

Collective Phenomena in Complex Networks

Ali Jadbabaie

JR East Professor of Engineering

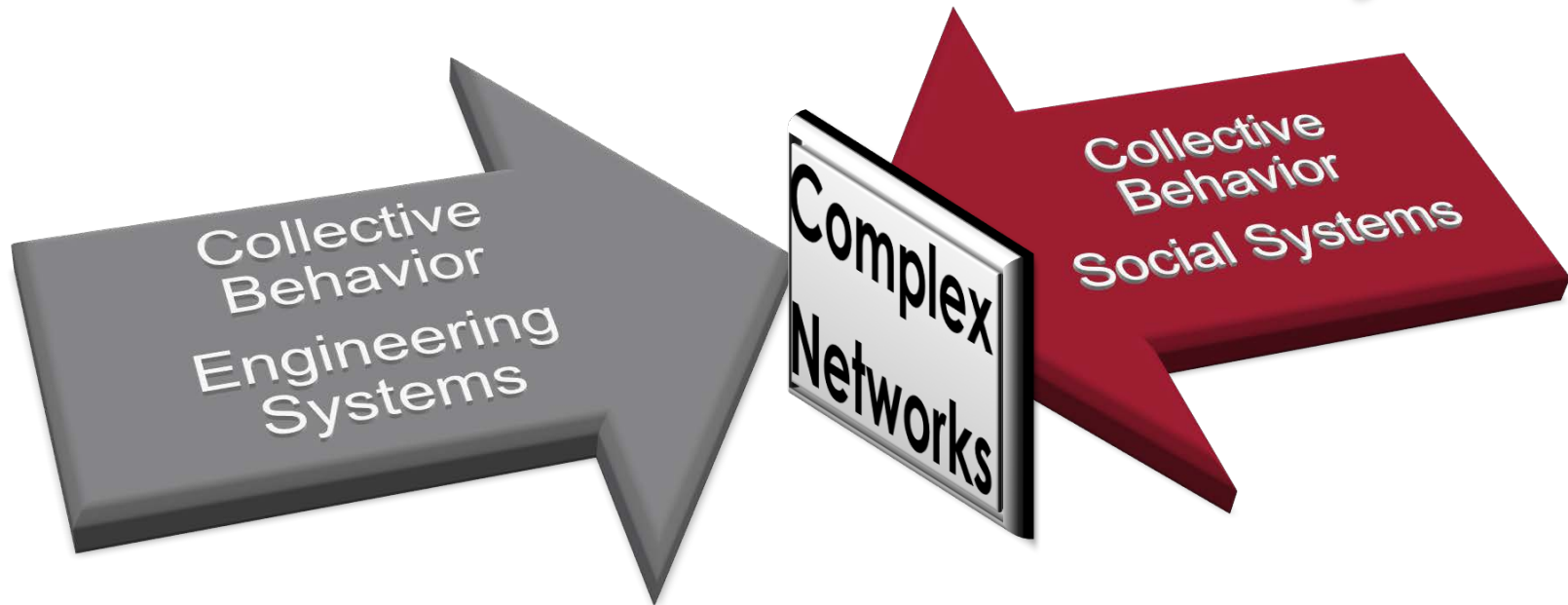
Associate Director, Institute for Data, Systems, and Society

Director, Sociotechnical Systems Research Center

<http://web.mit.edu/~jadbabai/www>



Research Summary:



Flocking & Motion Coordination
Synchronization, Consensus
Optimization for Machine Learning

Social Learning
Opinion Dynamics
Controlling Epidemics

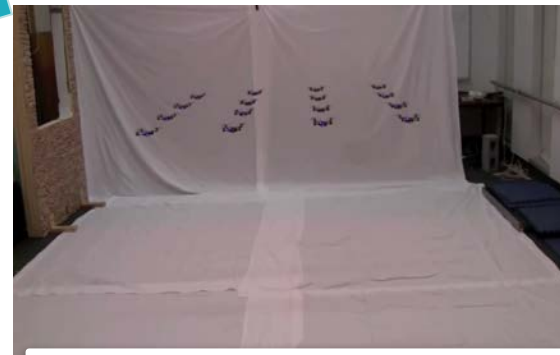
Tools: Network Science, Systems Theory, Dynamics and Control
Domains: Autonomy, Sensor & Infrastructure Networks, Social Phenomena and Social Behavior

From Single to Multi-agent Systems

Complexity of dynamics



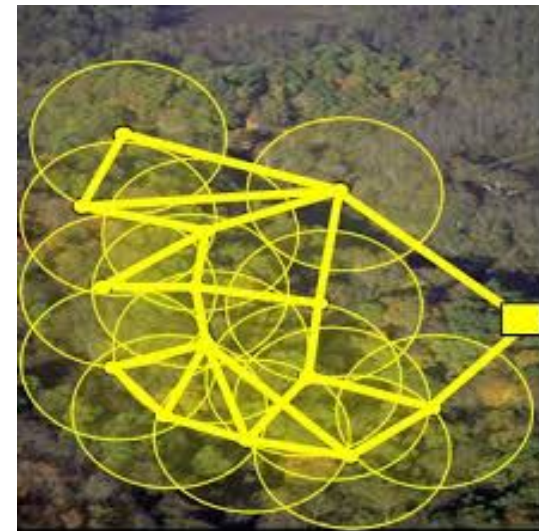
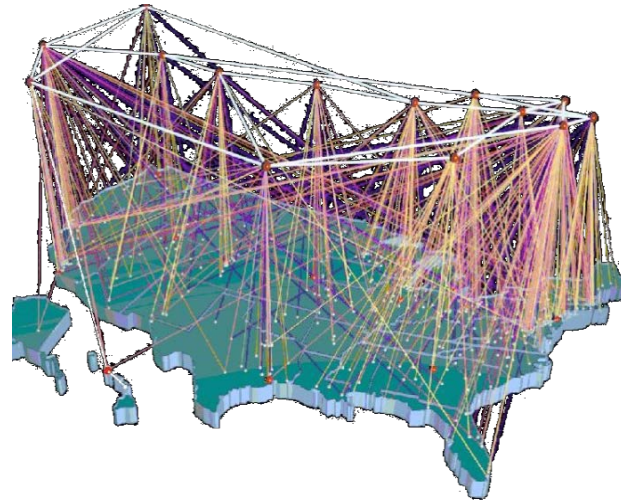
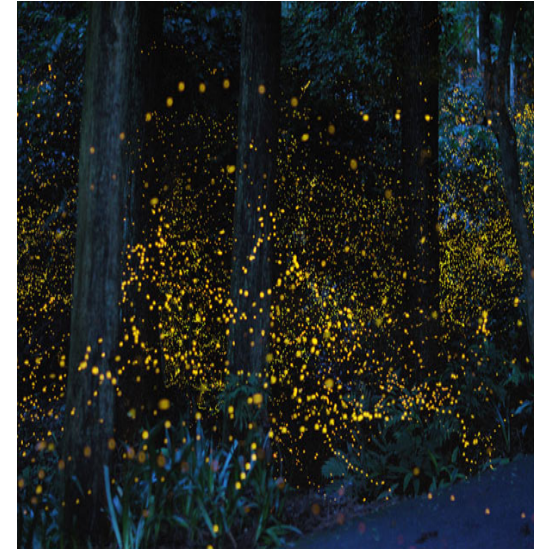
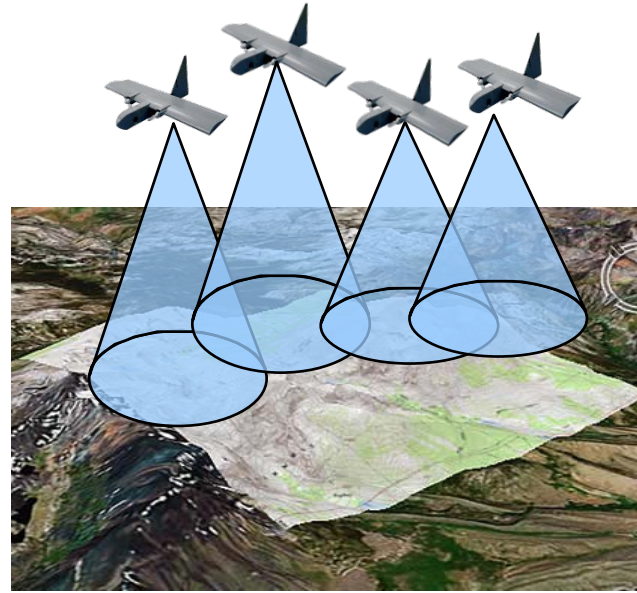
Single agent, circa 2001
Optimization-based control



multi-agent, circa 2012,
Melinger, Kushliev, Kumar

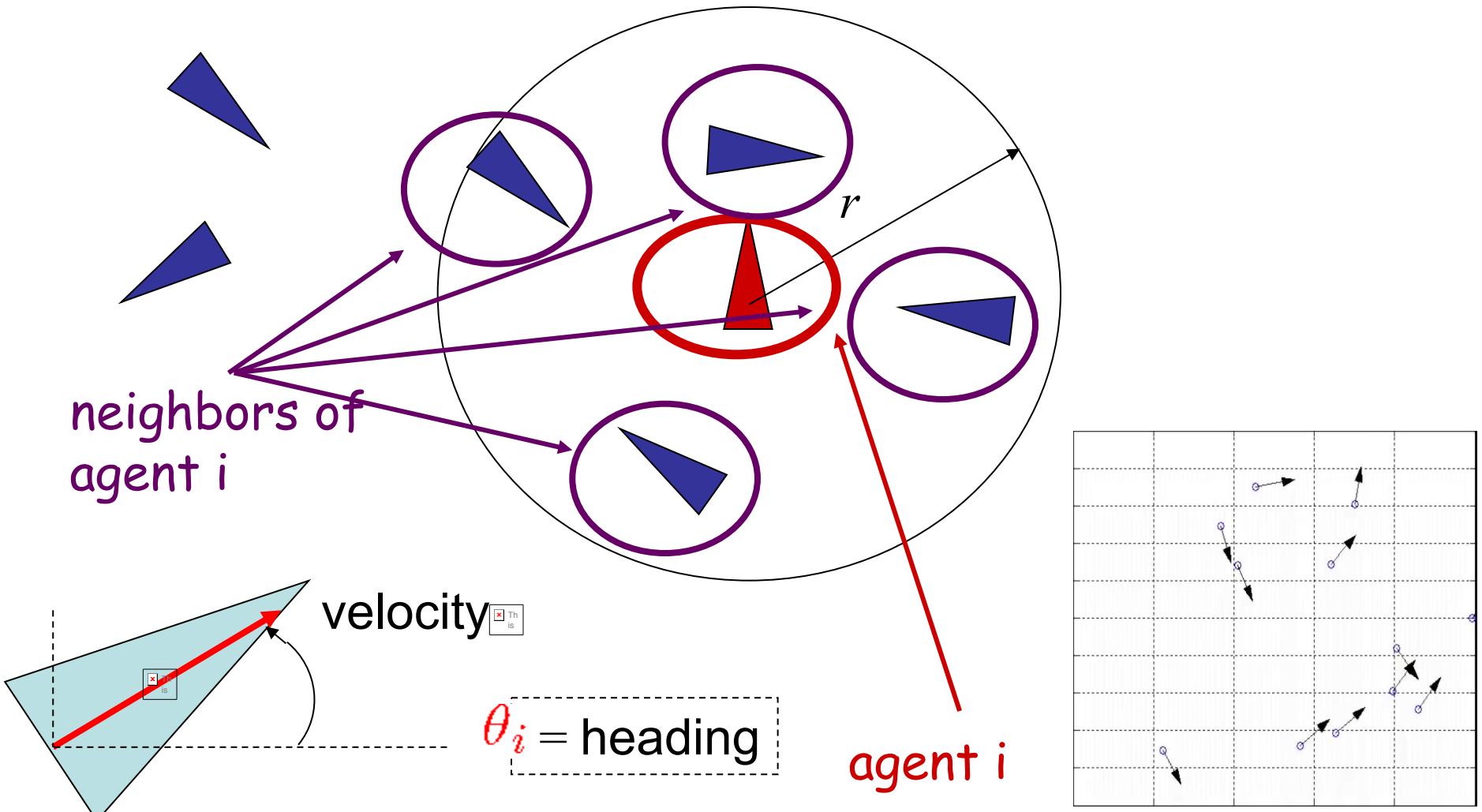
Complexity of interconnection

Collective Behavior: Autonomous Systems



A Kinematic Model of Flocking: Vicsek '95

'Average' heading of each agent and its neighbors, move one step

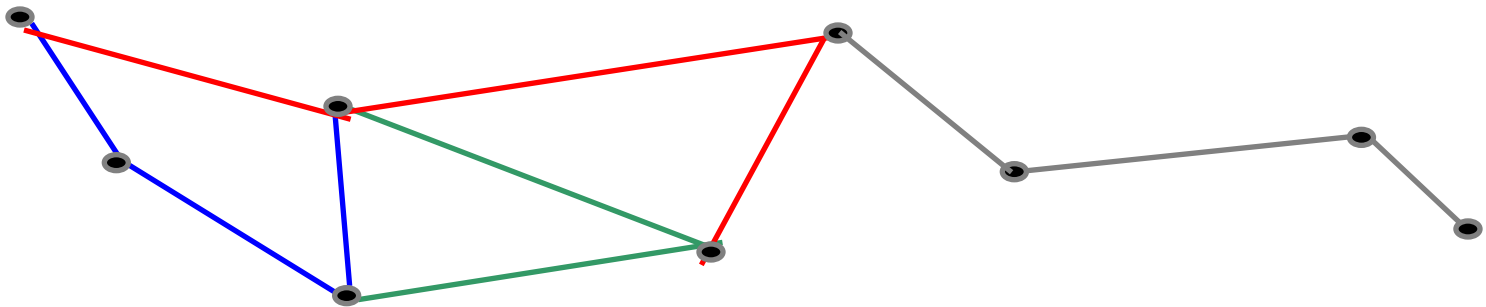


Conditions for Coordination

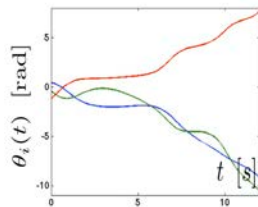
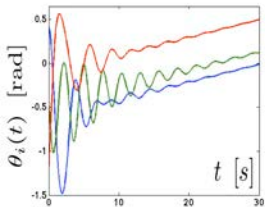
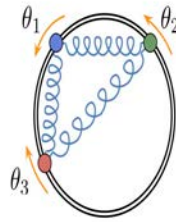
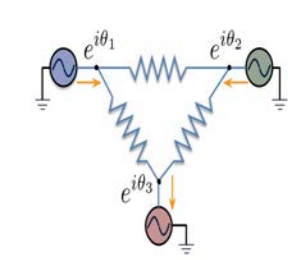
Difficulty: motion changes the network, and the network affects the motion

Theorem (Jadbabaie et al. 2003) *if sequence of graphs **jointly connected**, then all agents will asymptotically flock together.*

- Extensions to dynamic models, leader-follower and other formations, nonlinear dynamics, heterogeneous time delays, random networks,....
- Long history in parallel & distributed computation (Tsitsiklis '84, Seneta 77,81)
- Applications to: distributed optimization, statistics and inference, control, robotics, signal processing, opinion dynamics, social networks, gossip algorithms



Synchronization in Complex Networks



$$\frac{d\theta_i}{dt} = \omega_i + \frac{K}{N} \sum_{j=1}^N A_{ij} \sin(\theta_j - \theta_i)$$

Kuramoto, '1975

Swing equations in the grid
Flashing fire flies
Mexican waves
Rhythmic applause
Pacemaker cells & neurons

Effect of network structure?
Frequency mismatch?
Onset of synchronization?
Stability of synchronous state

Multi-agent Autonomous Platooning:

Sabau & Jadbabaie 2015-17



Can follow an uncooperative leader



Robust to presence of road disturbances



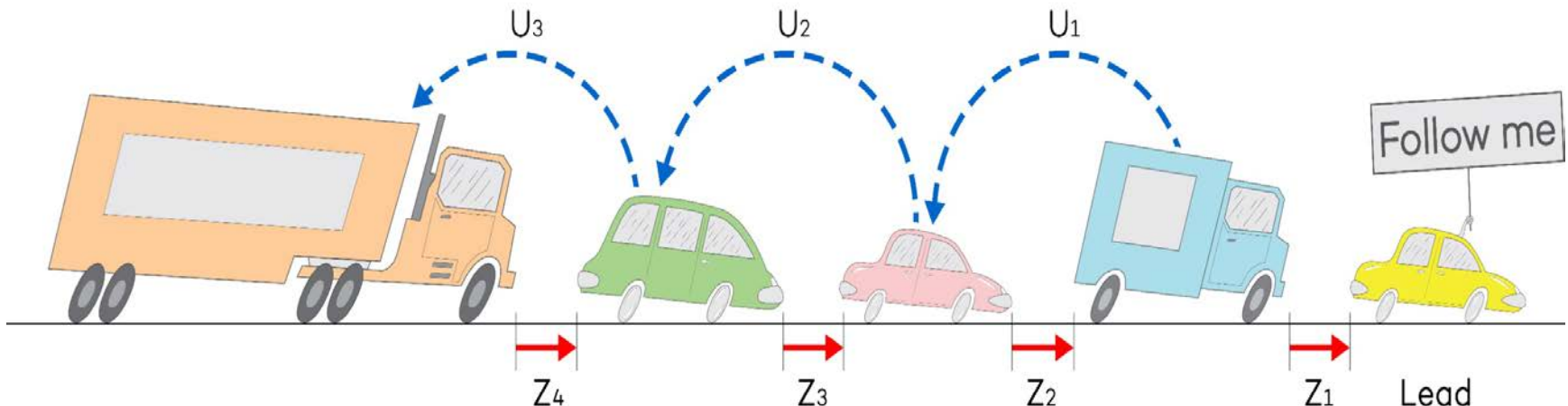
Guaranteed collision-free performance.



Eliminates the ``accordion'' effect (string instability)



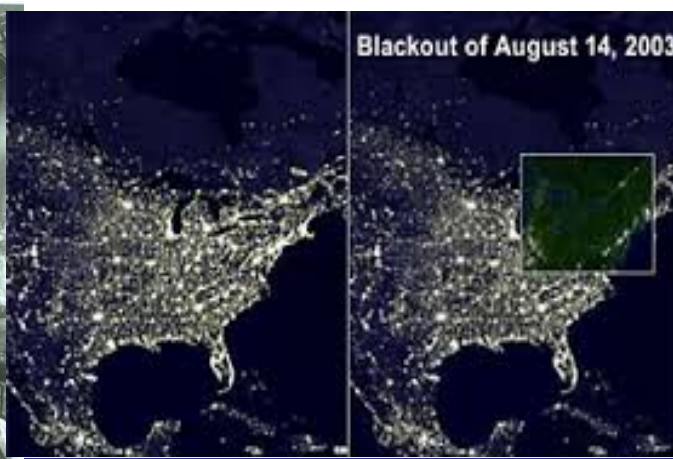
Compensate for Vehicle to Vehicle delays



Collective Behavior in Social Systems: Opinion Aggregation, Collective action, and Cascades

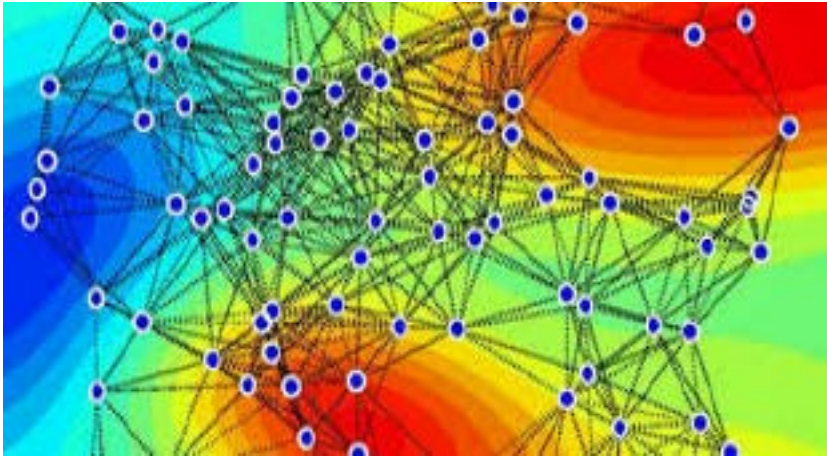


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Cascading failures: Why do shocks result in cascades?
How to intervene to maximize/minimize the spread?
What is the role of network structure? Popcorns vs dominos?

Social Learning and Distributed Inference



- How to combine opinions of peers (or sensor measurements) and observations to *learn* an unknown state?
- How does the network structure affect learning?



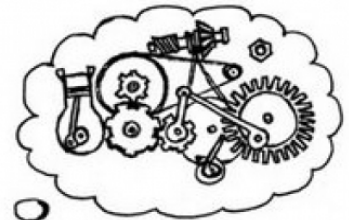
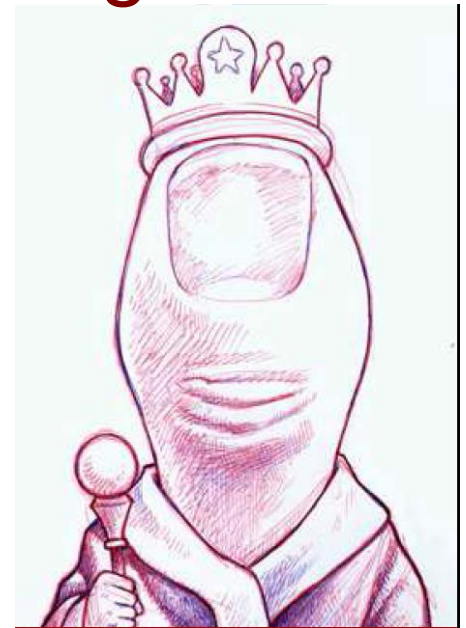
Social Learning and Opinion Pooling

- Opinions of individuals often influenced by private observations and opinion of friends, neighbors, ...
- How to combine private observations and peer opinions?



Rational vs. Rule of Thumb Learning Rules

- How to combine private observations and peer opinions?
 - Bayesian *Acemoglu et al. 2011, 2016, Mossel & Tamuz (2015)*
 - Non Bayesian: *DeGroot'74, Golub & Jackson 2010, Jadbabaie et al. 2012-14*
- Model how cognitive burden of Bayesian updating can lead to adhoc, rule of thumb rules



Social Learning & Rational Inference

Bayesian updating in networks: too much of a cognitive burden

Major departure: Imperfect Recall

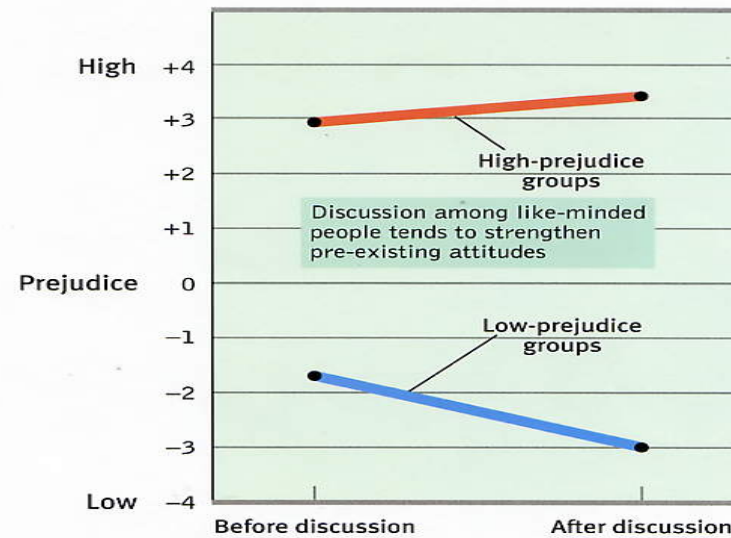
Take neighbors beliefs as sufficient statistic of what they know

$$\mu_{it+1} = \text{BU}\left(f_i(\mu_i^t); \omega_{it+1}\right), \mu_i^t = (\mu_{jt})_{j \in N_i},$$

- What properties should f_i have for learning to occur?
- How does the network structure affect learning?
- f_i can be linear, log-linear, but cannot have too much “logarithmic curvature”

Group Polarization and Learning

● **Group Polarization:** Tendency to shift to more extreme opinions after group discussion



Theorem (Molavi, Tahbaz-Salehi, Jadbabaie 2016): Networked societies
Subject to imperfect recall, learn if and only if opinions are non-polarizing.
Learning is non-generic, “knife-edge” and fragile.

Rational Decision Making in Organizations

- Experts/professional committees
- Examining witness testimonials
- Congressional debates
- Jury deliberations
- Medical diagnoses



Controlling Epidemics



Brockmann D, Helbing D. The hidden geometry of complex, network-driven contagion phenomena. *Science*. 2013.;342(6164):1337-42.

- How to stop Epidemics with limited resources?
- How to maximize spread of ideas / products ?
- What is the effect of network structure?

Controlling Epidemics in Networks

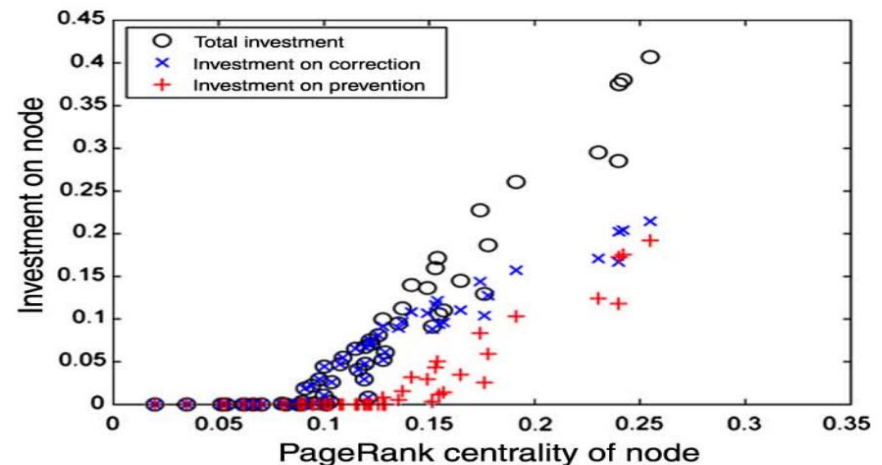
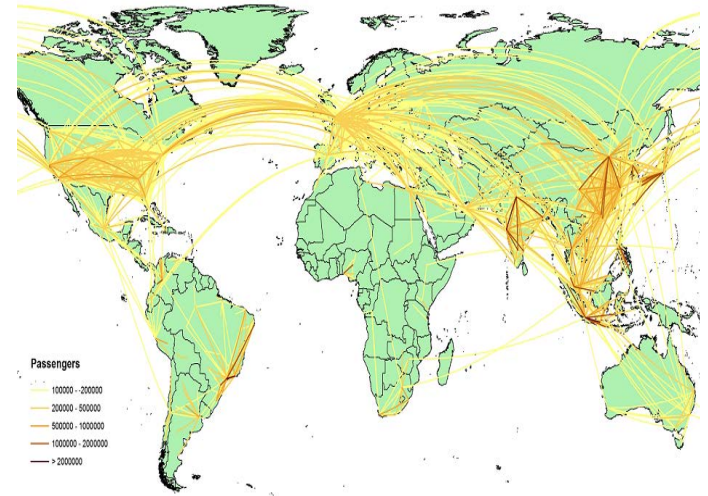
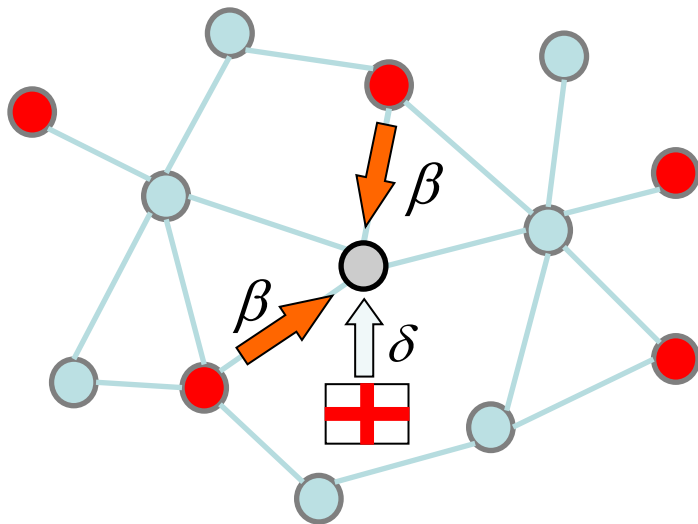
Key Question – how to optimally intervene to stop or maximize contagion ?

Existing control results:

- Centrality-based measures
- Greedy heuristics

Our approach: Geometric Programming

- Directed, weighted graphs
- Heterogeneous agents



Maximizing Spread: Quality or seeding ?

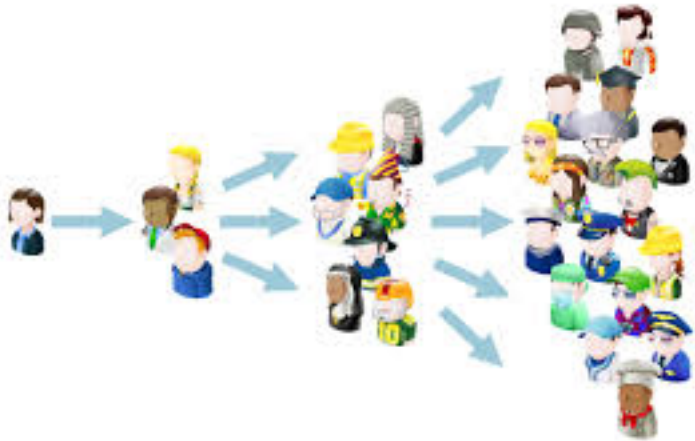
- Two parties compete for minds / ideas / market share in a network
 - Can “seed” a few initial infections (via persuasion, incentives, etc.)
 - Can invest in quality of product / message
 - parties have limited budget. Invest in quality or seeding?
- Subsequent contagion propagated via the network
- What happens when strategic firms play? s there an equilibrium? What role does structure play?



Maximizing Spread: Word of Mouth

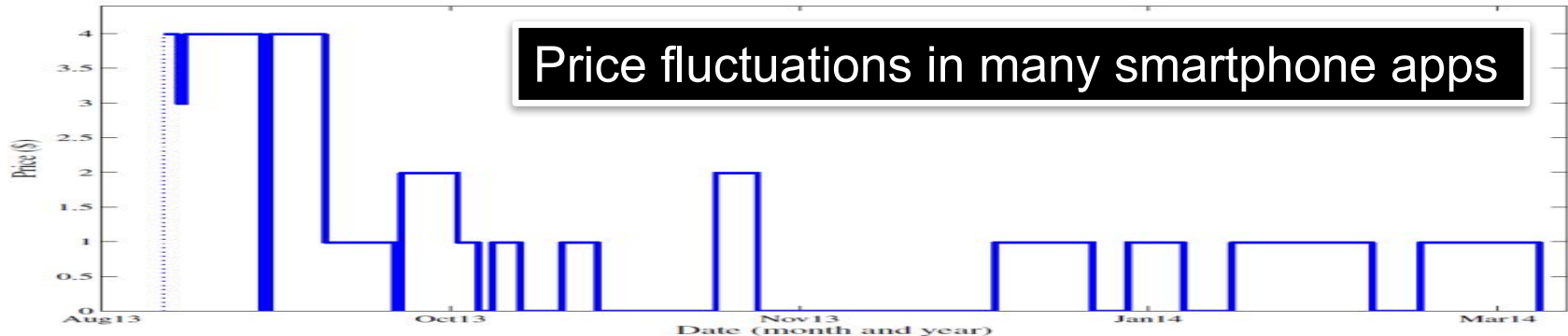
How to spread a product/idea in a large network?

Strategic Agents: Adopt the new technology when the price is right, and enough friends adopt

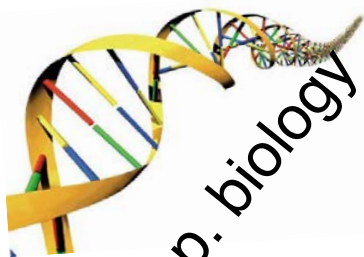


- If priced too low, many adopt, but no profit
- If priced too high, limited spread

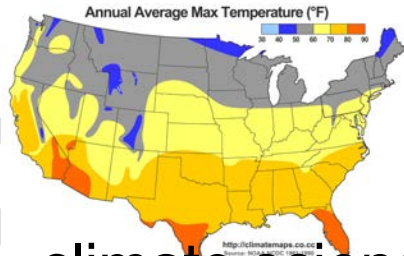
Theorem: Periodically sell at cost



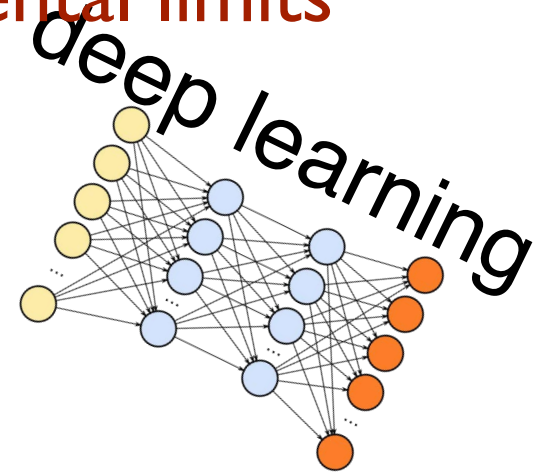
optimization for ML: fundamental limits



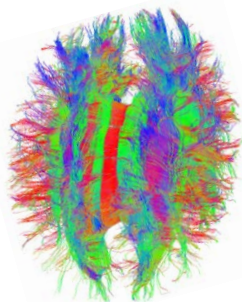
comp. biology



climate science



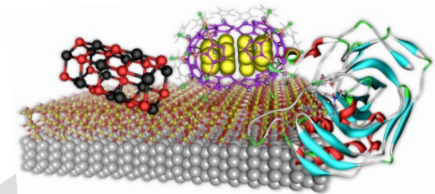
deep learning



medical imaging



semantic search



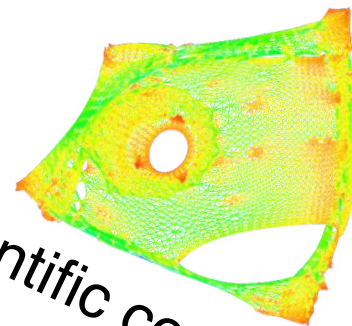
comp. material science



self-driving cars



scientific computing



Non-convex, incremental, distributed, second order methods

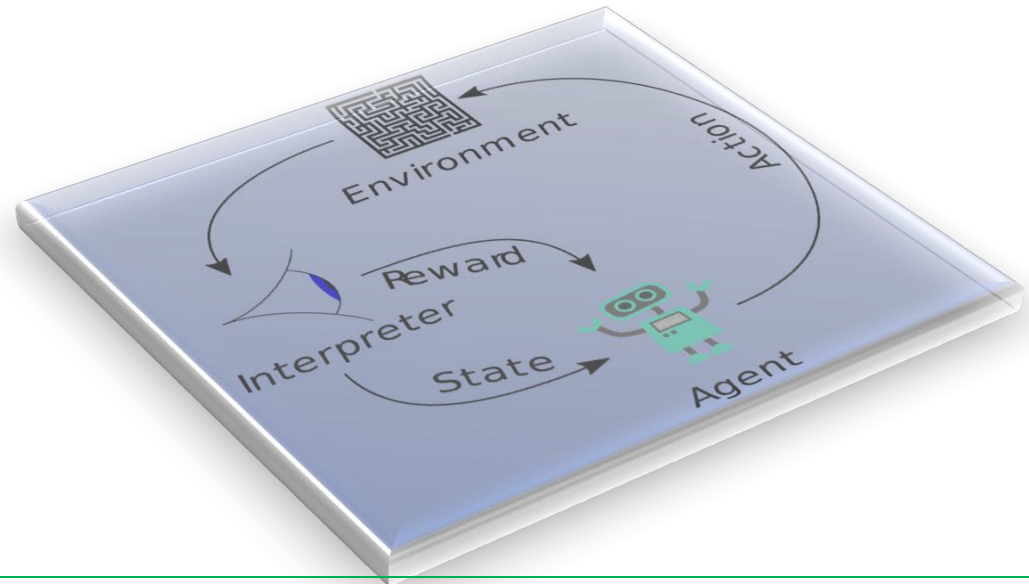
Empirical Risk Minimization and Online Learning

$$\min_{x \in \mathcal{M}} \mathbb{E}[F(x, \xi)]$$

$$\min_{x \in \mathcal{M}} \frac{1}{n} \sum_i f_i(x)$$

Optimizing Empirical Risk
for deep neural networks

Yun, Sra, Jadbabaie (2018)



General online protocol:

- Observe some side information
- Make a prediction / decision
- Observe data/outcome (or some partial information)
- Adjust the model

Autonomous Transportation + Urbanization

social potential of driverless cars.

Source: Bank of Mellon NY

Advantages

Safety
90%

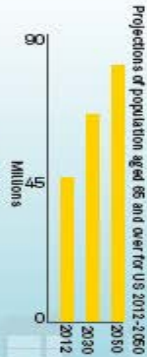
of road traffic accidents are currently caused by human error.¹

Driverless cars could translate into approximately **36,000 lives saved each year** and approximately **US\$ 488 billion.**

Social
Greater mobility provided to elderly and disabled people.

People over the age of 65 expected to double in US by **2050.**²

Economic
Morgan Stanley estimates Autonomous cars will result in **US\$ 1.3 trillion** savings every year for the US economy, globally this translates into **US\$ 5.6 trillion.**³



Economic benefits for drivers

- Fuel costs
- Productivity gains
- Accident costs

Potential obstacles

Liability
Who accepts responsibility in the case of an accident?

Legislation
US infrastructure deficiencies mean that **US\$ 10.8 billion** had to be found to keep the Highway Trust Fund solvent until May 2015.⁴

Consumer adoption

Once people accept and trust the systems, adoption rates are expected to climb.

Welcome to the autonomous car

2030

1 Society, 18 December 2013.
2 January 2015.

3 Source: "House Approves Highway Trust Fund patch," CBS 18 December 2013.
<http://www.cbsnews.com/news/house-approves-highway>

4 Investment Managers are appointed by BNY Mellon Investment Management Limited ("BNYMIM EMEA") or affiliated fund operators undertake portfolio management activities in relation to and services entered into by clients with BNYMIM EMEA or

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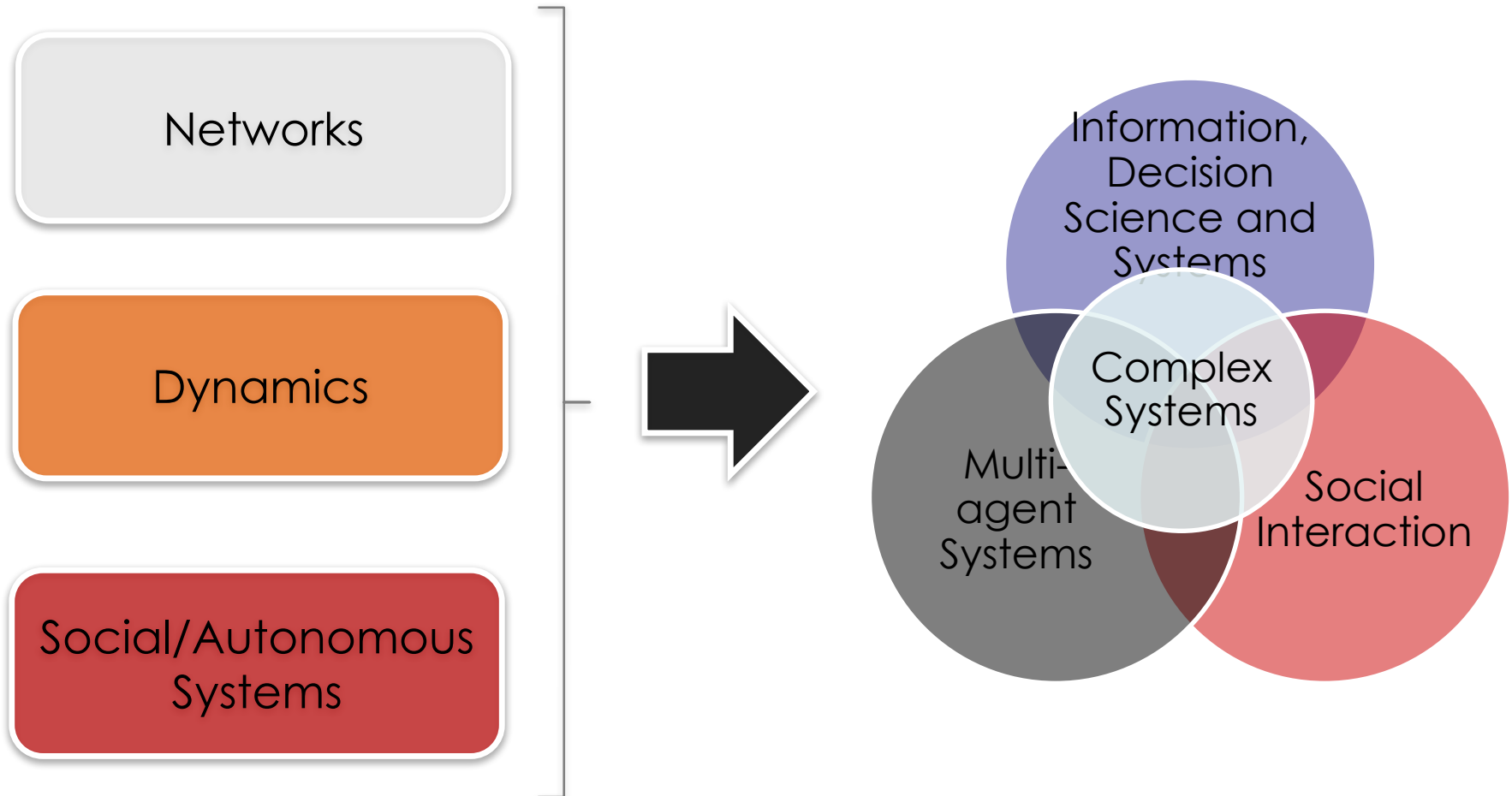
Social Behavior of drivers,
Institutional behavior
Of markets/regulators

COMING : HOME DELIVERY BY DRONE...



DAVE GRANLUND © www.davegranlund.com

Summary



**MIT INSTITUTE FOR DATA,
SYSTEMS, AND SOCIETY**



IDSS

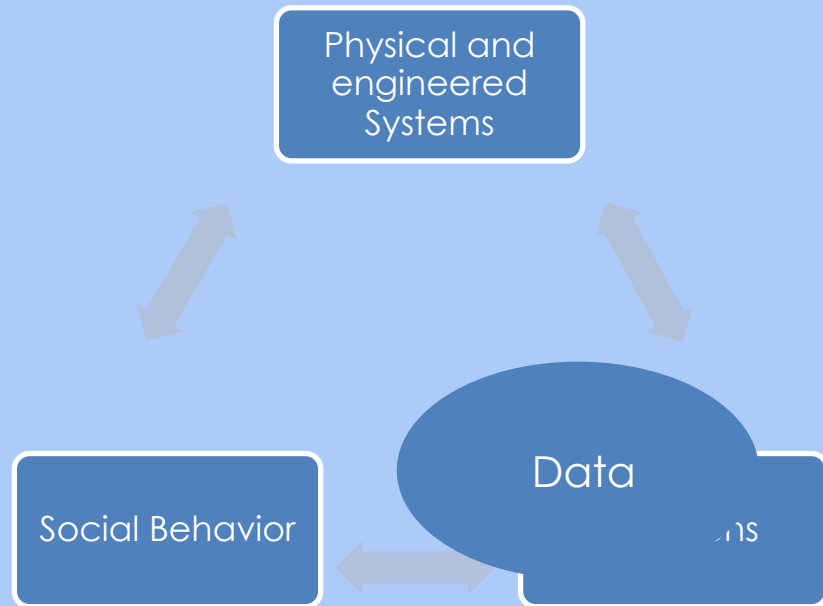
January, 2018

Our society depends on ***massively complex, data-rich systems***. Increasingly, the answers to society's most serious challenges lie in ***the ability to extract from these systems and vast data sets new patterns, new pathways, and new solutions***.

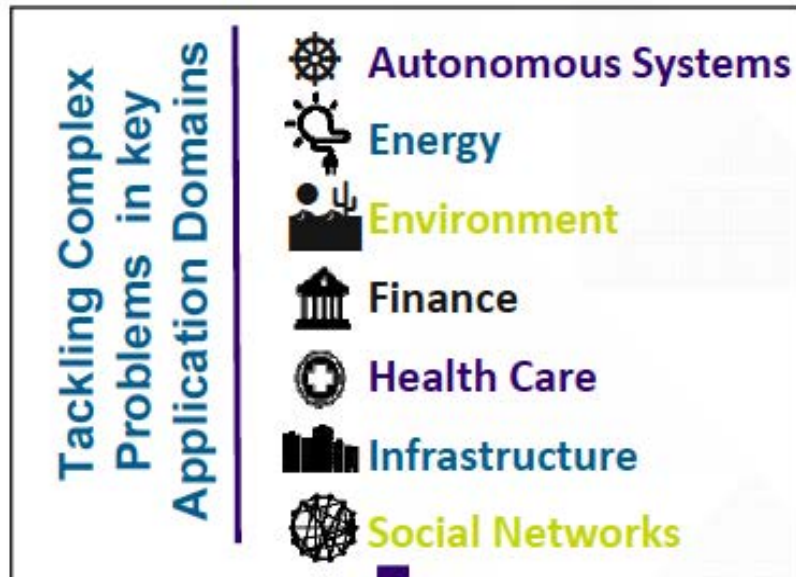
- MIT President Rafael Reif



“Addressing **complex societal challenges** by advancing education and research at the intersection of **statistics, data science, information and decision systems**, and **social sciences**.”

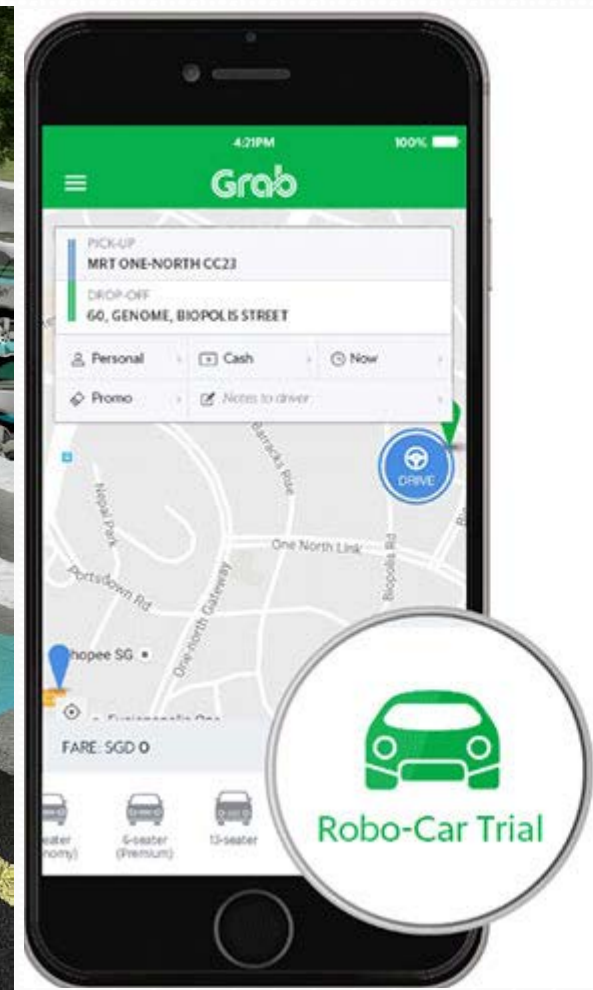
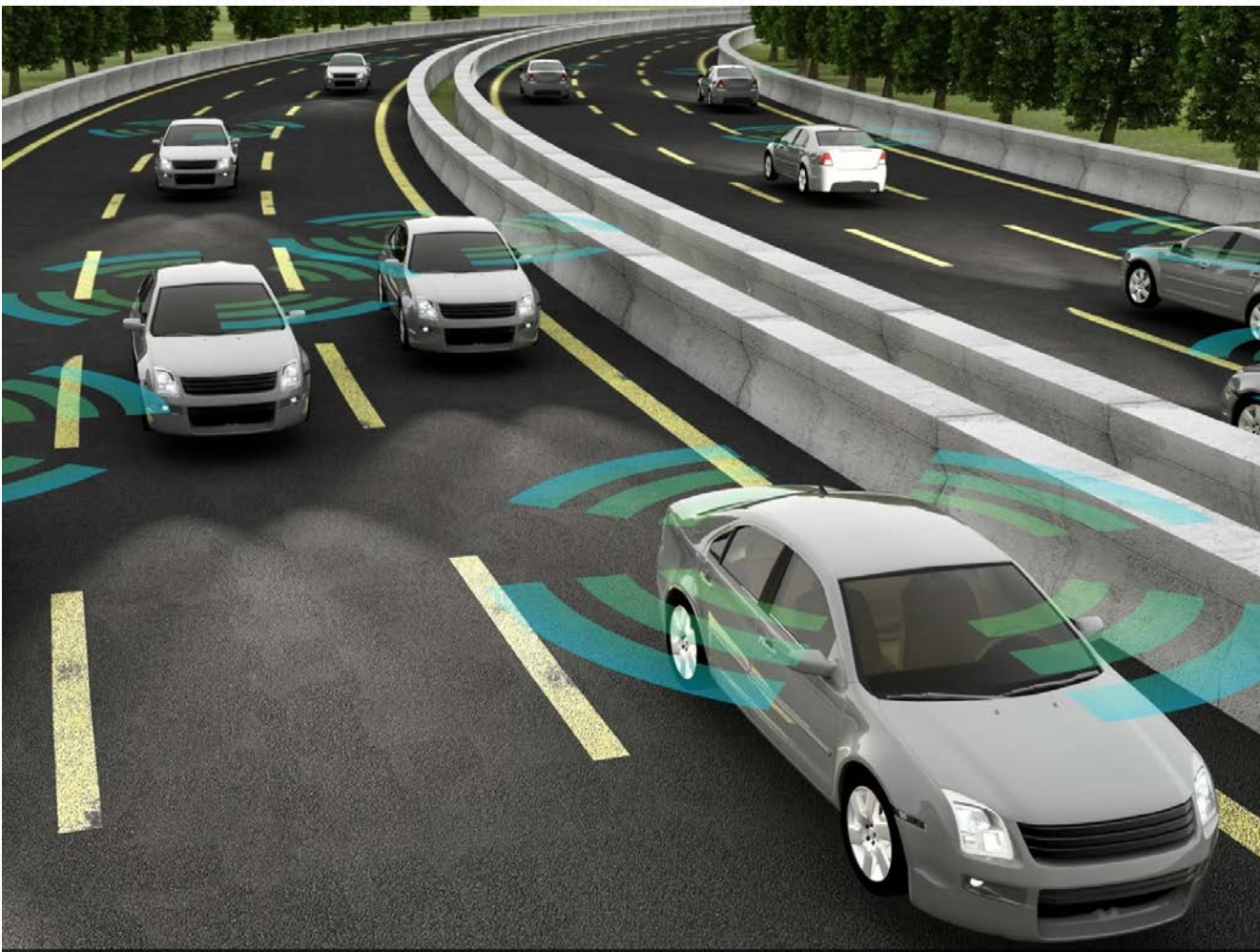


MULTI-DISCIPLINARY RESEARCH THAT ADDRESSES COMPLEX SOCIETAL CHALLENGES

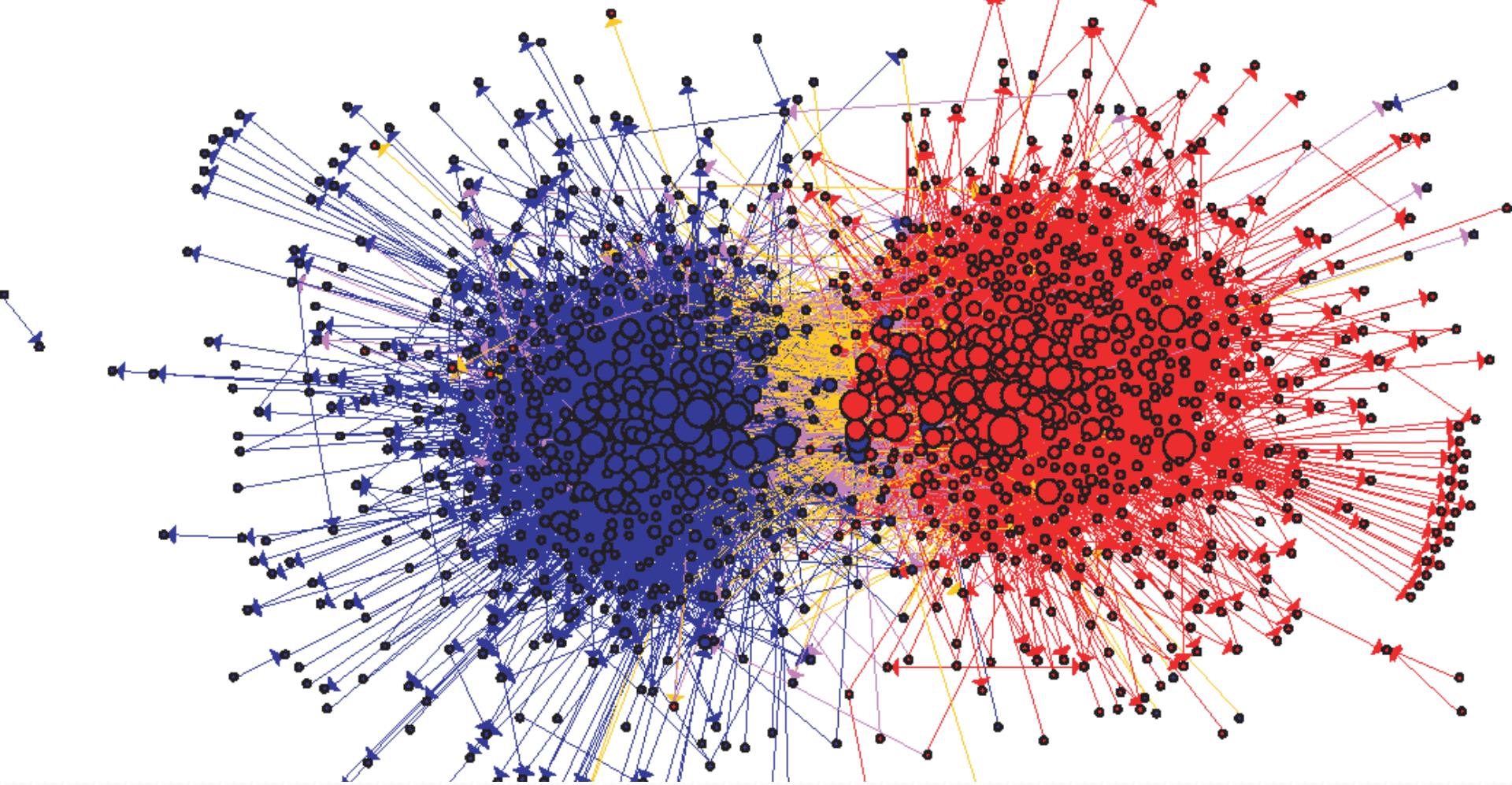


Our ability to understand data and develop actionable models across interconnected systems, is at the core of being able to uncover new insights and solutions.

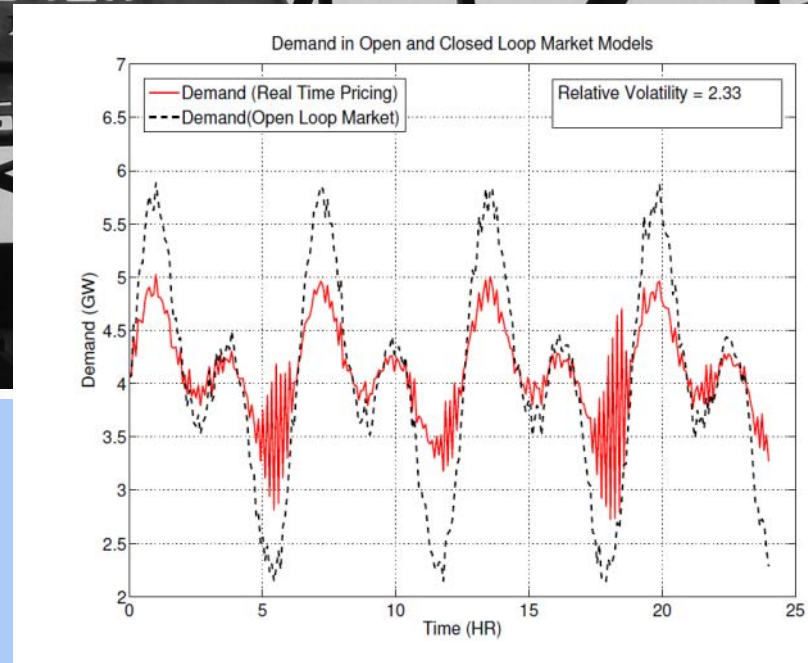
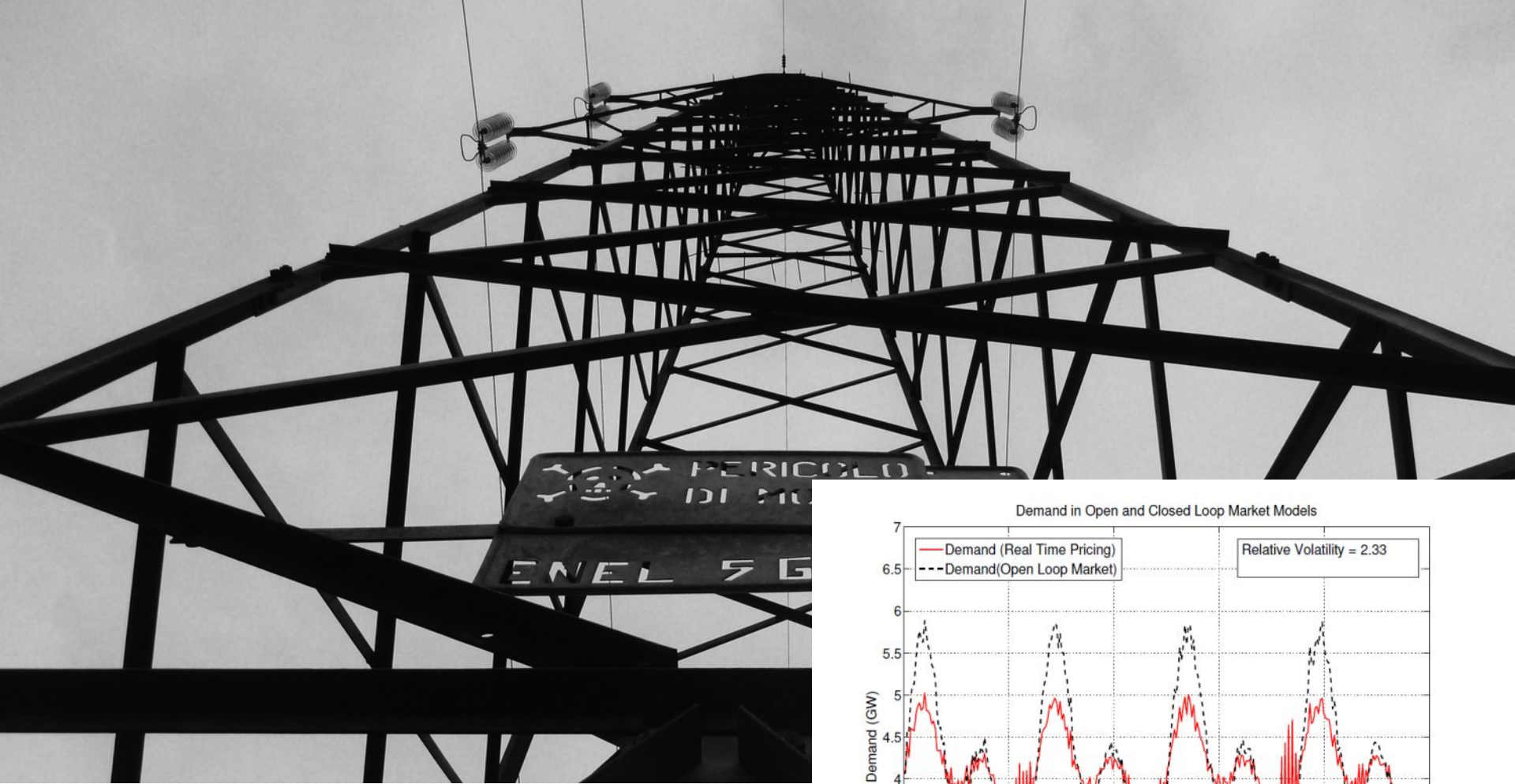




Connected, Self-Driving Cars Present New Solutions and Challenges



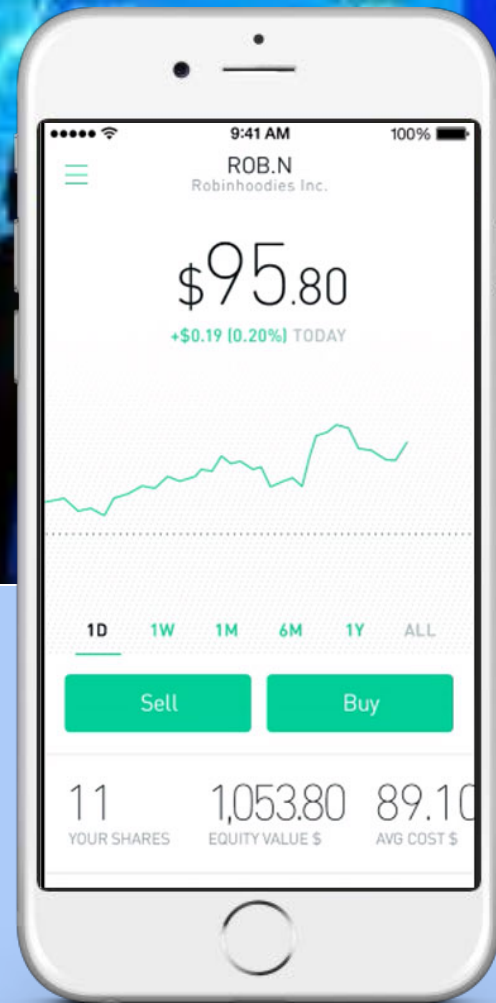
**Social Media is Transforming the Way We Share
--- and Consume -- Information**



Power Systems and Pricing Increasingly Complex



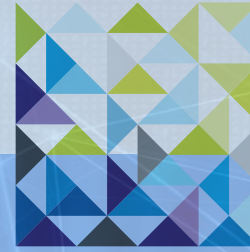
How Do We Understand Systemic Risk in a Global, Interconnected Market?





How Can We Predict Choices and Behavior to Develop Better Goods and Services?

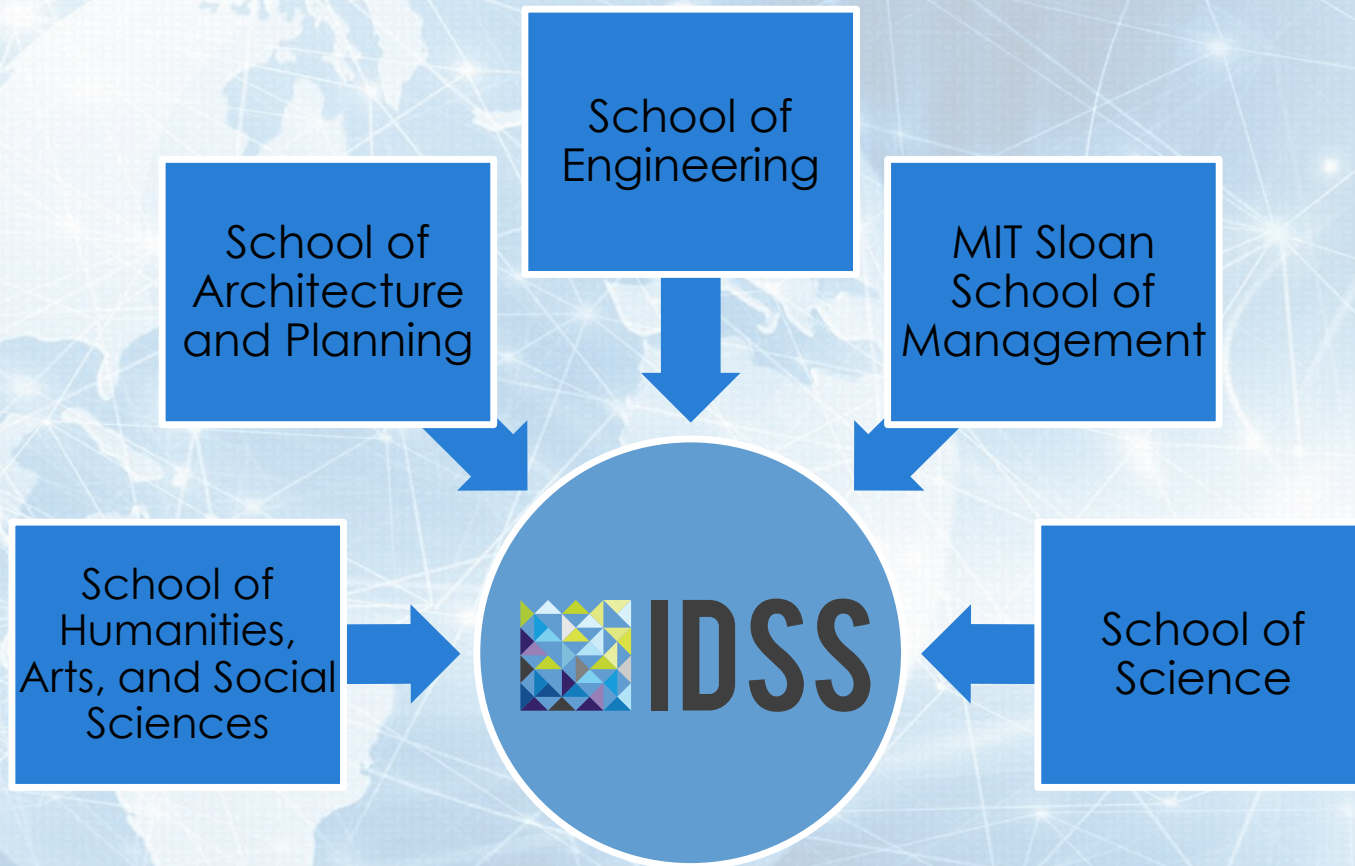
**MIT INSTITUTE FOR DATA,
SYSTEMS, AND SOCIETY**



IDSS

Teaching and Research

IDSS Has the Support and Involvement of All Five Schools of MIT



- **Founded in 2015**
- **Interdisciplinary research at the core of its mission**
- **Supported by more than 80 core and affiliate faculty from across MIT**

IDSS Advances Both Teaching and Research

Teaching

- PhD, Social and Engineering Systems (SES) (40 students and growing)
- PhD in Statistics (planned 2018)
- Masters in Technology Policy (~120 students)
- Statistics Minor for Undergraduates
- MicroMasters in Data Science (planned 2018)
- Professional Education Courses (Online)

Research

- IDSS SES PhD Program
- Laboratory in Information and Decision Systems (LIDS)
- Statistics and Data Science Center
- Sociotechnical Systems Research Center

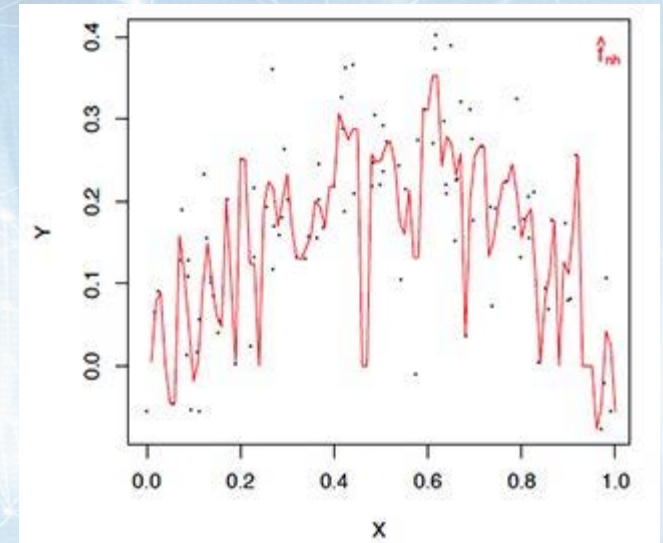
Teaching: The PhD in Social and Engineering Systems is in its Second Year with a total of 20 students

- SES PhD students study problems that correspond to significant societal challenges, in areas such as autonomous systems, energy systems, finance, social networks, and urban systems.
- Work includes analytical research that can be used to inform policy making.
- Graduates will go on to roles in academic departments in various fields, serve in the public sector (from research labs to regulatory agencies), as well as pursue careers in the private sector (from industry to consulting).



Teaching: The PhD in Statistics and Data Science to Launch in Fall, 2018

- The first Statistics and Data Science PhD at MIT, with an emphasis on 21st Century Statistics
- Interdisciplinary: aspiring students will apply through a “home department” such as Economics or Math.
- The program will offer students advanced training in statistics and data science, in combination with the strength and guidance of established departments at MIT.



Teaching: The Masters in Technology Policy is a 2-year Program, Comprising about 60 Students

- TPP addresses societal challenges through research and education at the intersection of technology and policy.
- Through our two-year Master of Science Program, we prepare students for leadership in government and industry.
- Degree requirements include graduate-level coursework in engineering, policy, economics and law, and an interdisciplinary thesis focused on a technology policy issue.



Teaching: An Online Course Designed for Professional Learners

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- Featuring top faculty members at MIT
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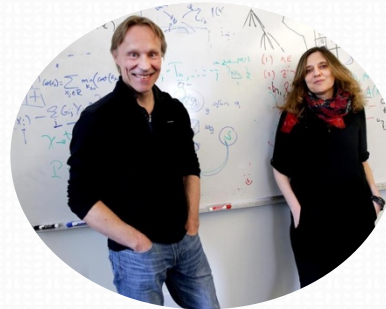
An Online Data Science MicroMasters



John Tsitsiklis



Philippe Rigollet



Regina Barzilay &
Tommi Jaakkola



Esther Duflo
& Sara Ellison

Probability:
6.431x

Statistics:
18.6501x

Machine Learning
6.86x

Data Analysis
14.310x

- Four of MIT's highly courses, online
- Open to employees, students worldwide for free
 - Verified certificates available for a small fee (~\$300/course)
- Classes start fall 2019



IDSS + **edX**



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- Employee Opportunities
 - Efficient investment in employee training
 - Boost retention, productivity
- Robust Verification of students
 - ID Verified Students
 - Proctored Exams
 - Anti-Cheating Algorithms





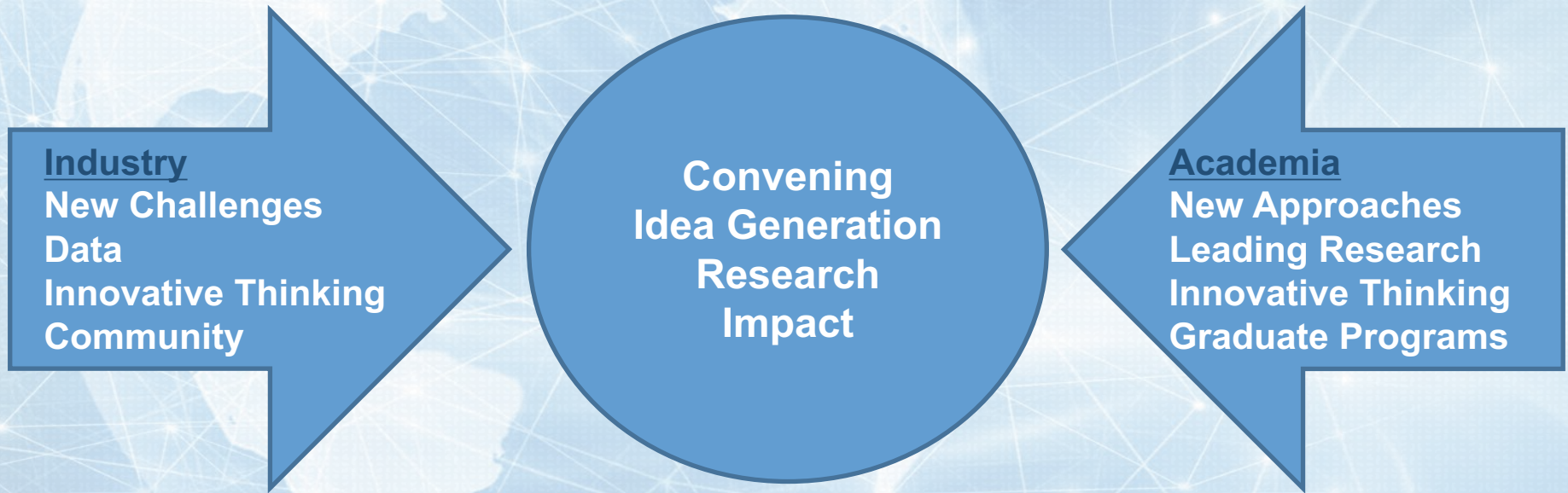
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strategy and technology consultants

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Industry Partnerships Support Research, Collaboration and Impact



The Industry Partnership Program Offers a Relationship with IDSS Faculty and Students Across Areas of Interest to the Industry

- Meet with faculty and graduate students and identify opportunities to incorporate members' data assets into graduate student research projects
- Attend members-only workshops in the fall and the spring, for small group discussion with faculty, researchers and students
- Meet and interact with IDSS graduate students for the purpose of recruiting, both for internships and positions after graduation
- Join the IDSS External Advisory Board, which offers strategic advice and guidance to IDSS
- Attend conferences and workshops sponsored by IDSS and the Laboratory for Information and Decision Sciences (LIDS) as VIP guests
- Attend monthly seminars, hosted by IDSS, featuring MIT researchers, visiting faculty, and guests from industry and top universities