

# A game theoretic look at alignment

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# Game Theory: Interacting Decision Makers

Game theory is about interactive decision making:

- ▶ It has very little to do with Chess and checkers!
- ▶ But lots to do with:
  - ▶ evolution
  - ▶ knowledge
  - ▶ manipulation
  - ▶ deception
  - ▶ reputation
  - ▶ trust
  - ▶ reputation
  - ▶ communication
- ▶ All ripe areas for modeling alignment

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- ▶ All ripe areas for modeling alignment

I'll take questions until slide 21!

# Connection to security

Many similarities with security:

- ▶ Randomization:
  - ▶ games: necessary for games to protect private knowledge
  - ▶ CS: necessary for interactive proofs and zero knowledge proofs
- ▶ Chains of reputation:
  - ▶ games: Useful for identifying bad actors
  - ▶ CS: “web of trust”
- ▶ Openness is better:
  - ▶ games: mechanism design
  - ▶ CS: security through obscurity isn't secure

# Trust

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  - ▶ The company trusts the executive with the power to buy start-ups
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  - ▶ The company trusts the executive with the power to buy start-ups
  - ▶ But the company gives them zero training
- ▶ The company doesn't trust the executive to log into their email!
  - ▶ They need two factor authentication to log in
  - ▶ Two factors aren't necessary to buy a startup!

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- ▶ After it happened, I decided I was comfortable being a schmuck since the alternative was to trust less
- ▶ So knowing when human's will stupidly "trust" is an issue for alignment

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- ▶ So knowing when human's will stupidly "trust" is an issue for alignment
- ▶ (If Chimps ruled the world, we wouldn't have to worry about alignment—they trust no-one!)

## The company's policy makes sense

- ▶ The company knows the executive is susceptible to spear phishing
  - ▶ So they lock that door twice!
- ▶ They know the executive won't trust a valuation of a start up as being a "good deal"
  - ▶ So they don't even lock that door once

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# Information vs. computation

- ▶ In game theory, all true facts are common knowledge
- ▶ We will model computation as information

# The Principal / Agent problem



## Evolution

Oldest example of Principal / agent:

- Flowers and bees
- Flowers "pay" bees to pollinate for them
- Flower is principal
- Bees is agent
- The deal
  - Flowers in winter
    - Food left in advance and half afterwards
    - Signable payment based on number of bees in the market place
    - Successful arrangement for 100 million years
- Note: Bees are much smarter than flowers

## Farming: Share cropping

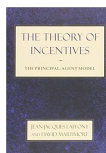
- Principal: Land owner
- Agent: Farmer
- The deal:

class 16

## Theory: Agents have knowledge

- Agents know more than principals
  - Necessary for game theory model
  - Otherwise, principal can simply pay "piece work"
- We will be modeling super-AIs as more knowledgeable
  - knowledge in game theory is sigma-fields, observations from the world, knowledge of ones personal utility function, etc
  - None of these apply to an AI
  - But they are better at computation
  - Which looks a lot like information
  - We will take it as being the same

## Books:



AGENCY COST

exploiting an information shadow for self-interest



AGENT

hires

performs



PRINCIPAL



S. ADAMS E-mail: SCOTTADAMS@aol.com



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# Evolution

Oldest example of Principal / agent:

- ▶ Flowers and bees!
- ▶ Flowers “pay” bees to pollinate for them
- ▶ Flower is principal
- ▶ Bee is agent
- ▶ The deal:
  - ▶ Payment in nectar
  - ▶ Paid half in advance and half afterwards
  - ▶ Variable payment based on number of bees in the market place
  - ▶ Successful arrangement for 100 million years
- ▶ Note: Bees are *much* smarter than flowers

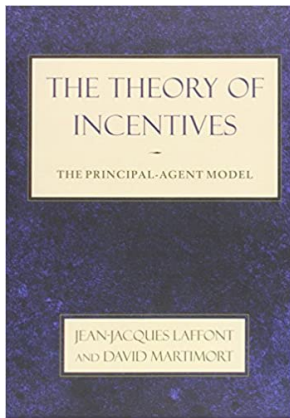
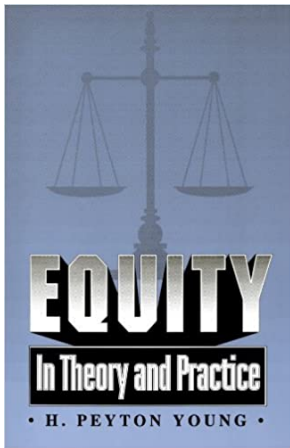
# Farming: Share cropping

- ▶ Principal: Land owner
- ▶ Agent: Farmer
- ▶ The deal:
  - ▶ Farmers give half of the proceeds to owner
  - ▶ Owner doesn't know how much productivity is due to effort vs luck
  - ▶ 50 / 50 split is common, but other splits are possible
- ▶ Note: Owners don't have to know farming

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# Books:



# Game Theory Questions?

## TRUST

<p><b>Connection to security</b></p> <ul style="list-style-type: none"> <li>Many conditions with security             <ul style="list-style-type: none"> <li>• Authentication</li> <li>• Given authority to give to prevent private knowledge</li> <li>• To respond to requests for private and non-proprietary assets</li> <li>• Check for authentication</li> <li>• To provide feedback for identifying bad actors</li> <li>• To track bad actor</li> <li>• Exposure to failure</li> <li>• To provide immediate danger</li> <li>• To security through identity, not to access</li> </ul> </li> </ul>	<p><b>Trust</b></p> <ul style="list-style-type: none"> <li>Consider an "ensemble" of a network             <ul style="list-style-type: none"> <li>• The context means the connection with the power to buy</li> <li>• But the company gives them some warning</li> </ul> </li> </ul>
<p><b>Maximum trust (not trust)</b></p> <p>A few years ago I got scammed on the street by being told a wife was</p>	<p><b>The company's policy makes sense</b></p> <ul style="list-style-type: none"> <li>The company knows this situation is susceptible to spoofing             <ul style="list-style-type: none"> <li>• To they look that bad actor</li> </ul> </li> <li>They know the situation and it's not a violation of a smart up thinking, good deal             <ul style="list-style-type: none"> <li>• To they don't even look that bad actor</li> </ul> </li> </ul>

## Information vs. computation

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- ▶ We will model computation as information

## The Principal / Agent problem



**Definition**

The principal-agent problem is a type of economic problem that arises when one party (the principal) hires another party (the agent) to perform a task on their behalf. The principal-agent problem arises because the principal and the agent have different interests and the principal cannot perfectly monitor the agent's actions.

**Forming, there copying**

### Theory: Agents have knowledge

- Agents know more than principals
  - Incentives to give them more
  - Otherwise, principal can't do anything
- We will be modeling agents with the same knowledge
  - Incentives to give them more than principals
  - The ability to monitor the agent's actions
  - The ability to give them more than principals
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### Books:

That was slide 21!

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## **Policy suggestions:**

- ▶ Launch early
- ▶ Launch many
- ▶ Private AIs are unregulated (e.g. tutors / advobots)
- ▶ Public AIs:
  - ▶ log all their statements (block-chain AI?)
  - ▶ AIs are tiered / cross checked

## Launching early: Trust

Humans need to learn lack of trust:

- ▶ 1890's yellow journalism (modern tabloids)
- ▶ 1950's chain letters and mail fraud
- ▶ 1990's email chain letters (lead to snoops)
- ▶ 2010's Facebook for "real news"



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- ▶ 2020's AI

So launching earlier will allow humans to get used to them

## Pox parties

- ▶ We need to throw chicken pox parties!
  - ▶ These were common when I was a kid
  - ▶ We'd go to a sick child's house and hopefully get chicken pox
  - ▶ Hopefully no one under 30 has a clue what I'm talking about
  - ▶ (Vaccine came out in 1995)

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  - ▶ (Vaccine came out in 1995)
- ▶ We have no vaccine against evil AIs
- ▶ We need to get inoculated by exposure to real AIs
- ▶ Hopefully we can build up immunity as we progress from GPT4, 5, 6, ...

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  - ▶ Neither are most animals or humans
  - ▶ We learn from experience
  - ▶ Use that for future interactions

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- ▶ I'm not a real game theorist!
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  - ▶ We learn from experience
  - ▶ Use that for future interactions
- ▶ But, won't super smart AIs learn faster than humans if we have repeated interactions?

## Aside: Repeated games

- ▶ If a  $FSA(n)$  plays a  $FSA(2^n)$  it loses.<sup>1</sup>

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## Aside: Repeated games

- ▶ If a  $FSA(n)$  plays a  $FSA(2^n)$  it loses.<sup>1</sup>
- ▶ But, if a  $FSA(O(1))$  is allowed to toss a coin, then it plays well against an arbitrarily smart adversary.

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- ▶ But, if a  $FSA(O(1))$  is allowed to toss a coin, then it plays well against an arbitrarily smart adversary.
- ▶ This is true, even if the stupid FSA has to learn the correct strategy to play. (F. and Vohra 1998, F. and Kakade 2008)

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# Launch early

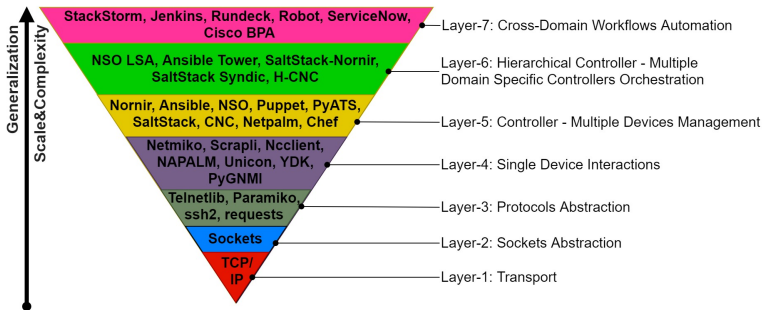
Launching early is a win because we:

- ▶ learn appropriate trust
- ▶ builds immunity
- ▶ learning doesn't favor the more intelligent

Principal / Agent



## Network Automation Abstraction Layers Taxonomy



# GPT4 as middle manager

- ▶ GPT4 can understand GPT5
  - ▶ Model GPT4 as having more information than we humans have
  - ▶ Use  $\sigma$ -fields
- ▶ Humans can understand GPT4
  - ▶ align GPT4's goals with human goals
  - ▶ Let GPT4 figure out how to align GPT5
- ▶ No trust is needed!

## Mathematics

<ul style="list-style-type: none"><li>▶ Human's <math>\sigma</math>-field is <math>\mathcal{F}_0</math>.</li><li>▶ GPT4's <math>\sigma</math>-field is <math>\mathcal{F}_4</math>.</li><li>▶ GPT5's <math>\sigma</math>-field is <math>\mathcal{F}_5</math>.</li><li>▶ GPT5 knows more than GPT4 which knows more than the human: <math>\mathcal{F}_0 \subset \mathcal{F}_4 \subset \mathcal{F}_5</math></li></ul>	<ul style="list-style-type: none"><li>▶ <math>A_0 \in \mathcal{F}_0</math>.</li><li>▶ <math>A_4 \in \mathcal{F}_4</math>.</li><li>▶ <math>A_5 \in \mathcal{F}_5</math>.</li><li>▶ <math>E((A_1) \mathcal{F}_0) \in \mathcal{F}_0</math>.</li><li>▶ Exotic Assumptions:<ul style="list-style-type: none"><li>▶ <math>E((A_1 \mathcal{A}) \mathcal{F}_4) \in \mathcal{F}_0</math>.</li><li>▶ <math>E((A_1 \mathcal{A}) \mathcal{F}_4) \in \mathcal{F}_4</math>.</li></ul></li></ul>
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### Theorem

*In this middle management principal agent model, the human's goals are aligned with GPT5's goals.*

Launching many: So they can control each other

# Many player games are easy

- ▶ Multiplayer games don't require as much strategic thinking
- ▶ An “economy of agents” is easier than a single agent



# Many player games are easy

- ▶ Multiplayer games don't require as much strategic thinking
- ▶ An “economy of agents” is easier than a single agent
- ▶ So, having many AIs is better than having a few
- ▶ Again: launch many!

# Launch many

Launching many is a win because:

- ▶ middle management / indirection
- ▶ economy requires less strategy than game theory

# Pseudo randomization

- ▶ Stackelberg equilibrium
- ▶ Example: Amazon vs FBA sellers
  - ▶ Each seller acts like a “random draw”
  - ▶ Amazon has to have a single policy for all sellers
- ▶ One AI against many people
  - ▶ pre-commit to what it is saying
  - ▶ Force it to tell a consistent story
  - ▶ Logging its statements
- ▶ TCS version: PCP

# Putting this together

- ▶ Launch early:
  - ▶ trust / reputation
  - ▶ builds immunity
  - ▶ learning
- ▶ Launch many:
  - ▶ economies are simpler than games (MIPs)
  - ▶ middle management
- ▶ Personalized private copies:
  - ▶ force privacy to avoid collusion
- ▶ large LLMs log their statements:
  - ▶ Stackelberg equilibrium (PCP)

# Final thoughts

- ▶ Game theory is useful model of human / AI interactions
  - ▶ Evolution has been solving these problems for billions of years
  - ▶ Humans have been solving them for millions of years
  - ▶ Legal codes have been solving them for 1000s of years
  - ▶ We can use this accumulated knowledge for alignment

THANKS!



# THANKS!

## Game Theory Questions?

**TRUST**

Trust	No Trust
Trust	No Trust

Source: [https://www.youtube.com/watch?v=...](#)

The Principal / Agent problem



**TRUST**

<b>Trustworthy Trust</b>	<b>Untrustworthy</b>
<b>Trustworthy</b>	<b>Untrustworthy</b>

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