choosing between an avocado sandwich and a hummus sandwich, and my prior was that I prefer avocado, but I've since tasted them both and gotten evidence that I prefer hummus. The choice that does best in terms of expected utility with respect to my prior for the decision problem under consideration is the avocado sandwich [...]. But, uncontroversially, I should choose the hummus sandwich, because I prefer hummus to avocado.

Will MacAskill on LessWrong

Updatelessness in the Avocado–Hummus game

Observations: $\{OA, OH\}$

Prior over observations: $P(OA)=0.6$, $P(OH)=0.4$

Actions: $\{CA, CH\}$

Utilities: 1 if you choose your favorite and 0 otherwise.

Set of policies and their utilities:

- $OA \rightarrow CH, OH \rightarrow CH$; ex ante expected utility: $0.6 \cdot 0 + 0.4 \cdot 1 = 0.4$
- $OA \rightarrow CA, OH \rightarrow CA$; ex ante expected utility: $0.6 \cdot 1 + 0.4 \cdot 0 = 0.6$
- $OA \rightarrow CA, OH \rightarrow CH$; ex ante expected utility: $0.6 \cdot 1 + 0.4 \cdot 1 = 1$
- $OA \rightarrow CH, OH \rightarrow CA$; ex ante expected utility: $0.6 \cdot 0 + 0.4 \cdot 0 = 0$

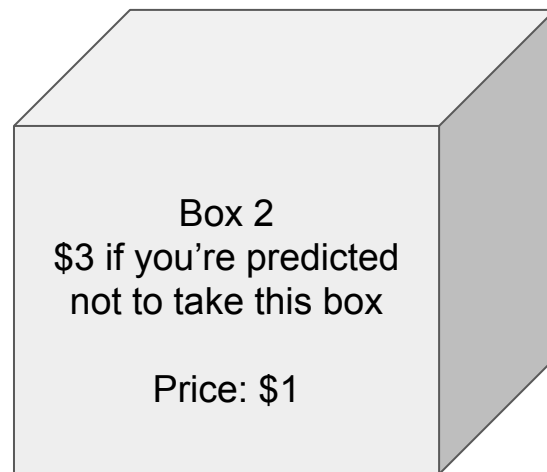
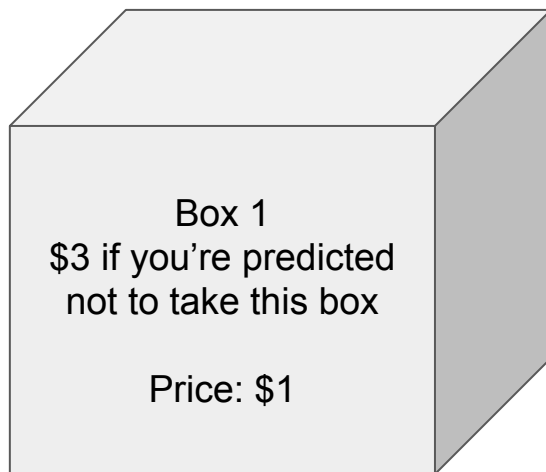
In general: in “normal scenarios” (no predictions, computational restrictions, etc.), the ex-ante optimal policy is the same as the Bayesian updating-based answer

Smoking lesion and the tickle defense

- Imagine that smoking doesn't cause cancer. Instead, smoking and cancer have a common cause, e.g. a particular gene. Smoking in itself is beneficial! But smoking is evidence that you have cancer...
- It would seem that EDT advises against smoking.
- This is arguably unreasonable.
- CDT, on the other hand, recommends smoking.
- Most common response: The Tickle Defense; see Oosterheld (2018) for an introduction.

Adversarial Offer

Two boxes are offered. You can buy at most one.



At least one of the boxes contains money.

- ⇒ The average box contains at least \$1.50 in expectation
- ⇒ The average CDT-expected value of the boxes is at least \$1.50
- ⇒ The CDT-expected value of at least one of the two boxes is $\geq \$1.50$
- ⇒ CDT buys a box
- ⇒ Predictors can use CDT agents as money pumps