

Announcements

- HW 3 postponed to this Thursday
- Project proposal due this Thursday as well

CS6501:Topics in Learning and Game Theory (Fall 2019)

Bayesian Persuasion

Instructor: Haifeng Xu

- Prediction markets and peer prediction study **how to elicit information** from others
- This lecture: when you have information, **how to exploit it?**
 - Relevant to mechanism design

Outline

- Introduction and Bayesian Persuasion
- Algorithms for Bayesian Persuasion

Two Primary Ways to Influence Agents' Behaviors

- Design/provide incentives
 - Auctions



Two Primary Ways to Influence Agents' Behaviors

- Design/provide incentives
 - Auctions
 - Discounts/coupons

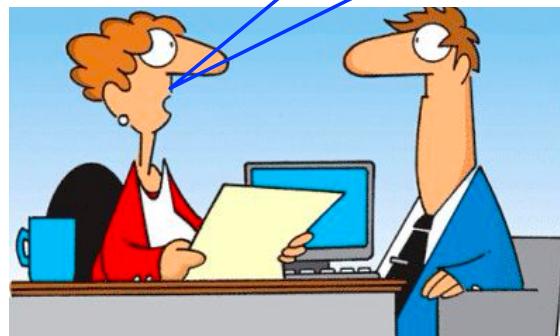


Two Primary Ways to Influence Agents' Behaviors

- Design/provide incentives

- Auctions
- Discounts/coupons
- Job contract design

Bonus depends on performance, and is up to \$1M!



Two Primary Ways to Influence Agents' Behaviors

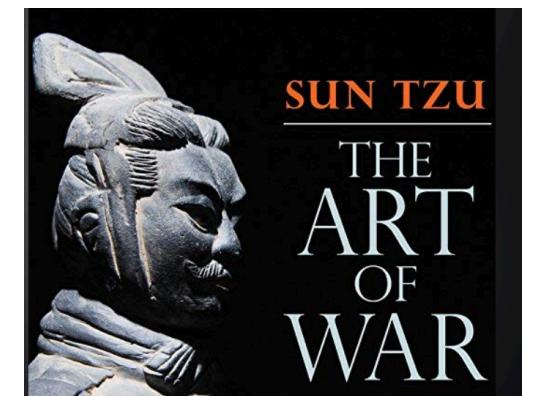
- Design/provide incentives
 - Auctions
 - Discounts/coupons
 - Job contract design
- Influence agents' beliefs
 - Deception in wars/battles



Mechanism Design

All warfare is based on deception. Hence, when we are able to attack, we must seem unable; when using our forces, we must appear inactive...

-- Sun Tzu, *The Art of War*



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Mechanism Design

- Influence agents' beliefs
 - Deception in wars/battles
 - Strategic information disclosure

Strategic inventory
information disclosure

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Size:

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Size Chart

Color: White Blackware



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Two Primary Ways to Influence Agents' Behaviors

- Design/provide incentives
 - Auctions
 - Discounts/coupons
 - Job contract design
- Influence agents' beliefs
 - Deception in wars/battles
 - Strategic information disclosure
 - News articles, advertising, tweets, etc.



Mechanism Design

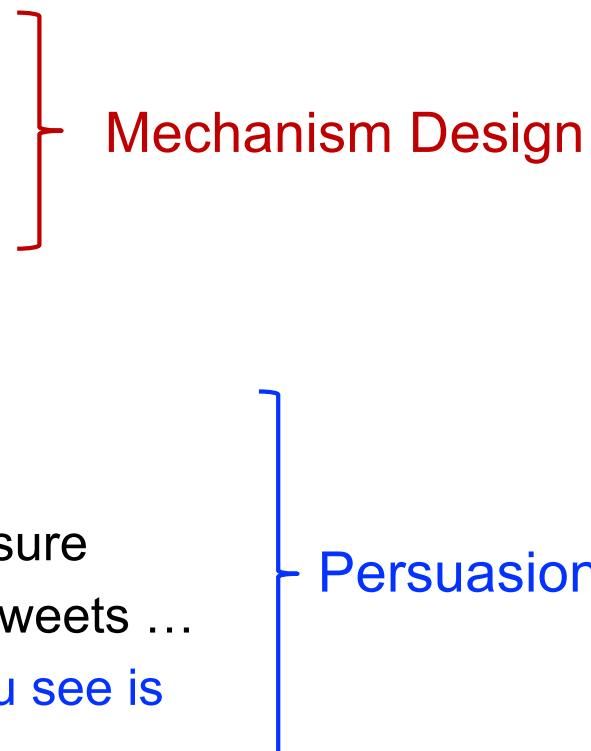


InsideEVs Tesla Pickup Truck Render Looks Bold, Sinister And Bad In Black Dressed in all black, this Tesla pickup truck render had a certain 'bad' appearance to it. It surely is bold and the black hue gives it a sinister look ... 3 days ago 

Business Insider Nordic Elon Musk repeatedly insults lawyer during bizarre deposition Elon Musk called the lawyer who interviewed him for a Tesla shareholder lawsuit 'a bad human being' and other insults during a bizarre ... 6 days ago 

CNET Tesla's Model 3 is great to drive, but what's it like to own? Not as bad as we'd have expected, actually. Tesla's quality seems to have improved more or less steadily since hitting a low point this year in ... 6 days ago 

Two Primary Ways to Influence Agents' Behaviors

- Design/provide incentives
 - Auctions
 - Discounts/coupons
 - Job contract design
 - Influence agents' beliefs
 - Deception in wars/battles
 - Strategic information disclosure
 - News articles, advertising, tweets ...
 - In fact, most information you see is there for a goal
- 
- Mechanism Design
- Persuasion

Persuasion is the act of exploiting an informational advantage in order to influence the decisions of others

- Intrinsic in human activities: advertising, negotiation, politics, security, marketing, financial regulation,...
- A large body of research

One Quarter of GDP Is Persuasion

*By DONALD McCLOSKEY AND ARJO KLAMER**

— The American Economic Review Vol. 85, No. 2, 1995.

Example: Recommendation Letters



- Advisor vs. recruiter
- 1/3 of the advisor's students are **excellent**; 2/3 are **average**
- A fresh graduate is randomly drawn from this population
- Recruiter
 - Utility $1 + \epsilon$ for hiring an excellent student; -1 for an average student
 - Utility 0 for not hiring
 - A-priori, only knows the advisor's student population

$$(1 + \epsilon) \times 1/3 - 1 \times 2/3 < 0$$

hiring

Not hiring

Example: Recommendation Letters



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 - Utility $1 + \epsilon$ for hiring an excellent student; -1 for an average student
 - Utility **0** for not hiring
 - A-priori, only knows the advisor's student population
- Advisor
 - Utility 1 if the student is hired, 0 otherwise
 - Knows whether the student is excellent or not

Example: Recommendation Letters



What is the advisor's optimal “recommendation strategy”?

- Attempt 1: always say “excellent” (equivalently, no information)
 - Recruiter ignores the recommendation
 - Advisor expected utility 0

Remark

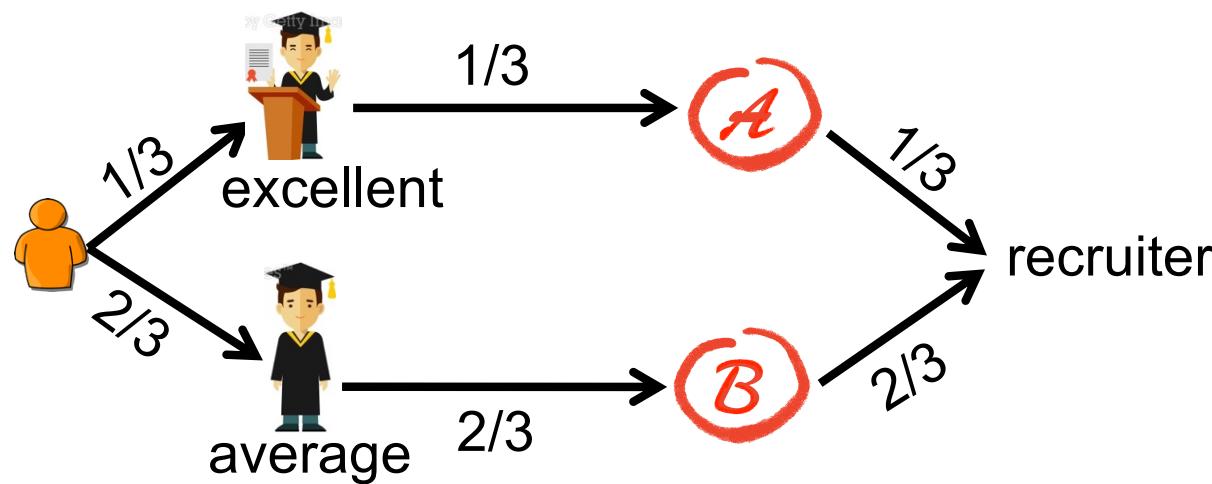
Advisor commitment: cannot deviate and recruiter knows his strategy

Example: Recommendation Letters



What is the advisor's optimal “recommendation strategy”?

- Attempt 2: honest recommendation (i.e., full information)
 - Advisor expected utility $1/3$

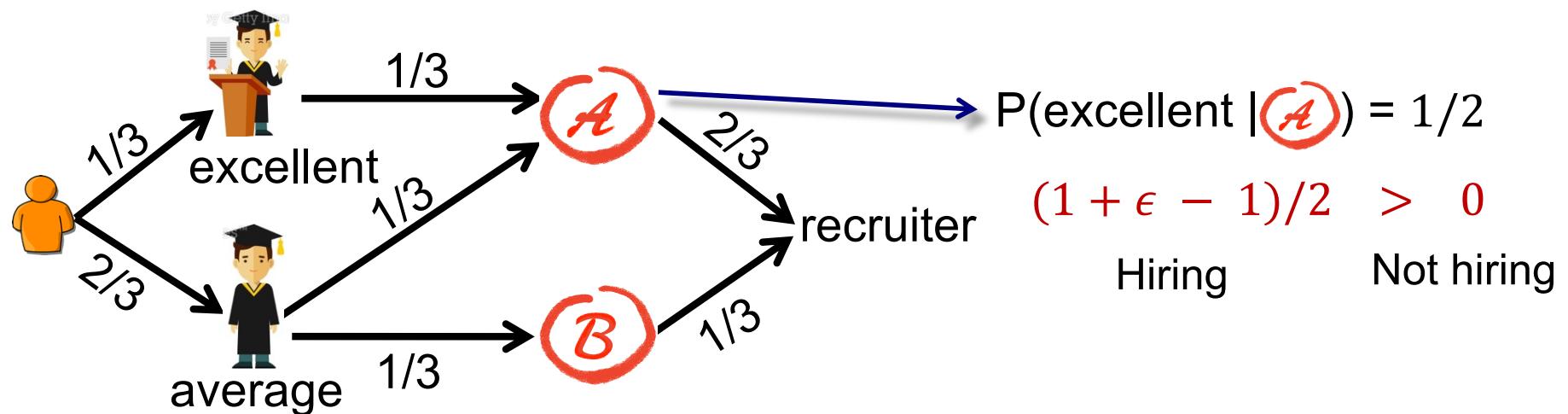


Example: Recommendation Letters



What is the advisor's optimal "recommendation strategy"?

- Attempt 3: noisy information → advisor expected utility $2/3$



Model of Bayesian Persuasion

- Two players: persuader (**Sender, she**), decision maker (**Receiver he**)
 - Previous example: advisor = sender, recruiter = receiver
- Receiver looks to take an action $i \in [n] = \{1, 2, \dots, n\}$
 - Receiver utility $r(i, \theta)$ $\theta \in \Theta$ is a random **state of nature**
 - Sender utility $s(i, \theta)$
- Both players know $\theta \sim \text{prior dist. } \mu$, but Sender has an **informational advantage** – she can observe realization of θ
- Sender wants to strategically reveal info about θ to “persuade” Receiver to take an action she likes
 - Concealing or revealing all info is not necessarily the best

Well...how to reveal **partial** information?

Revealing Information via Signaling

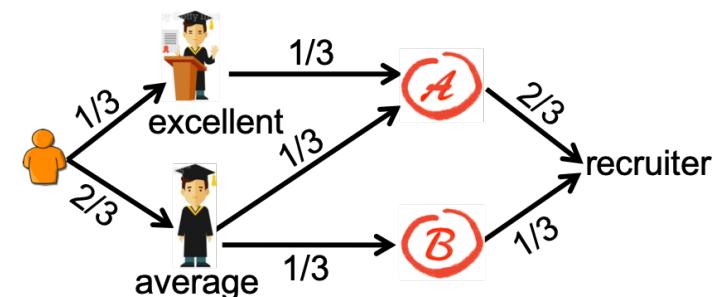
Definition: A signaling scheme is a mapping $\pi: \Theta \rightarrow \Delta_\Sigma$ where Σ is the set of all possible signals.

π is fully described by $\{\pi(\sigma, \theta)\}_{\theta \in \Theta, \sigma \in \Sigma}$ where $\pi(\sigma, \theta) = \text{prob. of sending } \sigma \text{ when observing } \theta$ (so $\sum_{\sigma \in \Sigma} \pi(\sigma, \theta) = 1$ for any θ)

Note: scheme π is always assumed public knowledge, thus known by Receiver

Example

- $\Theta = \{Excellent, Average\}$, $\Sigma = \{A, B\}$
- $\pi(A, Average) = 1/2$



Revealing Information via Signaling

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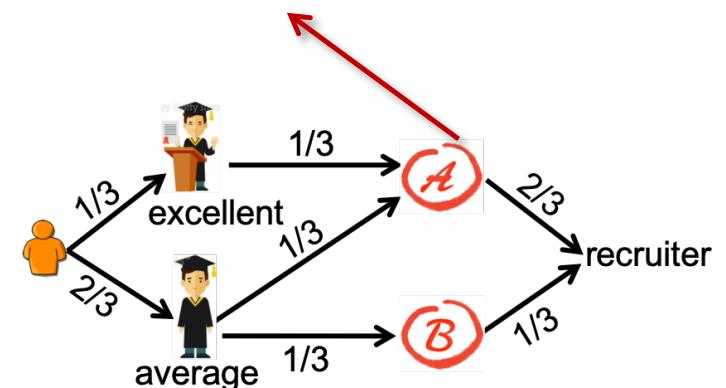
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What can Receiver infer about θ after receiving σ ?

Bayes updating:

$$\Pr(\theta | \sigma) = \frac{\pi(\sigma, \theta) \cdot \mu(\theta)}{\sum_{\theta'} \pi(\sigma, \theta') \cdot \mu(\theta')}$$

$$\Pr(\text{excellent} | A) = 1/2$$



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Would such noisy information benefit Receiver?

- Expected Receiver utility conditioned on σ :

$$R(\sigma) = \max_{i \in [n]} \left[\sum_{\theta \in \Theta} r(i, \theta) \cdot \frac{\pi(\sigma, \theta) \cdot \mu(\theta)}{\sum_{\theta'} \pi(\sigma, \theta') \cdot \mu(\theta')} \right]$$

- $\Pr(\sigma) = \sum_{\theta'} \pi(\sigma, \theta') \cdot \mu(\theta')$

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- $\Pr(\sigma) = \sum_{\theta'} \pi(\sigma, \theta') \cdot \mu(\theta')$
 - $\Pr(\sigma) \cdot R(\sigma) = \max_i \sum_{\theta \in \Theta} r(i, \theta) \cdot \pi(\sigma, \theta) \cdot \mu(\theta)$
- Expected Receiver utility under π : $\sum_{\sigma} \max_i \sum_{\theta \in \Theta} r(i, \theta) \cdot \pi(\sigma, \theta) \cdot \mu(\theta)$

Revealing Information via Signaling

Fact. Receiver's expected utility (weakly) increases under any signaling scheme π .

Proof:

- Expected Receiver utility without information: $\max_i \sum_{\theta \in \Theta} r(i, \theta) \cdot \mu(\theta)$
- Expected Receiver utility under π : $\sum_{\sigma} \max_i \sum_{\theta \in \Theta} r(i, \theta) \cdot \pi(\sigma, \theta) \cdot \mu(\theta)$

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- Let $i^* = \arg \max_i \sum_{\theta \in \Theta} r(i, \theta) \cdot \mu(\theta)$, we have

$$\begin{aligned}\sum_{\sigma} \max_i \sum_{\theta \in \Theta} r(i, \theta) \cdot \pi(\sigma, \theta) \cdot \mu(\theta) &\geq \sum_{\sigma} \sum_{\theta \in \Theta} r(i^*, \theta) \cdot \pi(\sigma, \theta) \cdot \mu(\theta) \\ &= \sum_{\theta \in \Theta} r(i^*, \theta) \cdot [\sum_{\sigma} \pi(\sigma, \theta)] \cdot \mu(\theta) \\ &= \sum_{\theta \in \Theta} r(i^*, \theta) \cdot \mu(\theta)\end{aligned}$$

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Revealing Information via Signaling

Fact. Receiver's expected utility (weakly) increases under any signaling scheme π .

Remarks:

- Signaling scheme does increase Receiver's utility
- More (even noisy) information always helps a decision maker (DM)
 - Note true if multiple DMs (will see examples later)

Corollary. Receiver's expected utility is maximized when Sender reveals full info, i.e., directly revealing the realized θ .

Because any other noisy scheme π can be improved by further revealing θ itself

Revealing Information via Signaling

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But this is not Sender's goal...

Sender Objective: carefully pick π to maximize her expected utility

Outline

- Introduction and Bayesian Persuasion
- Algorithms for Bayesian Persuasion

Q: What worries you the most when designing $\pi = \{\pi(\theta, \sigma)\}_{\theta \in \Theta, \sigma \in \Sigma}$?

- Don't know what is the set of all possible signals Σ ...
- Like in mechanism design, too many signals to consider in this world
 - Again, you can use "looking 45° up to the sky" as a signal
- **Key observation:** a signal is mathematically nothing but a posterior distribution over Θ
 - Recall the Bayes updates: $\Pr(\theta | \sigma) = \frac{\pi(\sigma, \theta) \cdot \mu(\theta)}{\sum_{\theta'} \pi(\sigma, \theta') \cdot \mu(\theta')}$
- It turns out that n signals suffice

Revelation Principle

Fact. There always exists an optimal signaling scheme that uses at most n ($= \#$ receiver actions) signals, where signal σ_i induce optimal Receiver action i

➤ Conditioned on any signal σ

- Receiver infers $\Pr(\theta|\sigma) = \frac{\pi(\sigma, \theta) \cdot \mu(\theta)}{\sum_{\theta} \pi(\sigma, \theta') \cdot \mu(\theta')}$
- Receiver takes optimal action $i^* = \arg \max_{i \in [n]} \sum_{\theta} \Pr(\theta|\sigma) r(i, \theta)$

➤ Now, if signal σ and σ' result in the same optimal action i^* , Sender can instead send a new signal $\sigma_{i^*} = (\sigma, \sigma')$ in both cases

- Claim: i^* is still the optimal action conditioned on σ_{i^*}

$$\sum_{\theta} \Pr(\theta|\sigma) r(i^*, \theta) \geq \sum_{\theta} \Pr(\theta|\sigma) r(i, \theta), \quad \forall i$$

$$\sum_{\theta} \Pr(\theta|\sigma') r(i^*, \theta) \geq \sum_{\theta} \Pr(\theta|\sigma') r(i, \theta), \quad \forall i$$

$$\begin{aligned} \Rightarrow & \sum_{\theta} [\Pr(\theta|\sigma)p + \Pr(\theta|\sigma')(1-p)] r(i^*, \theta) \\ & \geq \sum_{\theta} [\Pr(\theta|\sigma)p + \Pr(\theta|\sigma')(1-p)] r(i, \theta), \quad \forall i \end{aligned}$$

$\Pr(\theta|\sigma^*)$ is a convex combination of $\Pr(\theta|\sigma)$ and $\Pr(\theta|\sigma')$

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- Claim: i^* is still the optimal action conditioned on σ_{i^*}
- Both players' utilities did not change as receiver still takes i^* as Sender wanted

➤ Can merge all signals with optimal receiver action i^* as a single signal σ_{i^*}

Revelation Principle

Fact. There always exists an optimal signaling scheme that uses at most n ($= \# \text{ receiver actions}$) signals, where signal σ_i induce optimal Receiver action i

- Each σ_i can be viewed as an **action recommendation** of i

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Optimal Persuasion via Linear Program

➤ Input: prior μ , sender payoff $s(i, \theta)$, receiver payoff $r(i, \theta)$

➤ Variables: $\pi(\sigma_i, \theta)$

Sender expected utility
(we know Receiver will take i at signal σ_i)



$$\max \left[\sum_{\theta \in \Theta} \sum_{i=1}^n s(i, \theta) \cdot \pi(\sigma_i, \theta) \mu(\theta) \right]$$

$$\text{s.t. } \sum_{\theta \in \Theta} r(i, \theta) \cdot \pi(\sigma_i, \theta) \mu(\theta) \geq \sum_{\theta \in \Theta} r(j, \theta) \cdot \pi(\sigma_i, \theta) \mu(\theta), \quad \text{for } i, j \in [n].$$

$$\sum_{i=1}^n \pi(\sigma_i, \theta) = 1, \quad \text{for } \theta \in \Theta.$$

$$\pi(\sigma_i, \theta) \geq 0, \quad \text{for } \theta \in \Theta, i \in [n].$$

Optimal Persuasion via Linear Program

- Input: prior μ , sender payoff $s(i, \theta)$, receiver payoff $r(i, \theta)$
- Variables: $\pi(\sigma_i, \theta)$

σ_i indeed incentivizes Receiver best action i

$$\max \sum_{\theta \in \Theta} \sum_{i=1}^n s(i, \theta) \cdot \pi(\sigma_i, \theta) \mu(\theta)$$

$$\text{s.t. } \boxed{\sum_{\theta \in \Theta} r(i, \theta) \cdot \pi(\sigma_i, \theta) \mu(\theta) \geq \sum_{\theta \in \Theta} r(j, \theta) \cdot \pi(\sigma_i, \theta) \mu(\theta), \text{ for } i, j \in [n].}$$

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- Variables: $\pi(\sigma_i, \theta)$

$$\max \quad \sum_{\theta \in \Theta} \sum_{i=1}^n s(i, \theta) \cdot \pi(\sigma_i, \theta) \mu(\theta)$$

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π is a valid signaling scheme

Thank You

Haifeng Xu

University of Virginia

hx4ad@virginia.edu