# Automated Negotiation in Supply Chain Management

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Presentation at Rollins



- Ph.D. candidate, Computer Science, Brown University.
  - Research Area: Artificial Intelligence (AI) / Multi-Agent Systems.

Game Theory / Mechanism Design / Automated Negotiation.

- Originally from Caracas, Venezuela; in the U.S. since 2011.
- Prior to Brown, CS M.Sc. & MAT Mathematics at Indiana University.

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## Please feel free to interrupt me! Let's make this talk a conversation!



#### Part 1: Supply Chains

Part 2: Artificial Negotiation Agents for Supply Chains

**Part 3:** Future Research Plans



Part 1: Supply Chains

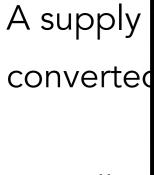
# Why are supply chains important? How do they function? How could they function better?

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# Supply Chains



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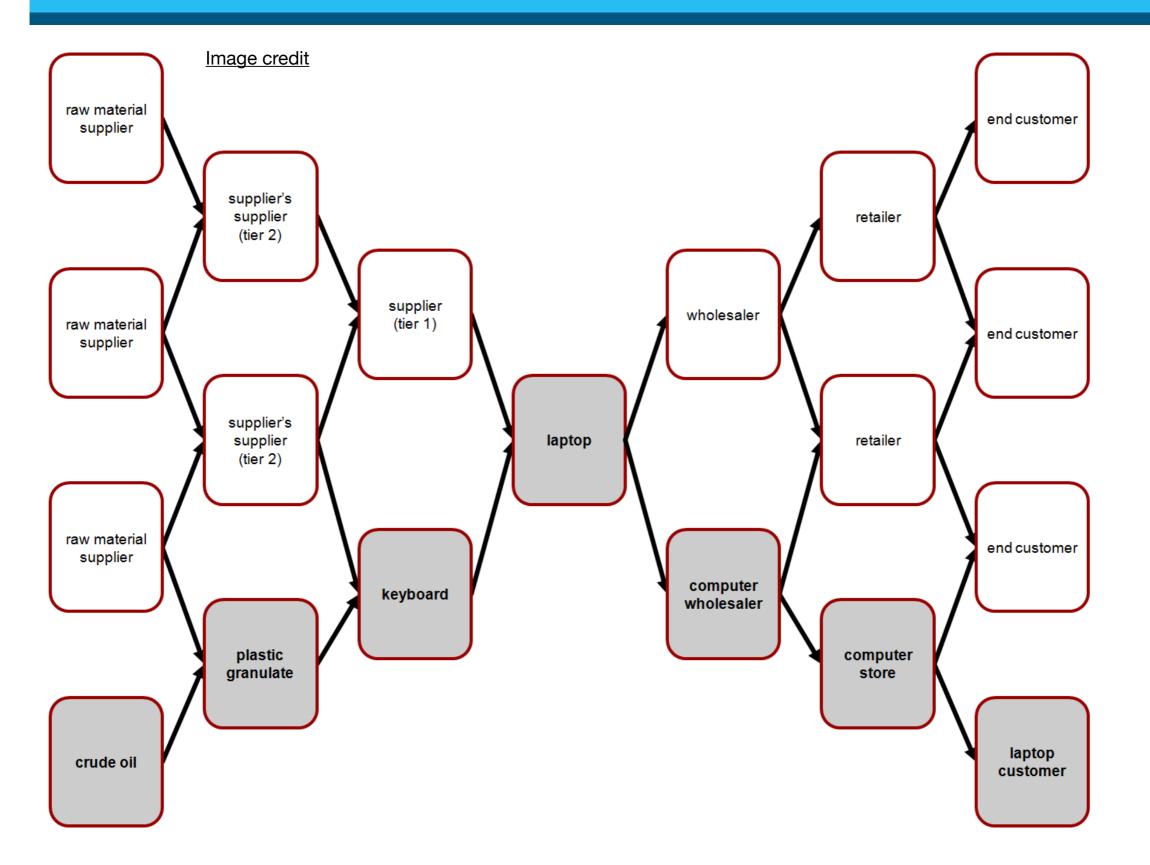
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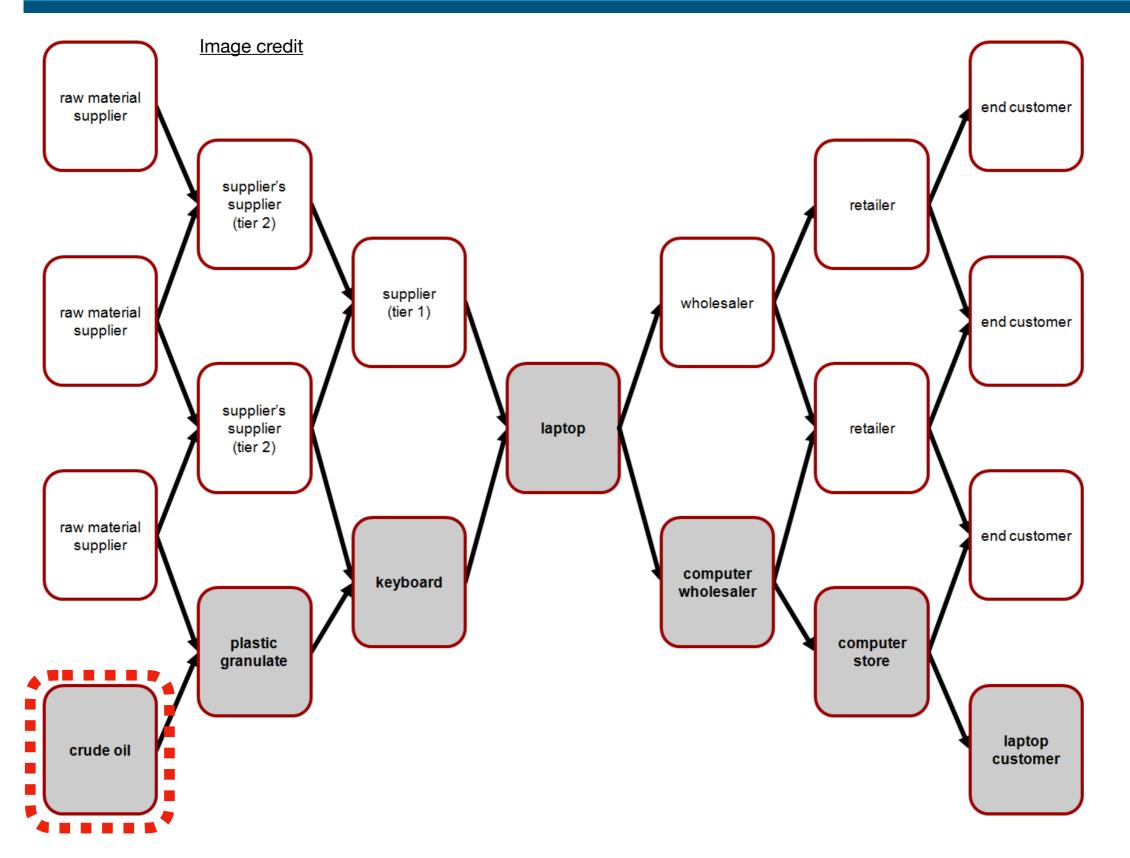
Participation in supply chains has contributed to **industrialisation** and high rates of **economic growth** in several developing economies (WTO).

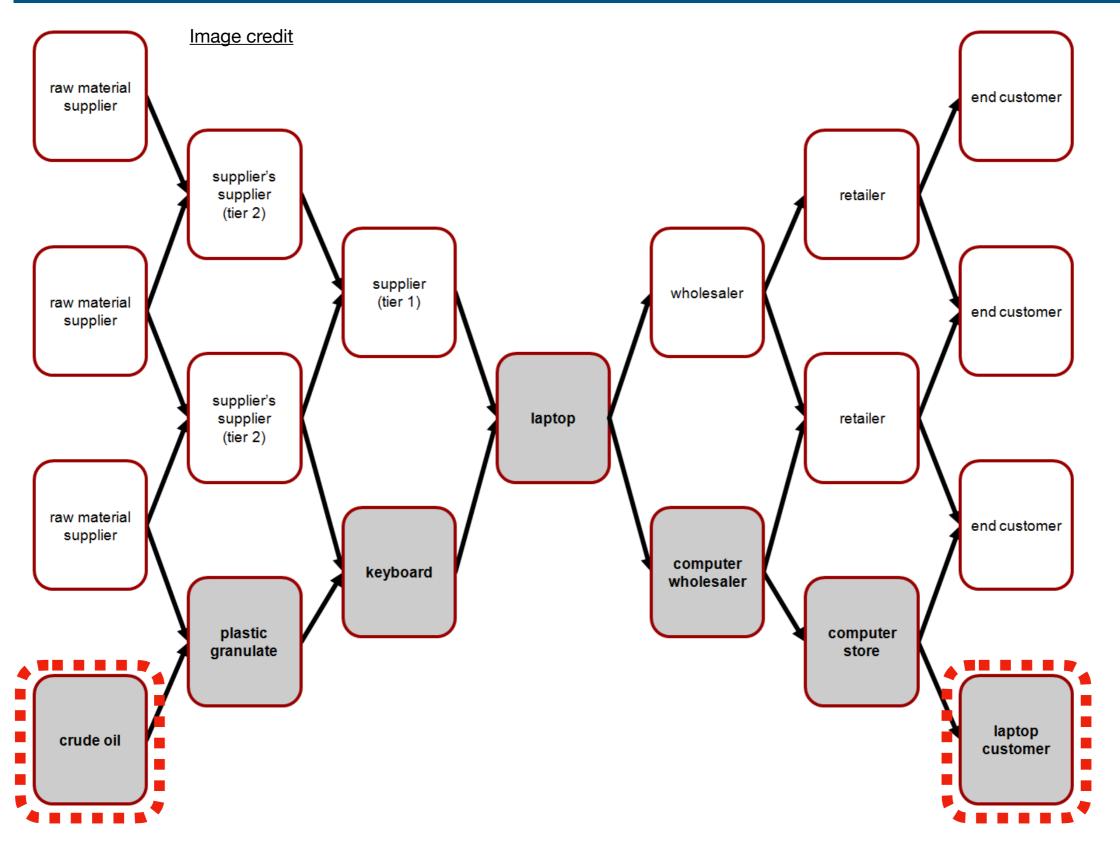
#### Supply chains are **decentralized**, i.e., no central coordinator.

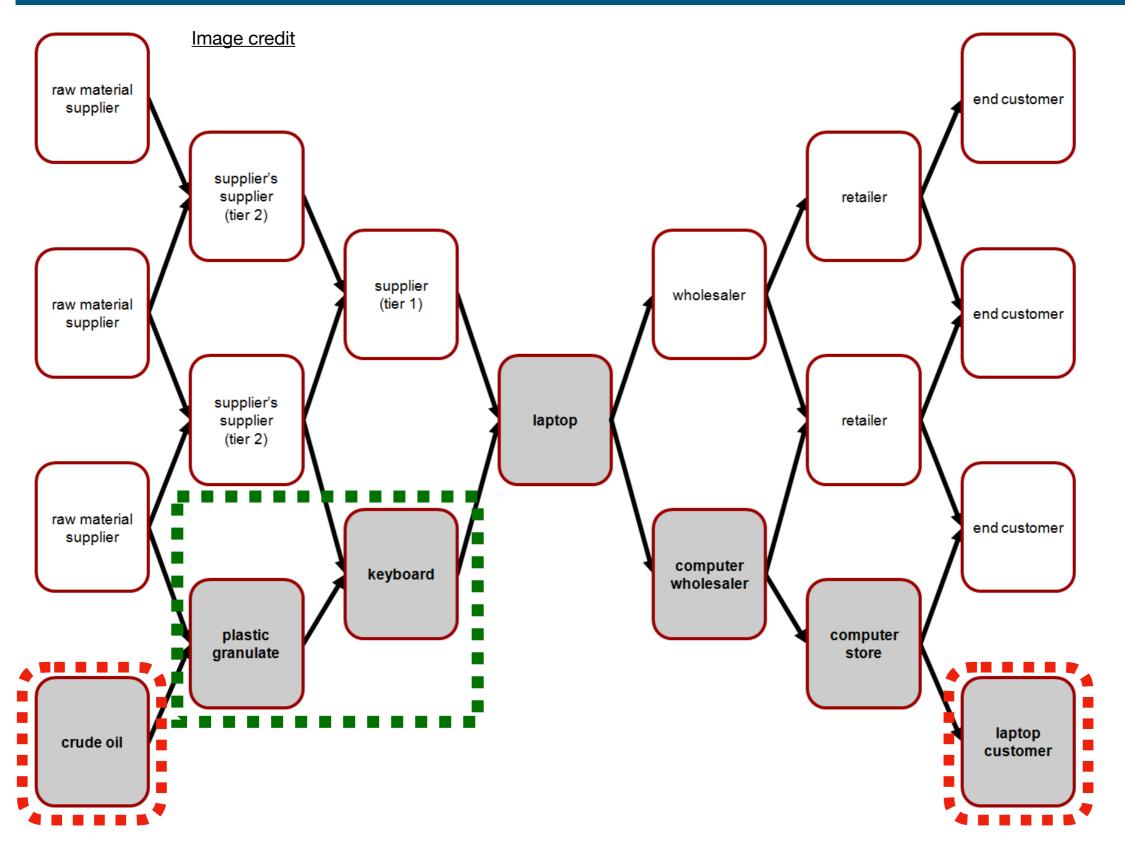
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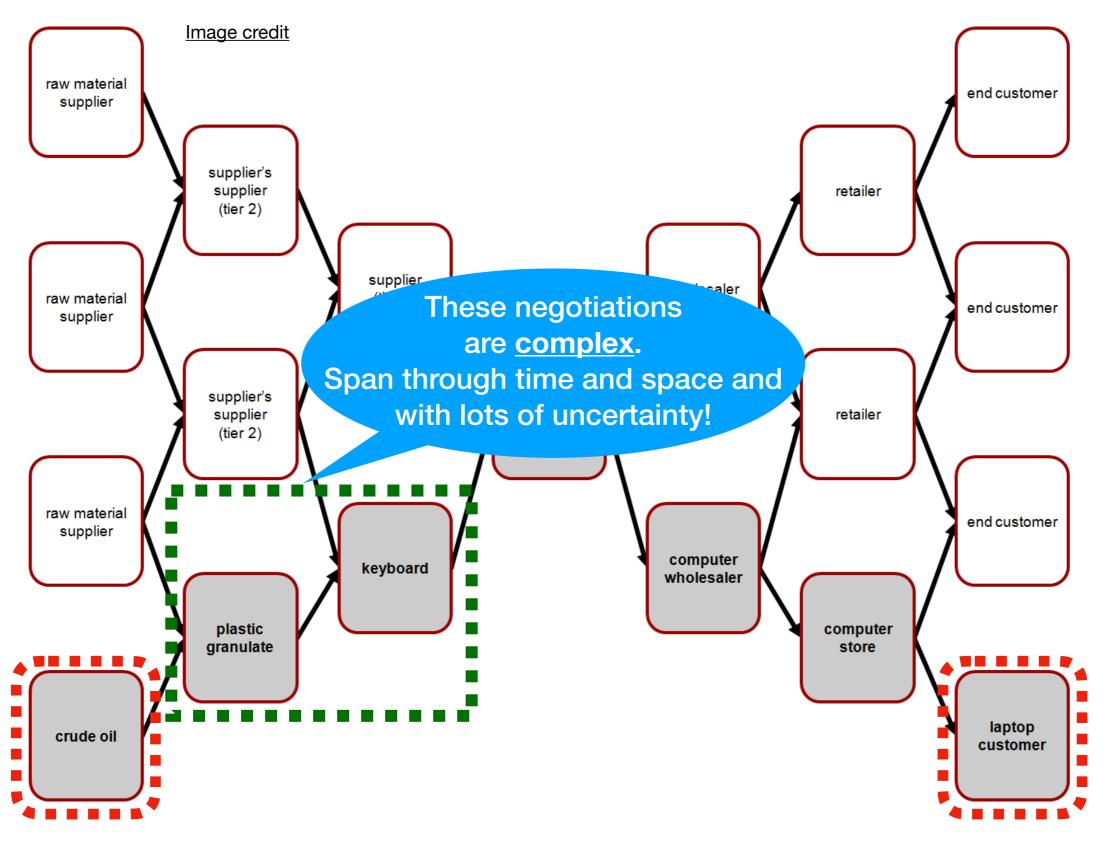
Instead, managers **negotiate** contracts to procure products.













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Defines how a negotiation is conducted e.g., alternate offers and counteroffers, one at a time

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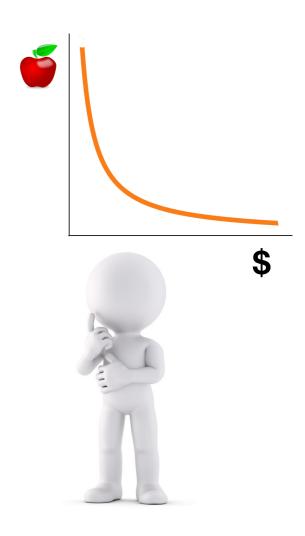
**Negotiation Strategy** 

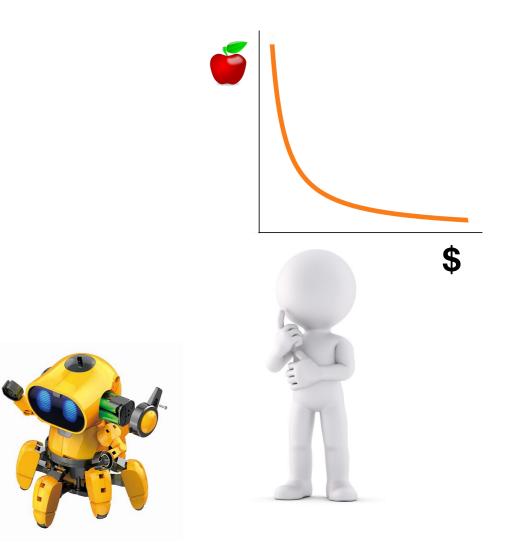
Defines how an actor behaves during negotiation e.g., be willing to accept "worst" deals as time passes

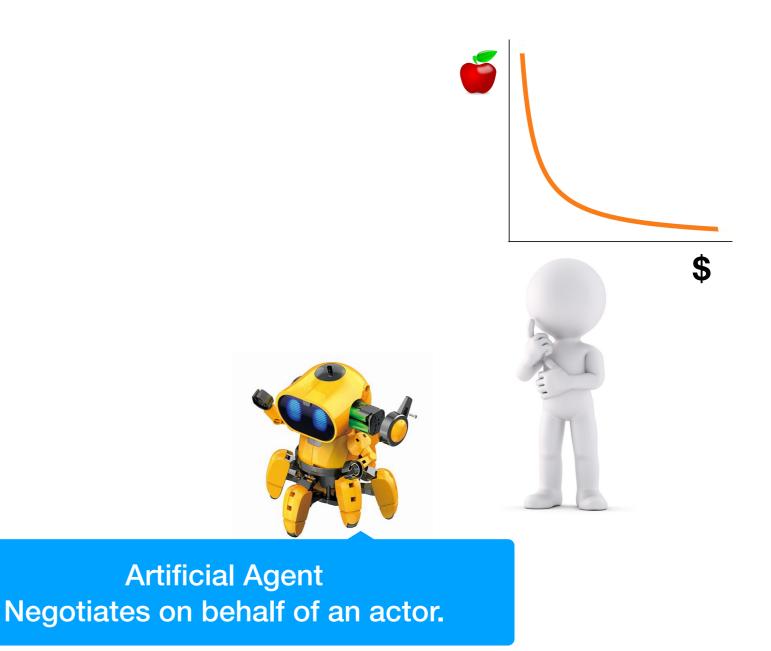


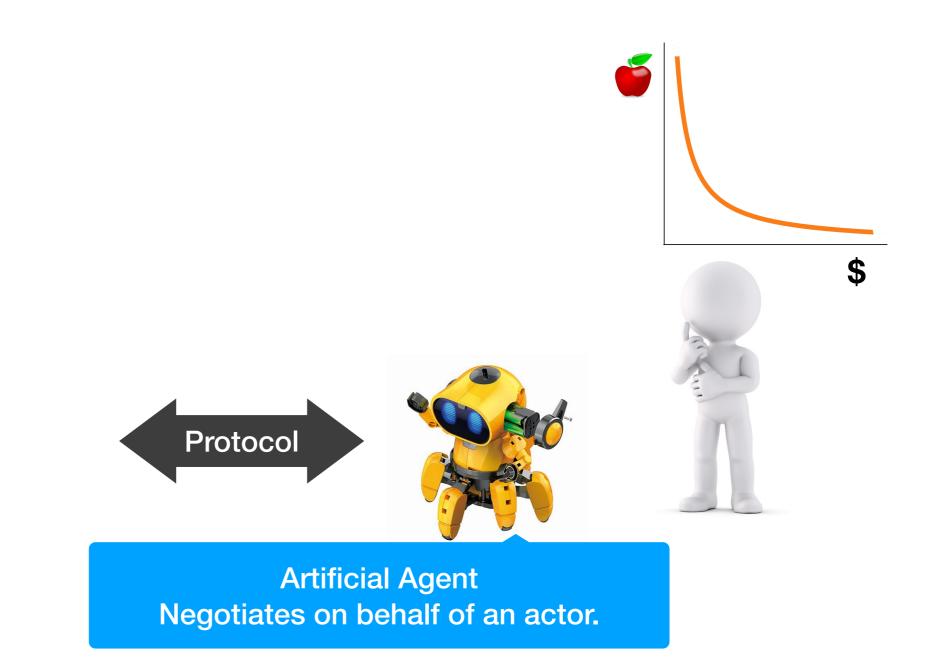
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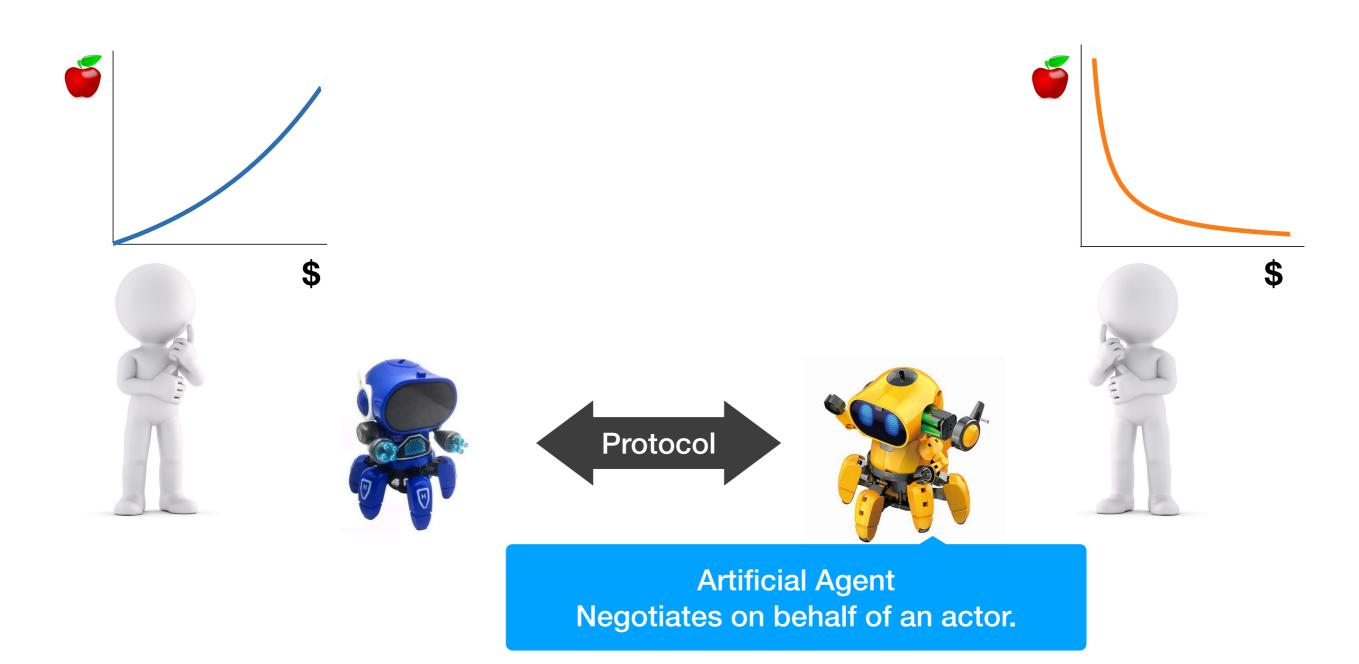


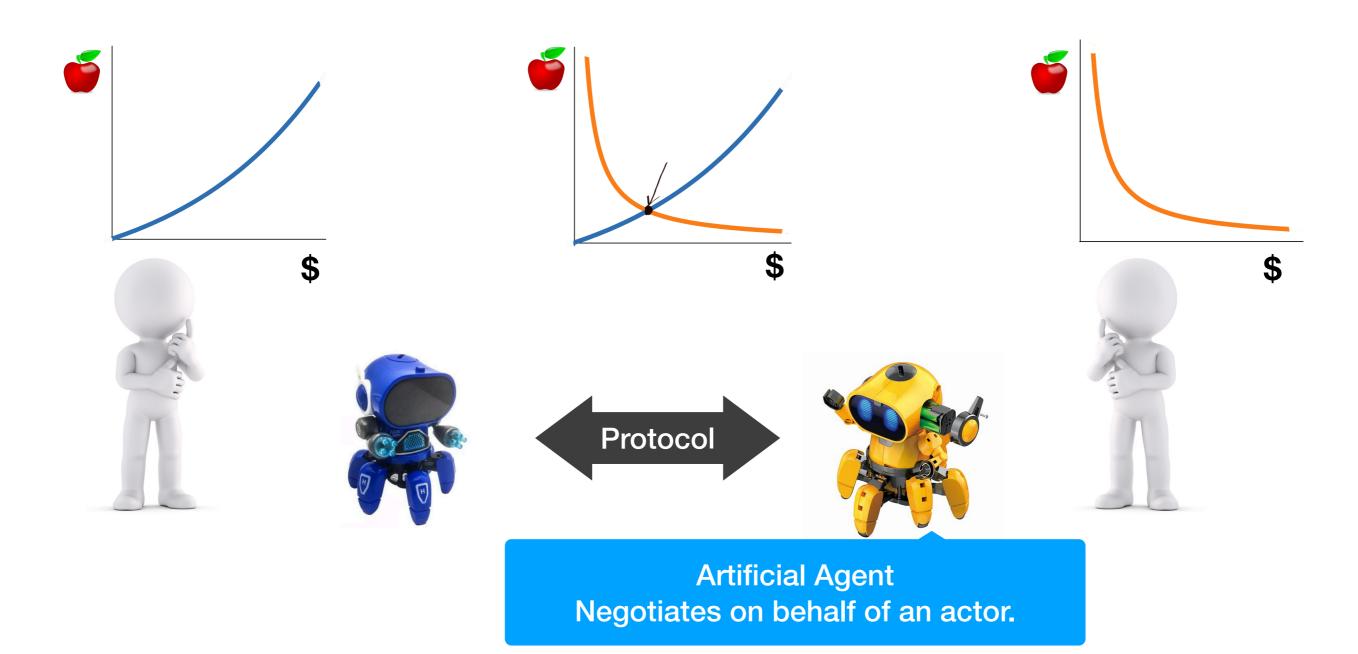


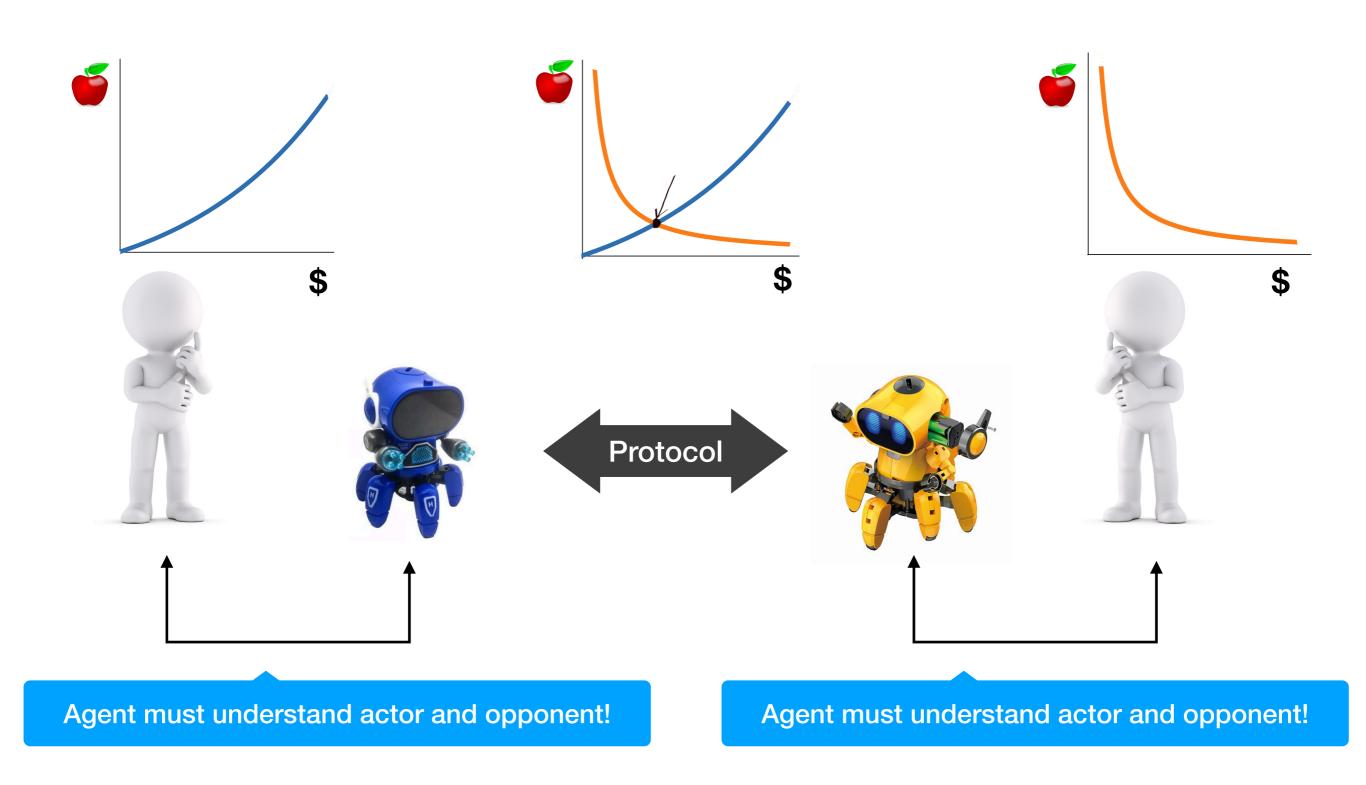


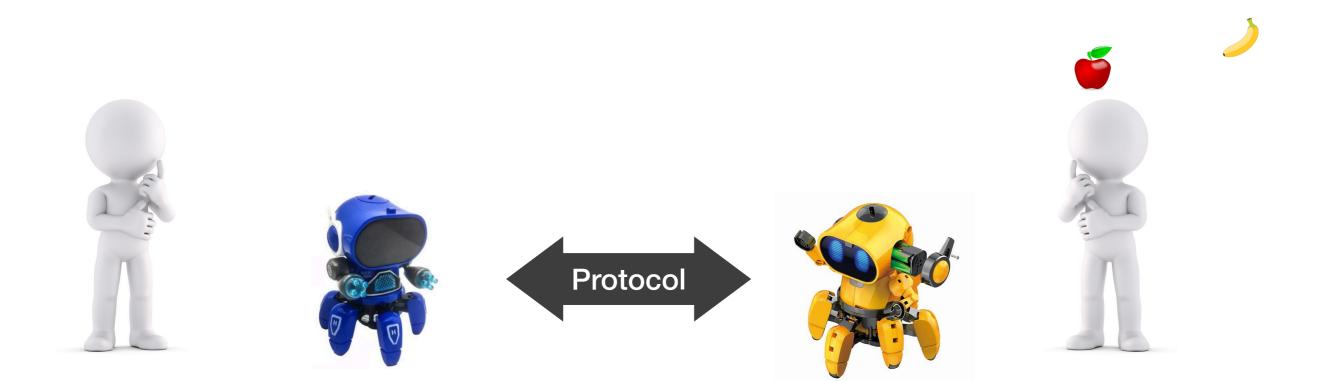


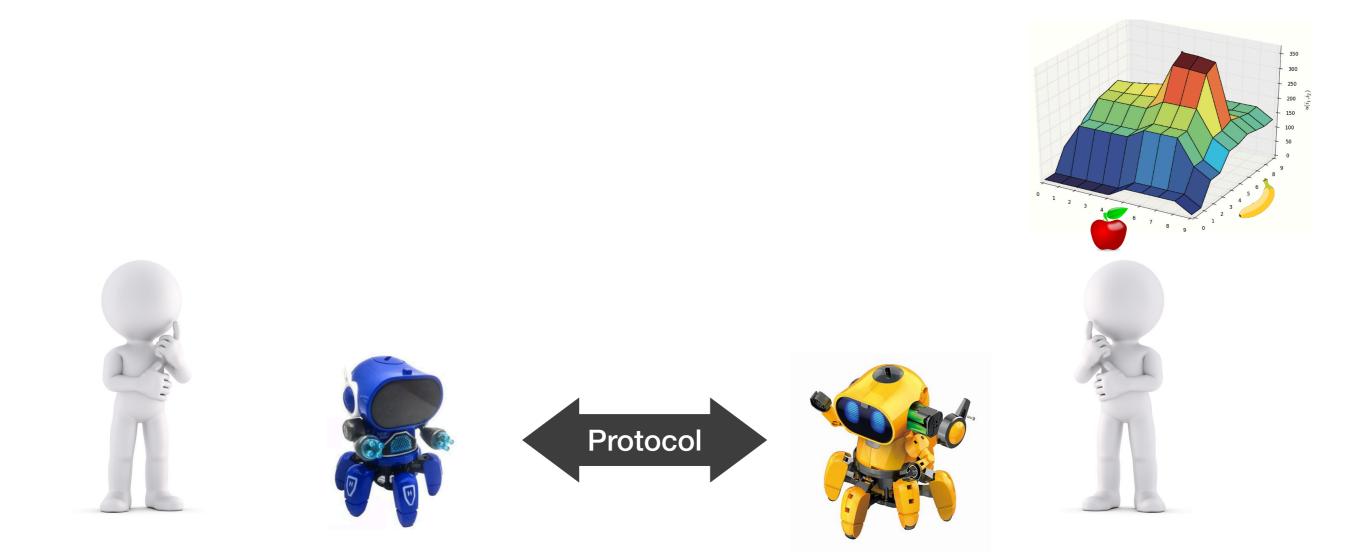


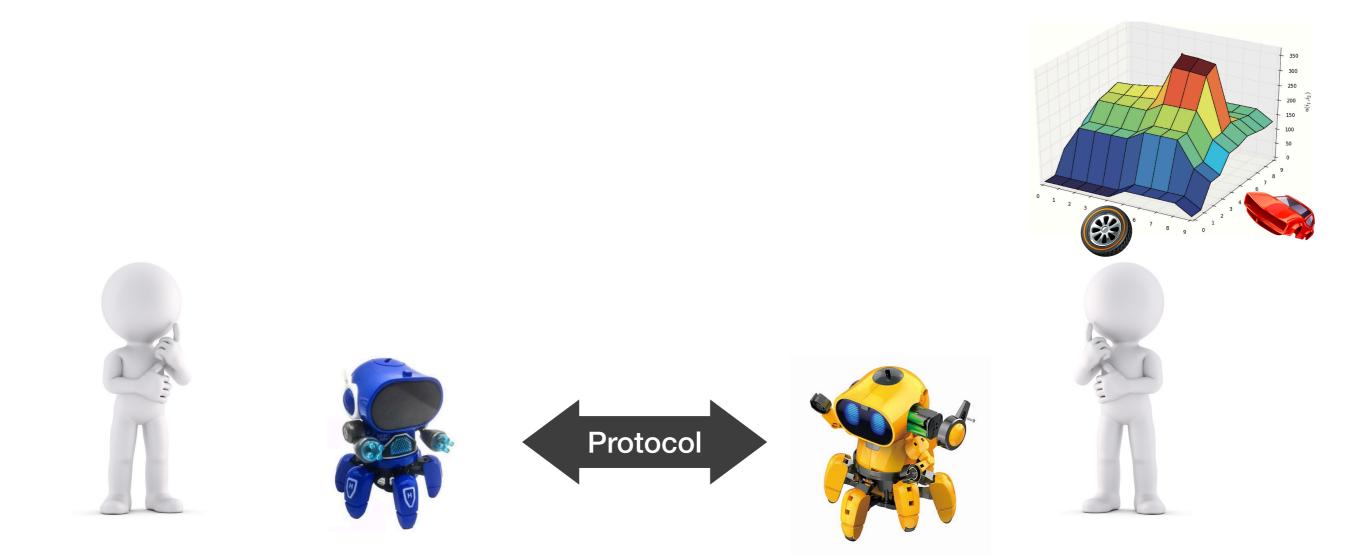




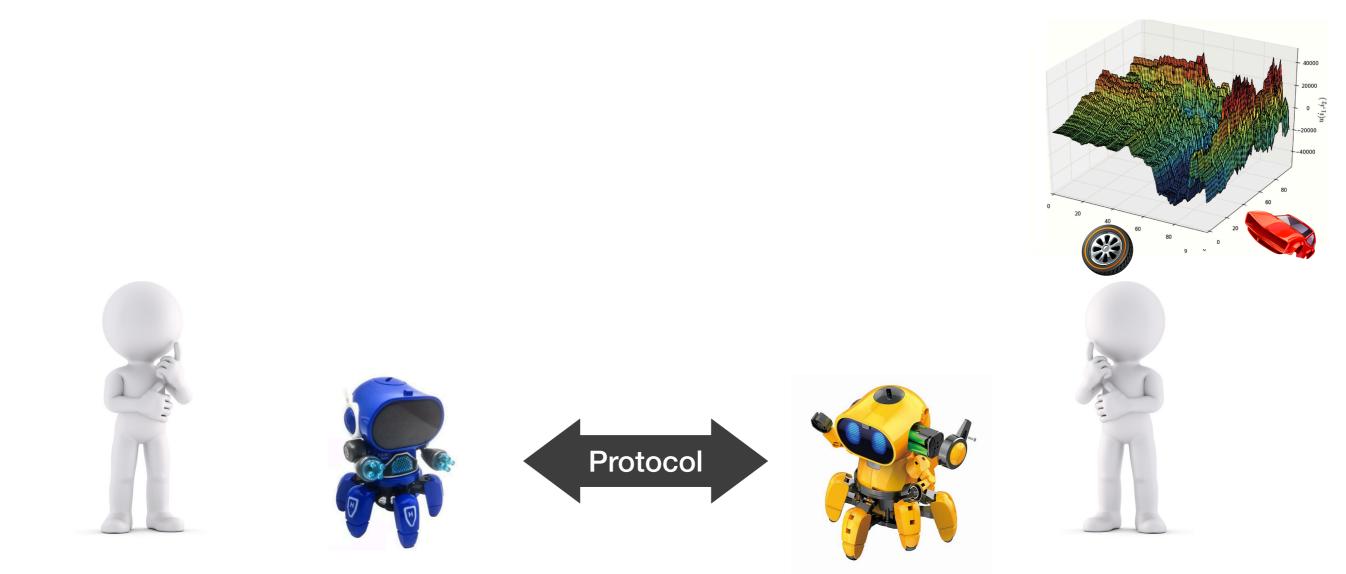




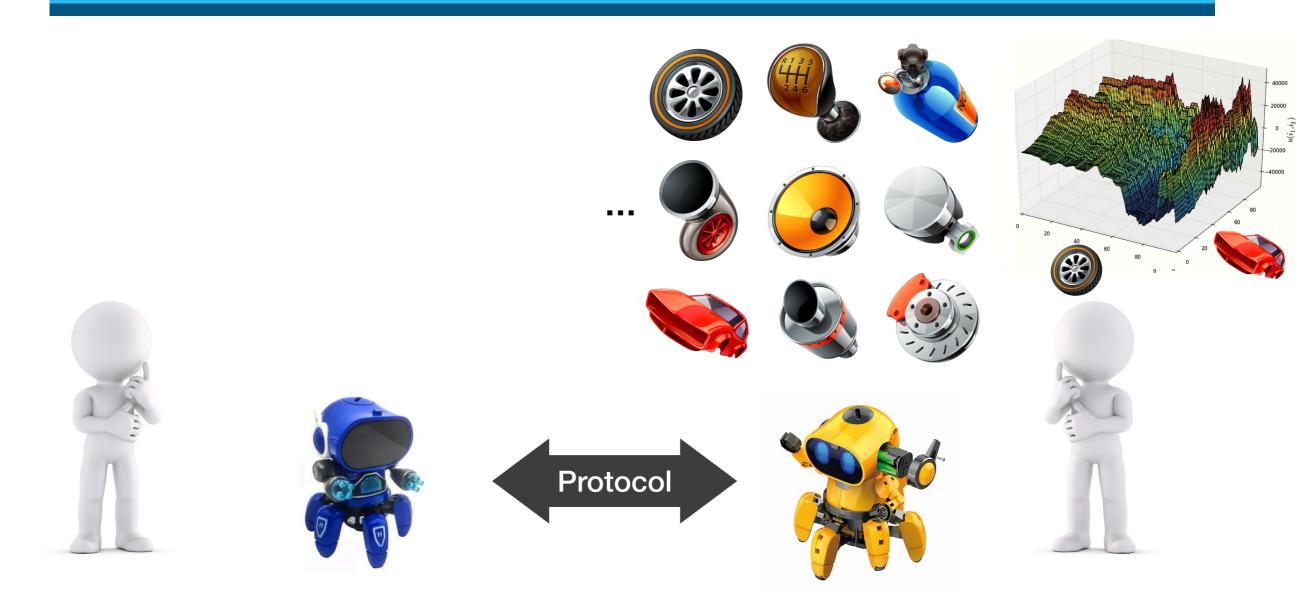




## Automated Negotiation



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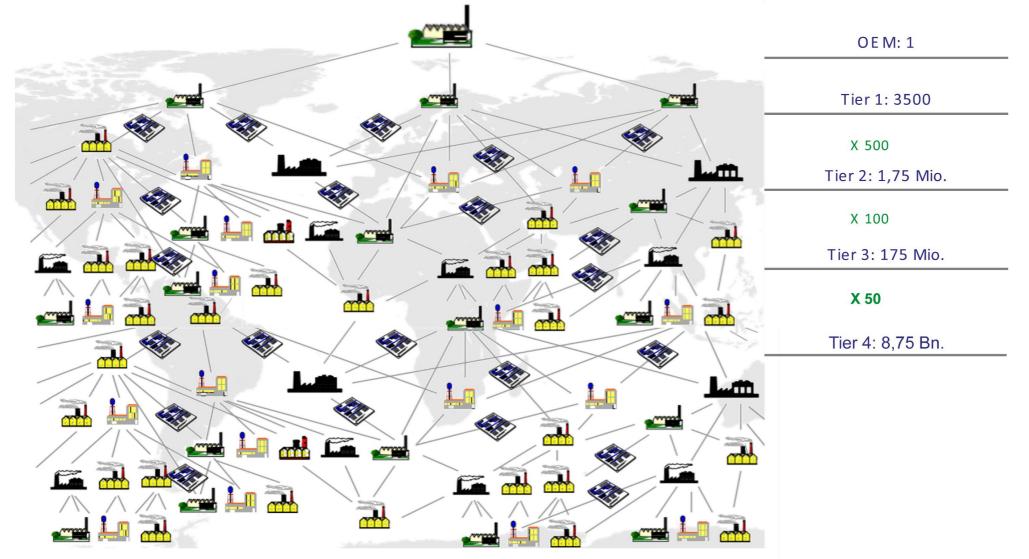
#### Automated Negotiation in Supply Chains

Why automate negotiations in supply chains?

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Why automate negotiations in supply chains?

# A complex supply chain



#### 50 to 80% suppliers are coming from outside Europe 3-8 levels in the Automotive supply chain

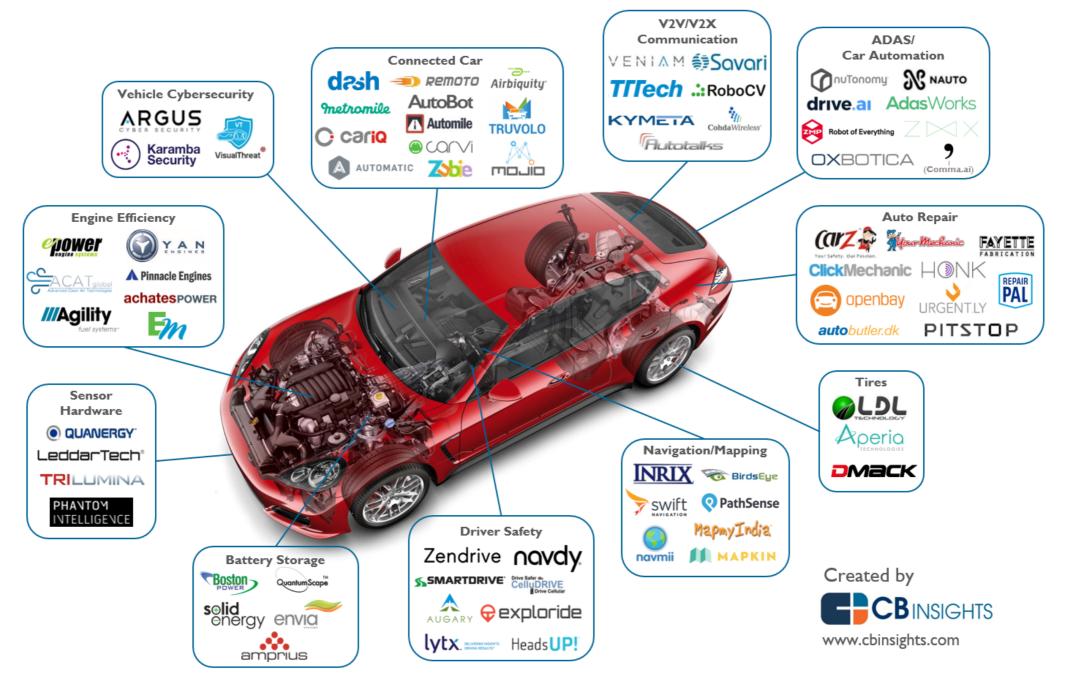
Copyright supply chain graphic and statistics: Edited from Timo Unger, Presentation "Automotive Industry & REACH Strategy & Challenges for Authorization", 2016 Copyright graphic map: Wikipedia Commons Picture by Crates, used under CC BY / Edited from original



## Automated Negotiation in Supply Chains

Why automate negotiations in supply chains?

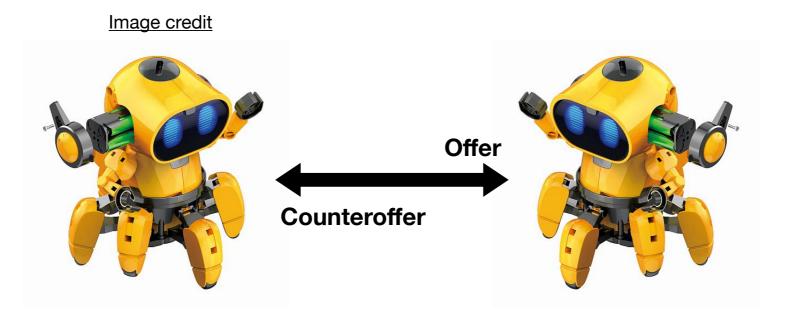
#### **Unbundling The Automobile**



Human negotiations lead to an estimated 17-40% value leakage (lots of lost \$\$) in some estimates (KPMG report: <u>https://bit.ly/3kDRy6l</u>)

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- Studies suggests that at least 15 companies are working to contract support systems (Forrester report: <u>https://bit.ly/3nwXEaY</u>)
- A recent proposal to standardize negotiation protocols for SCM and other applications (UN/CEFACT Project website: <u>https://bit.ly/38LOsLX</u>)



Part 2: Artificial Negotiation Agents for Supply Chains

How can we develop agents for supply chains? How can we measure our progress? Why (or why not) are games a good idea?

# Advancing Research via Not to be confused with video games!









Long-standing tradit



Garry Kasparov < @Kasparov63

Tweet

Many thanks to @RealAAAI, Amy Greenwald, and my illustrious co-panelists today. I'm happy to share by experiences and thought, but they are moving the world forward with their research and innovations. Fascinating conversation!

https://vimeo.com/389556398

#### 🔊 AAAI @RealAAAI · Feb 11

The AI History panel is beginning! With @Kasparov63, Michael Bowling, @murraycampbell, Hiroaki Kitano, David Silver, and Amy Greenwald.



7:39 PM · Feb 11, 2020 · TweetDeck

igence (AI) research.



#### International Automated Negotiating Agents Competition (ANAC)

From their website, http://web.tuat.ac.jp/~katfuji/ANAC2020:

"ANAC provides a unique benchmark for evaluating practical negotiation strategies in multi-issue **domains**".

#### <u>Leagues</u>

- \* Automated Negotiation
- \* Human-Agent
- \* LeagueWerewolf Game
- \* HUMAIN League
- \* Supply Chain Management (new, since 2019)



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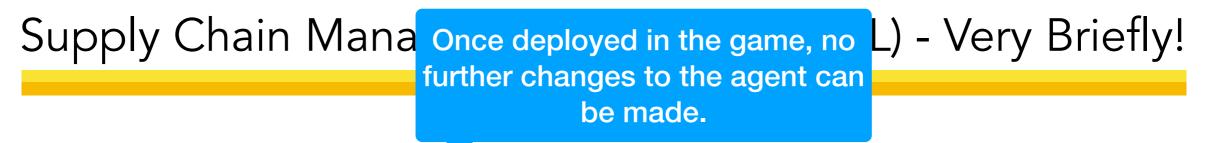
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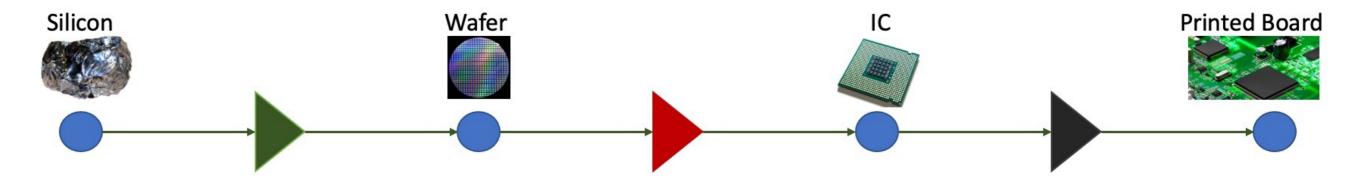
A league models some of the challenges that automated negotiation agents face in some domain

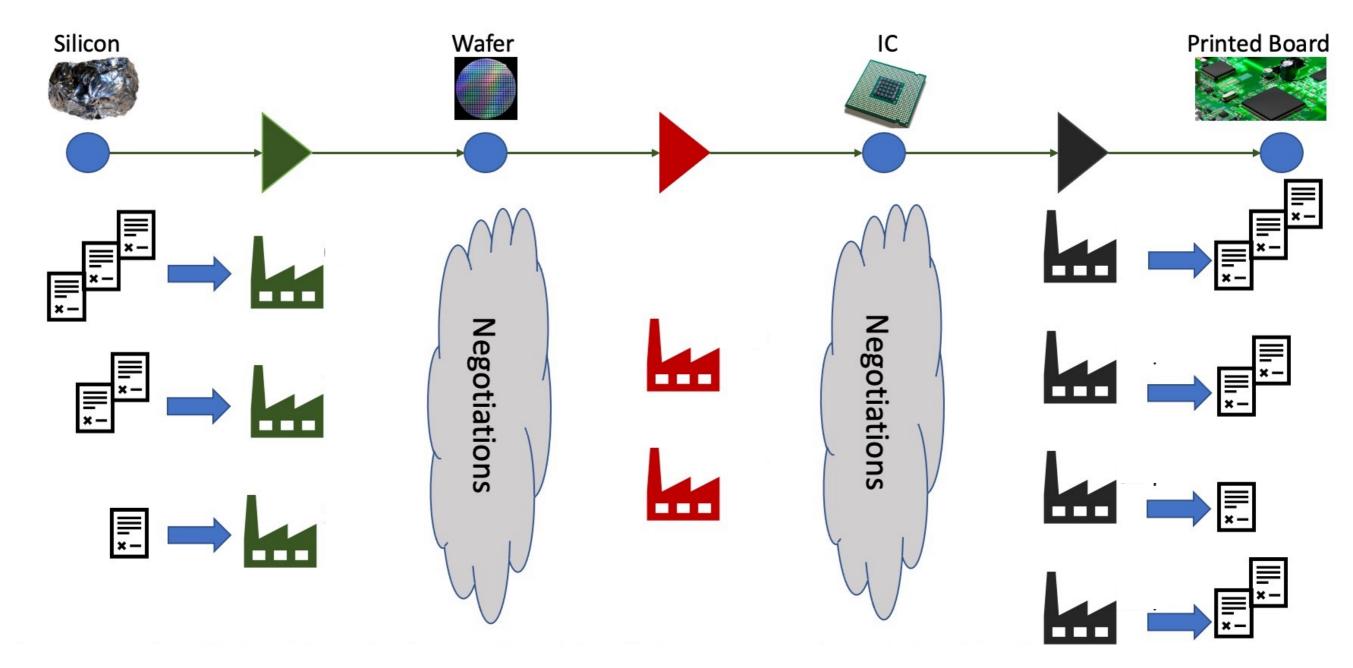


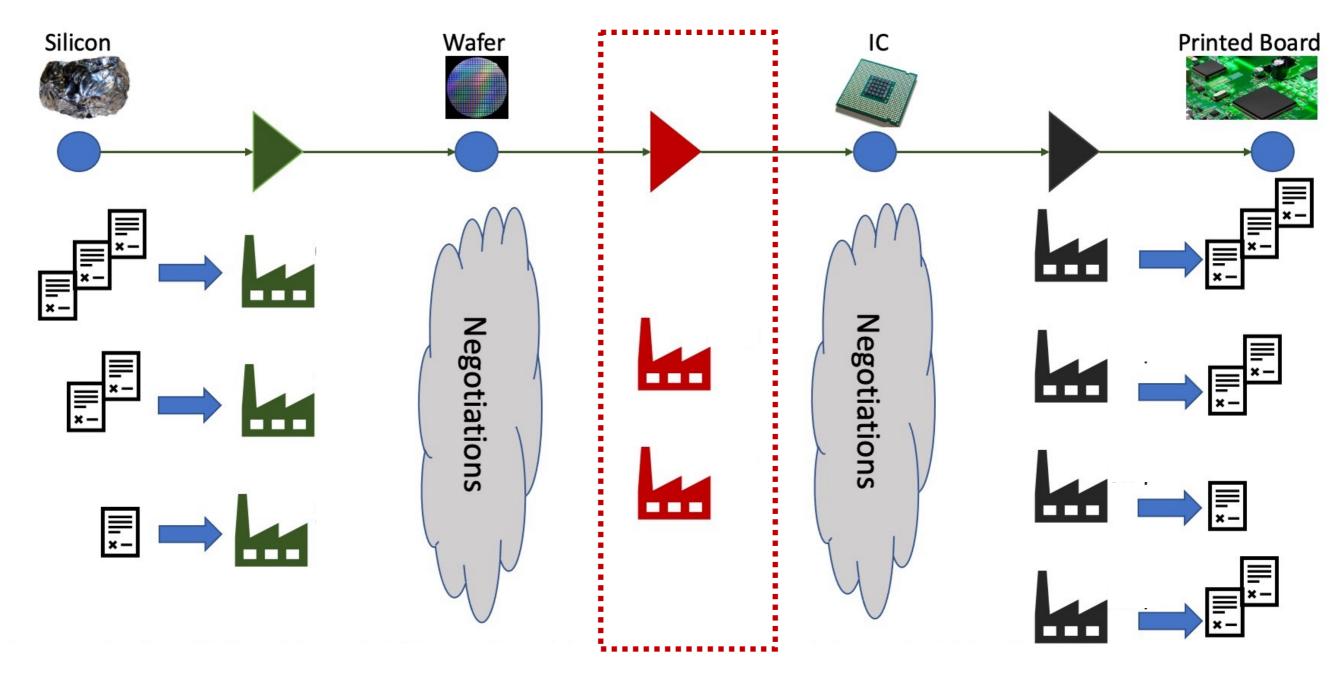


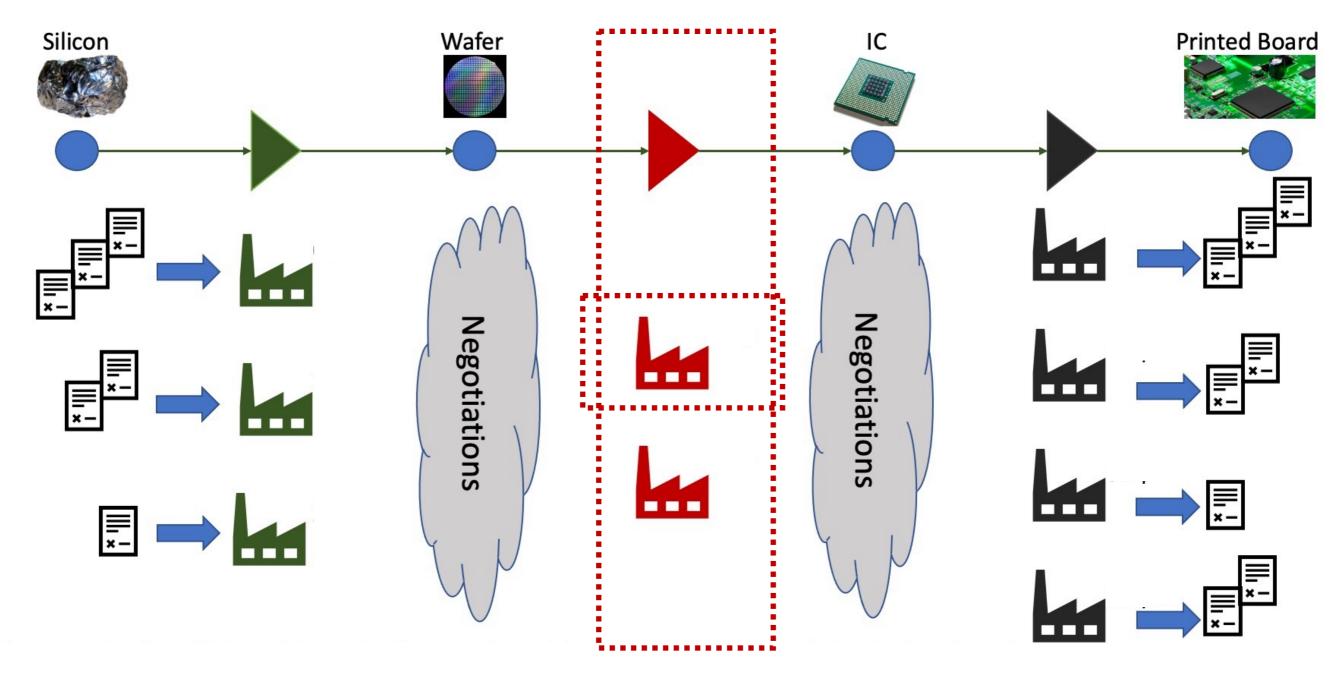
Design and build an **autonomous agent** that **negotiates** on behalf of a

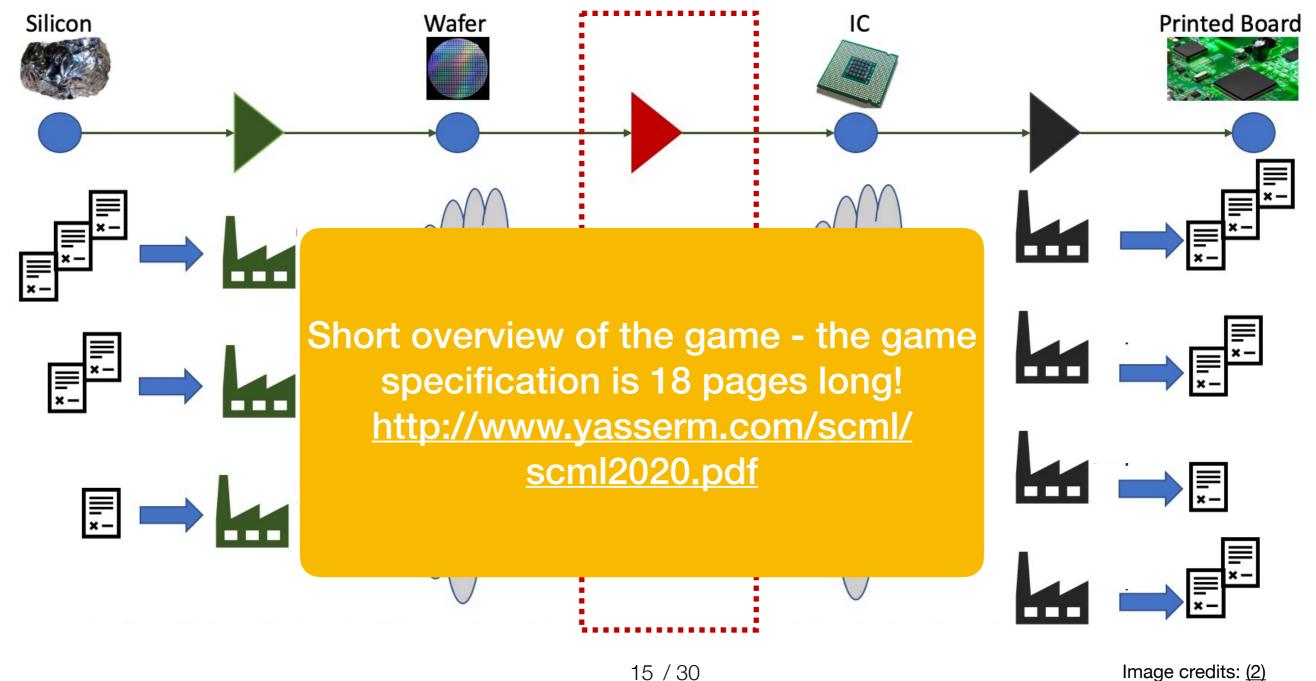
factory manager situated in a supply chain management simulation.











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This is an *intricate* challenge because your agent must (at least):

• **Devise** some utility function

[Start, Day 1, Day 2, Day 3, ... , End of game]

- **Devise** some utility function
- Negotiate with **multiple** other agents under **uncertainty**

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Resources	
Website	https://scml.cs.brown.edu/
Code	https://www.github.com/yasserfarouk/scml
YouTube Tutorials	<u>https://tinyurl.com/y5koqxfu</u>
Online Competition	https://scml.cs.brown.edu

## Our Agents for SCML

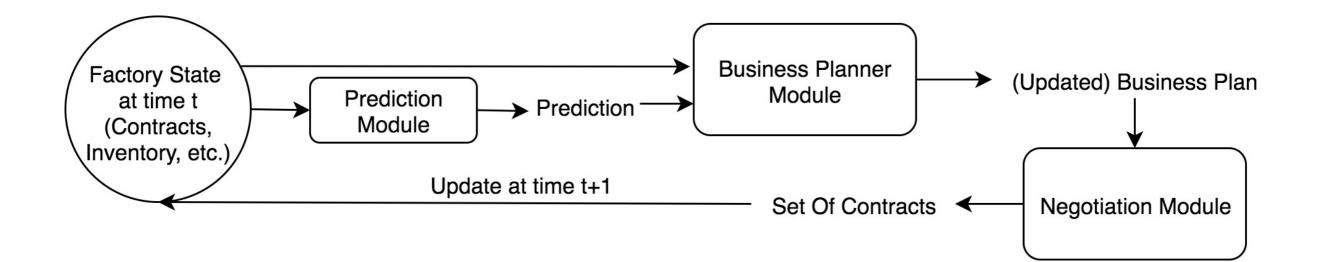
An agent's architecture in 3 modules:

- Prediction
- Business Planner
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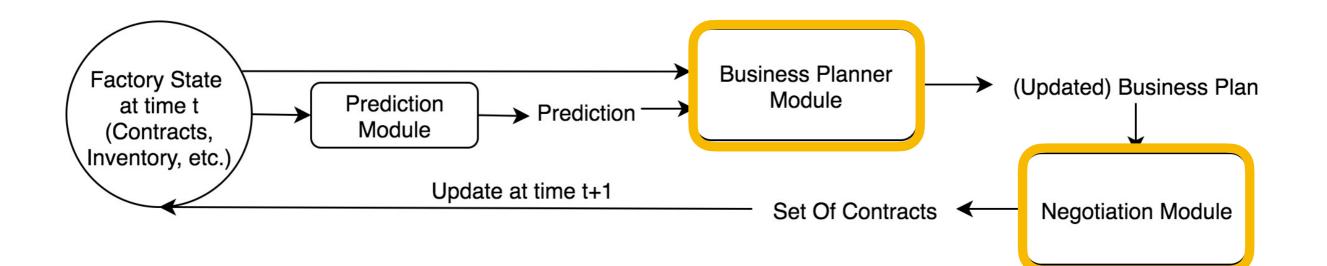
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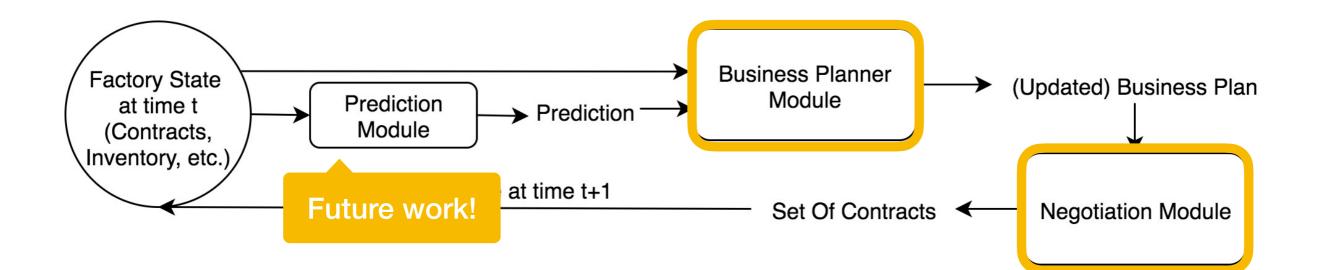
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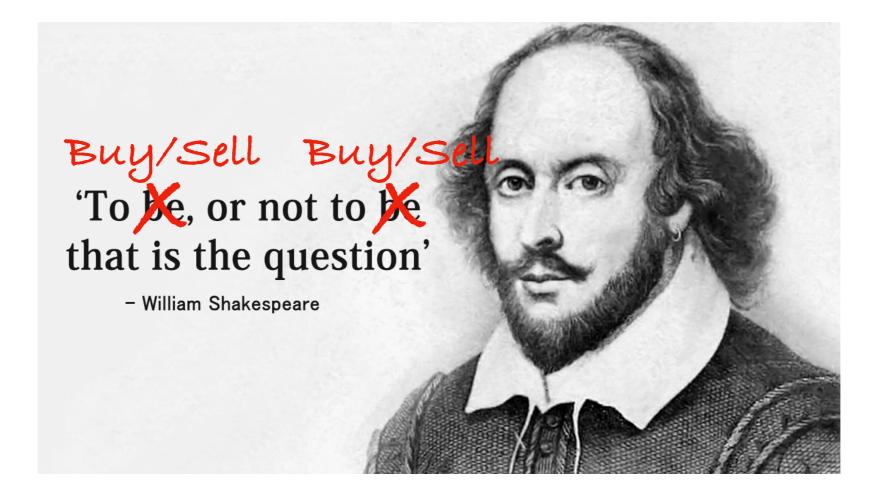
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## Our Agents for SCML - Business Plan

Uncertain **supply** (of raw materials) and **demand** (of finished products).

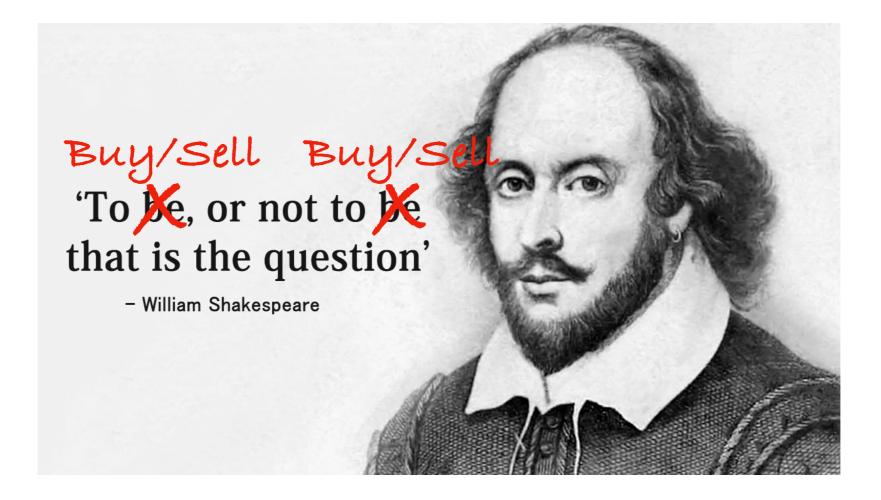
Still, agent needs to **buy** (raw materias) and **sell** (finished products) to earn profit.



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Let's start simple. Suppose there was no uncertainty. What to do then?

Suppose the agent **knows prices**:

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```
raw material, p_{raw} and finished product, p_{product}
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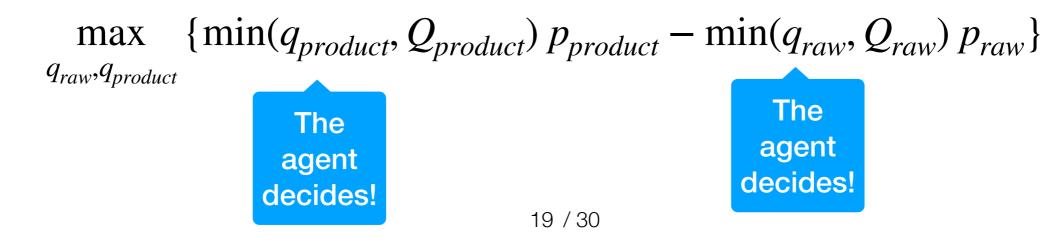
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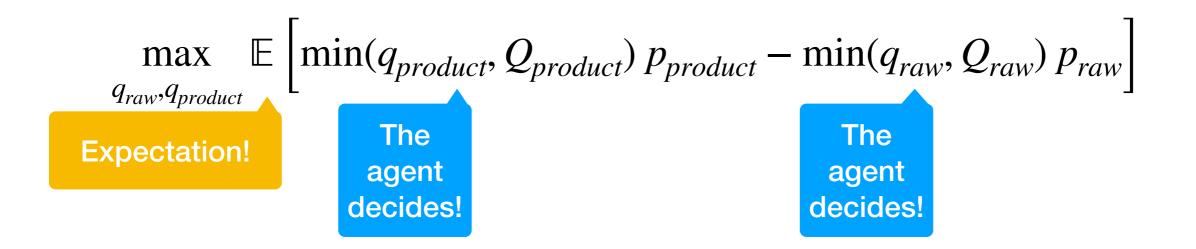
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 $\mathbb{E}\left[\min(q_{product}, Q_{product}) p_{product} - \min(q_{raw}, Q_{raw}) p_{raw}\right]$ max  $q_{raw}, q_{product}$ **Expectation!** 

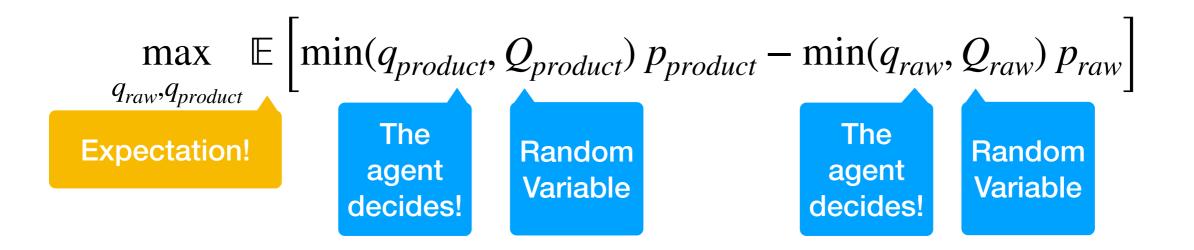
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$$\max_{q_{raw}^{t}, q_{product}^{t}} \mathbb{E}\left[\sum_{t=1}^{T} \min(q_{product}^{t}, Q_{product}^{t}) p_{product}^{t} - \min(q_{raw}^{t}, Q_{raw}^{t}) p_{raw}^{t}\right]$$

Subject to **inventory control** (cannot sell finished product before procuring raw materials!), and **acts accordingly**.

Ignoring *some* (ok, *a lot*! :-)) of details, our final agents for 2019/2020 solve this equation

$$\max_{\substack{q_{raw}^{t}, q_{product}^{t} \in I \\ actually multi-period!}} \mathbb{E}\left[\sum_{t=1}^{T} \min(q_{product}^{t}, Q_{product}^{t}) p_{product}^{t} - \min(q_{raw}^{t}, Q_{raw}^{t}) p_{raw}^{t}\right]$$

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A **plan** is a sequence  $q_{raw}^t$ ,  $q_{product'}^t$  for t = 1, 2, ..., T**Acts** means to implement the plan.

ac

Ignoring some (ok, Solving this equation is computationally challenging! solve this equation We efficiently solved it using dynamic programming and other algorithmic/AI techniques!

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High (low) utility for negotiations that drive the agent closer (farther) to **fulfilling** its plan.



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**Aspiration Negotiator**! (Time-based strategy, Faratin et.al. [4]) Start with some **aspiration level** (minimum utility willing to accept). To start, do not accept offers below the aspiration level.



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The plan can be used as a "**utility function**". How?

High (low) utility for negotiations that drive the agent closer (farther) to **fulfilling** its plan.

Aspiration Negotiator! (Time-based strategy, Faratin et.al. [4]) Start with some aspiration level (minimum utility willing to accept). To start, do not accept offers below the aspiration level. As time passes, concede some utility hoping to reach an agreement!



Actor (Institution or Individual) has! a

Implementing the plan **via** negotiations.

The plan can be used as a "**utility function**". How?

High (low) utility for negotiations that drive the agent closer (farther) to **fulfilling** its plan.

**Aspiration Negotiator**! (Time-based strategy, Faratin et.al. [4])

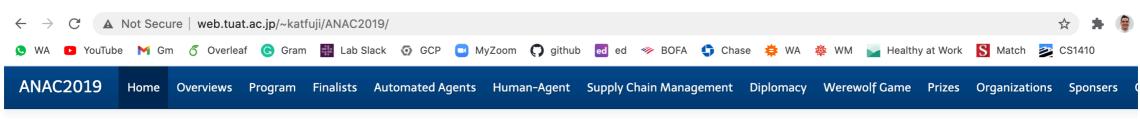
As L

Start with som	Lots of unspecified parameters!	to accept).
To start, do n	How to set initial aspiration level? How fast to concede?	
As time passe	Lots of opportunities for further research!	agreement!



Actor (Institution or Individual) has! a

### Our Agents for SCML - Results So Far



#### Winners

#### **Automated Agents League**

#### Individual Utility

	Winner AgentGG Shaobo Xu and Peihao Ren		Shaobo Xu and Peihao Ren	len University of Southampton	
			Ryohei Kawata	Tokyo University of Agriculture and Technology	
			Yuta Hosokawa	Tokyo University of Agriculture and Technolog	

#### Joint Utility

Winner winkyAgent Siqi Chen and Jie Lin		Siqi Chen and Jie Lin	Tianjin University	China
2nd FSEGA2019 Stancu Anca		Stancu Anca	Bábes Bolyai University	România
3rd	AgentGP	Tomoya Fukui	Nagoya Institute of Technology	Japan

#### Human-Agent League

1st	Draft	Bohan Xu and Shadow Pritchard and James Hale and Sandip Sen	University of Tulsa	USA
2nd	Dona	Eden Shalom Erez, Inon Zuckerman, Galit Haim	Ariel University and The College of Management Academic Studies	Israel

#### Supply Chain Management League

#### **Standard Category**

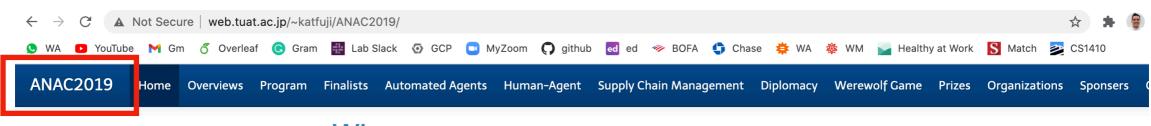
1	st	IFFM	Masanori Hirano, Taisei Mukai, Hiroyasu Matsushima, Kiyoshi Izumi	University of Tokyo	Japan
2	nd	NVM	Enrique Areyan Viqueira, Amy Greenwald	AIST and Brown University	Japan and USA
3	rd	SAHA	Nahum Alvarez	AIST	Japan

#### **Collusion Category**

1st	IFFM	Masanori Hirano, Taisei Mukai, Hiroyasu Matsushima, Kiyoshi Izumi	University of Tokyo	Japan
2nc	MVM	Enrique Areyan Viqueira, Amy Greenwald	AIST and Brown University	Japan and USA
3rd	I FJ2	Ryohei Kawata	Tokyo University of Agriculture and Technology	Japan

#### Sabotage Category

Honorable Mention Award Monopoly Ryoto Ishikawa and Yuta Hosokawa Tokyo University of Agriculture and Technology Japan



# Winners

### **Automated Agents League**

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- [	Winner	AgentGG	Shaobo Xu and Peihao Ren	University of Southampton	UK
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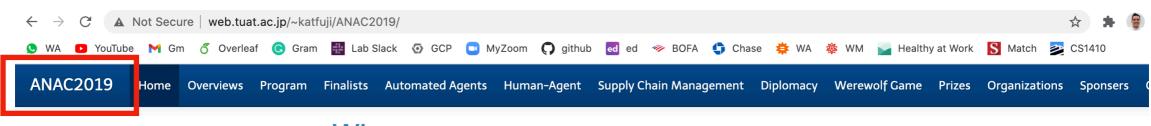
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## Supply Chain Management League

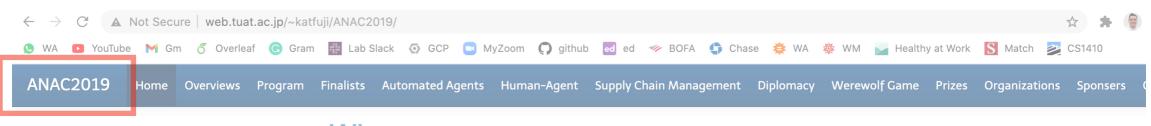
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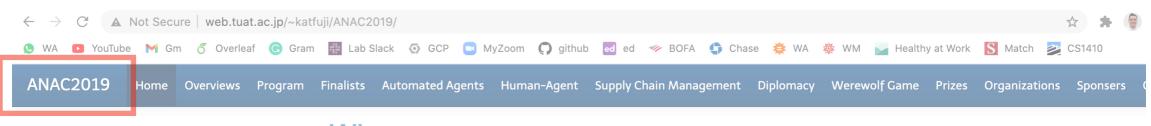
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#### Sabotage Category



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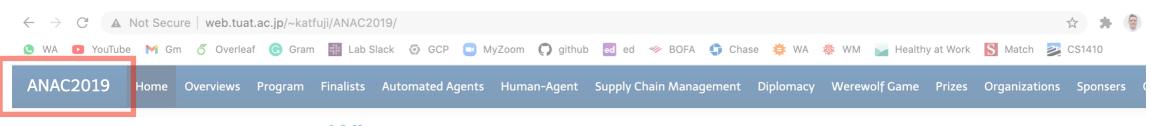
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#### Sabotage Category

▲ Not Secure web.tuat.ac.ip/~katfuji/ANAC2019/ [scml] Results of the Qualifications Round of ANAC 2020 SCML League (Standard) Brown × Yasser Mohammad (SCML Organization Committee) yasserfarouk@gmail.com via lists.cs.brown.edu Wed, Aug 26, 6:29 AM to scml 👻 Dear Enrique Areyan We are happy to inform you that your agent MontyHall was qualified for the final round of ANAC 2020 SCM league (Standard track). We received 22 submissions, 19 entered the qualifications after removing agents that did not fully comply with all submission requirements. Twelve agents were qualified for the Standard track and 6 agents were qualified for the Collusion track. We are looking forward to seeing (or virtually meeting) you at IJCAI 2020 in January. The final results of the league will be announced at the ANAC session as part of IJCAI's official competitions track. We thank you again for the time and effort you put into this agent. **Best Regards** ugrads MONTY HALL: BROWN UNIVERSITY'S AGENT FOR THE SUPPLY CHAIN MANAGEMENT LEAGUE OF THE Yasser Mohammad **2020 AUTOMATED NEGOTIATING AGENTS COMPETITION** representing the SCML Organization Commmittee Enrique Areyan Viqueira, Edward Li, Daniel Silverston, Amrita Sridhar, James Tsatsaros, Andrew Yuan, and Amy Greenwald **OVERVIEW.** At a high level, MontyHall's strategy is to buy inputs, immediately convert them into outputs, and Sent via Mail Merge for Gmail then sell said outputs. Monty begins its day by computing a business plan. For each day into the future (up to some Reply K Reply all Forward

Sabotage Category

lonorable Mention Award Monopoly Ryoto Ishikawa and Yuta Hosokawa Tokyo University of Agriculture and Technology Japar

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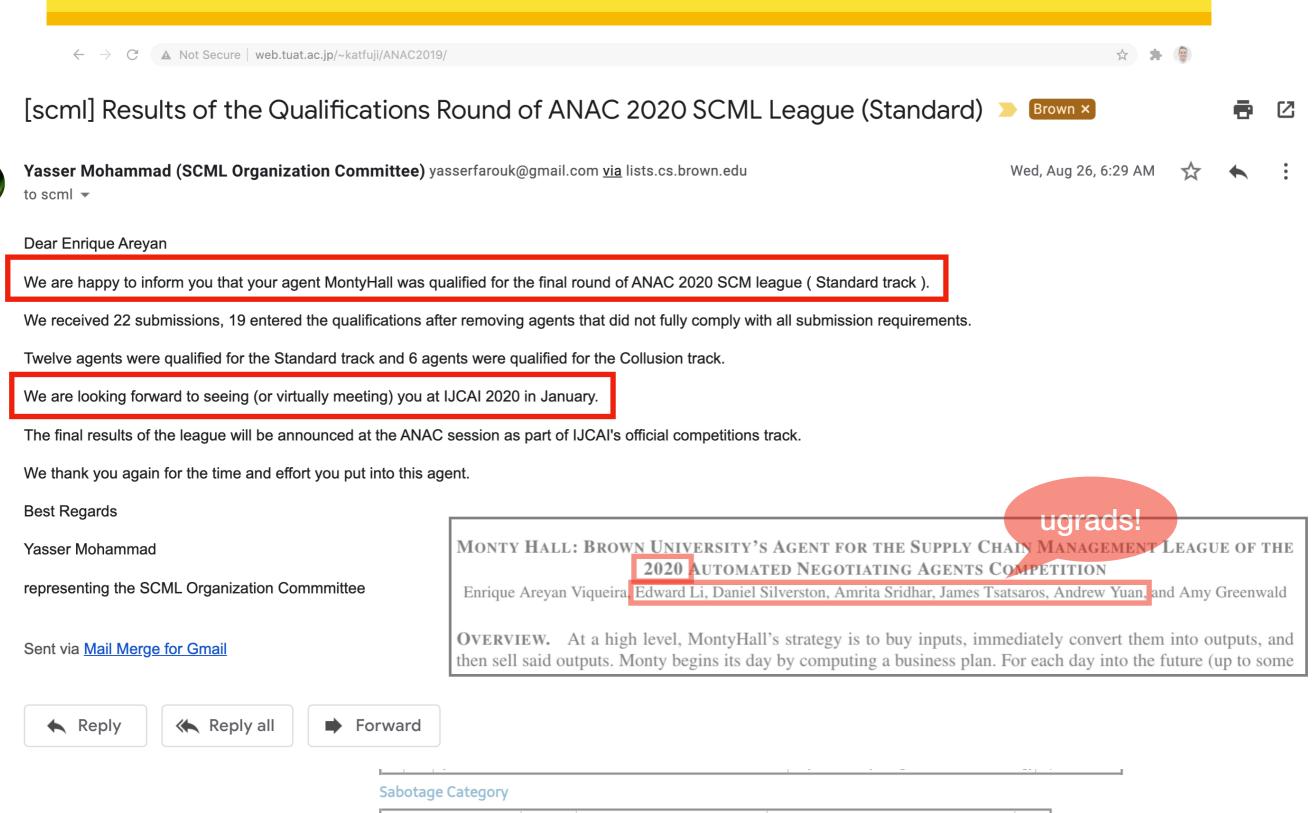


Image credit: http://theracquet.org/4747/news/ask-5-uwl-students-tell-their-future-plans/

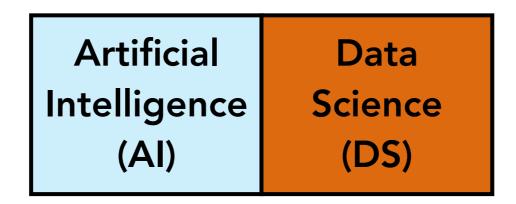


Part 3: Future Research Plans

We are no where near done! There are lots of research directions to pursue... A bit about other related and tangential research. We have an **agent architecture**, but many pieces need more work!

This research is **open-ended** and **interdisciplinary** by nature!

Artificial Intelligence (AI)	Data Science (DS)	Economics (EC)	Operations Research (OR)	+ more
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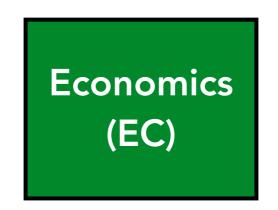


Our current prediction module (prices, supply, demand) just counts data.

Can we use machine learning techniques that provably "learn" from data?

In my research, I work on developing **machine learning** algorithms that **provably** learn concepts from data.

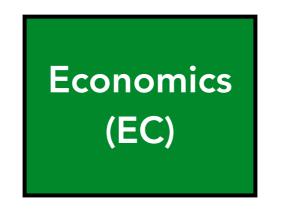
Improved Algorithms for Learning Equilibria in Simulation-Based Games. Enrique Areyan Viqueira, Cyrus Cousins, Amy Greenwald. 19th International Conference on Autonomous Agents and MultiAgent Systems (AAMAS20).



Are current rules of the ANAC SCML game the "right" ones?

In my research, I also develop machine learning techniques to design games (**mechanism design**). This is a **big**, open opportunity!

Empirical Mechanism Design: Designing Mechanisms from Data. Enrique Areyan Viqueira, Cyrus Cousins, Yasser Mohammad, Amy Greenwald. Uncertainty in Artificial Intelligence (UAI19).



Incentives players (researchers) to do their best and not just "hack" the game!

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Empirical Mechanism Design: Designing Mechanisms from Data. Enrique Areyan Viqueira, Cyrus Cousins, Yasser Mohammad, Amy Greenwald. Uncertainty in Artificial Intelligence (UAI19). The techniques and ideas develop for ANAC SCML extend **beyond**!

# Other domains we have tackled: advertisement exchange markets.

On Approximate Welfare-and Revenue-Maximizing Equilibria for Size-Interchangeable Bidders. Enrique Areyan Viqueira, Amy Greenwald, Victor Naroditskiy. 16th International Conference on Autonomous Agents and MultiAgent Systems (AAMAS17). The techniques and ideas develop for ANAC SCML extend **beyond**!

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As our economies grow more sophisticated, multi-agent systems research will become **even more important**.





# Collaborators



**Amy Greenwald** 



**Cyrus Cousins** 



# **Yasser Mohammad**



# **Marilyn George**

29 / 30

# Collaborators



**Amy Greenwald** 



**Cyrus Cousins** 



# **Yasser Mohammad**



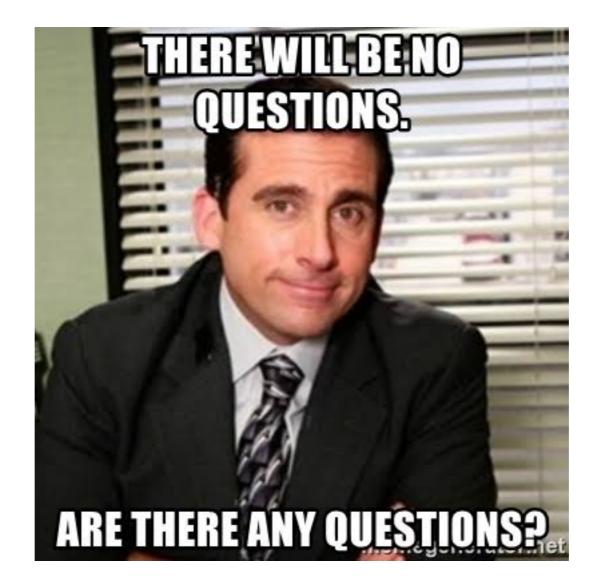
**Marilyn George** 

?

Future Collaborators! Students Colleagues Researchers All are welcome!

29 / 30

# Thank you for your attention!



Mohammad, Y., Viqueira, E. A., Ayerza, N. A., Greenwald, A., Nakadai, S., and Morinaga,
 S. (2019). <u>Supply chain management world</u>. In International Conference on Principles and
 Practice of Multi-Agent Systems, pages 153–169. Springer.

[2] World Trade Organization. <u>Supply Chain Perspectives</u>.

[3] Mohammad, Y. <u>PRIMA 2020 SCML Tutorial</u>.

[4] Faratin, P., Sierra, C., & Jennings, N. R. (1998). <u>Negotiation decision functions for</u> <u>autonomous agents</u>. Robotics and Autonomous Systems, 24(3-4), 159-182

[5] Hadfi, R., & Ito, T. (2015). <u>Complex multi-issue negotiation using utility hyper-graphs</u>. Journal of Advanced Computational Intelligence and Intelligent Informatics, 19(4), 514-522.