

# Announcements

- Please work on your course project
- HW3 will be posted around middle Nov

# CMSC 35401: The Interplay of Learning and Game Theory (Autumn 2022)

## Bayesian Persuasion

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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?**
  - Related to manipulate features to game a learning algorithm (later lectures)

# Outline

- Introduction and Bayesian Persuasion
- Algorithms for Bayesian Persuasion
- Persuading Multiple Receivers

# Two Primary Ways to Influence Agents' Behaviors

- Design/provide incentives
  - Auctions



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  - Discounts/coupons



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## ➤ Design/provide incentives

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- Discounts/coupons
- Job contract design



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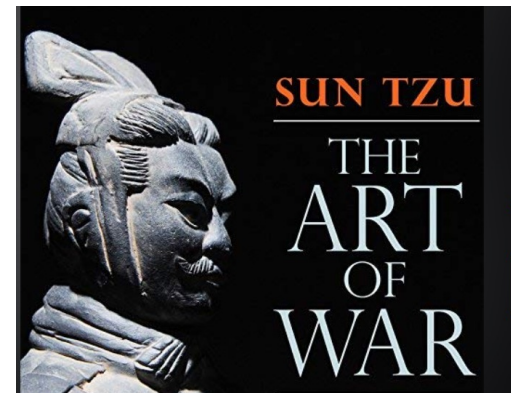
} Mechanism Design

## ➤ Influence agents' beliefs

- Deception in wars/battles

**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*





# Two Primary Ways to Influence Agents' Behaviors

## ➤ Design/provide incentives

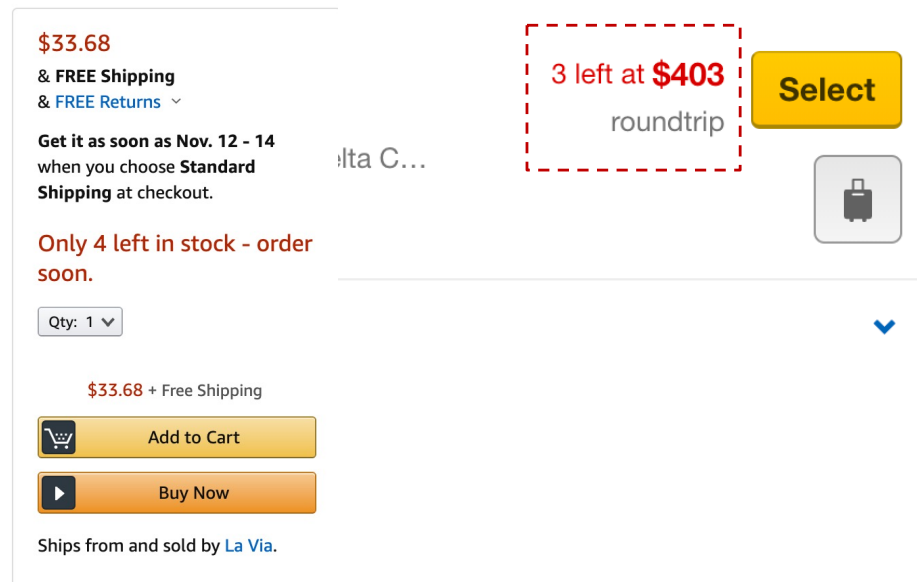
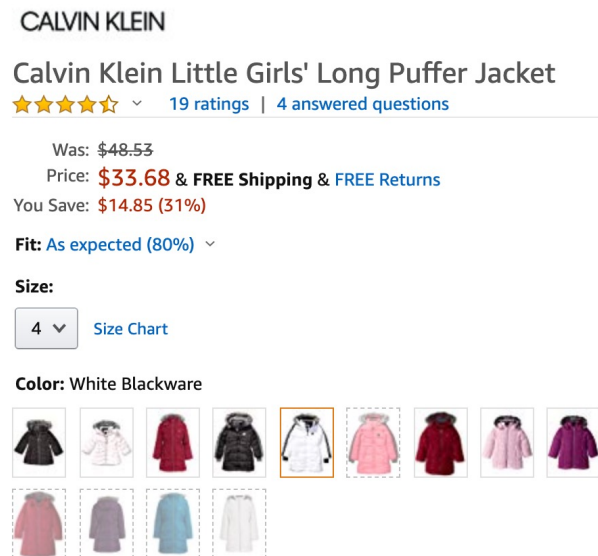
- Auctions
- Discounts/coupons
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## ➤ Influence agents' beliefs

- Deception in wars/battles
- Strategic information disclosure

Strategic inventory  
information disclosure



# Two Primary Ways to Influence Agents' Behaviors

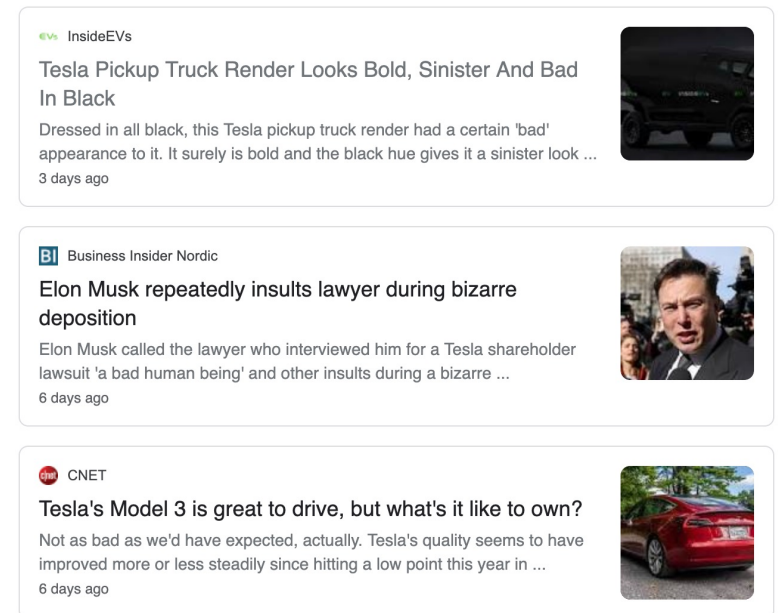
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- Auctions
- Discounts/coupons
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} Mechanism Design

## ➤ Influence agents' beliefs

- Deception in wars/battles
- Strategic information disclosure
- News articles, advertising, tweets, etc.



# Two Primary Ways to Influence Agents' Behaviors

## ➤ Design/provide incentives

- Auctions
- Discounts/coupons
- Job contract design



Mechanism Design

## ➤ Influence agents' beliefs

- Deception in wars/battles
- Strategic information disclosure
- News articles, advertising, tweets ...
- In fact, **most information you see is there with a purpose**



Persuasion  
(information design)

A whole course from Booth  
on this topic

**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



- 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
- 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

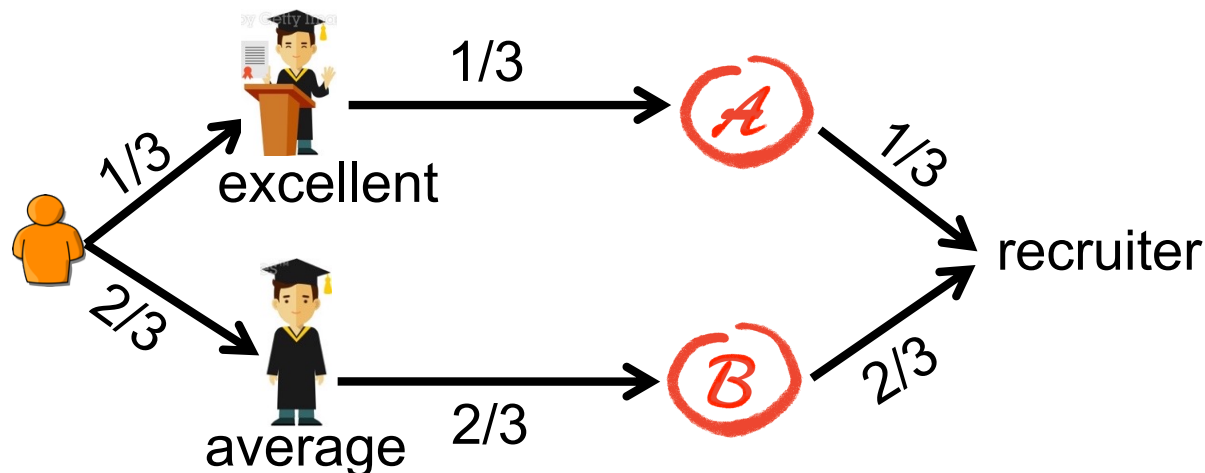
Assume advisor "commits" to some policy, and recruiter is fully aware this policy and will best respond

# Example: Recommendation Letters



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

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



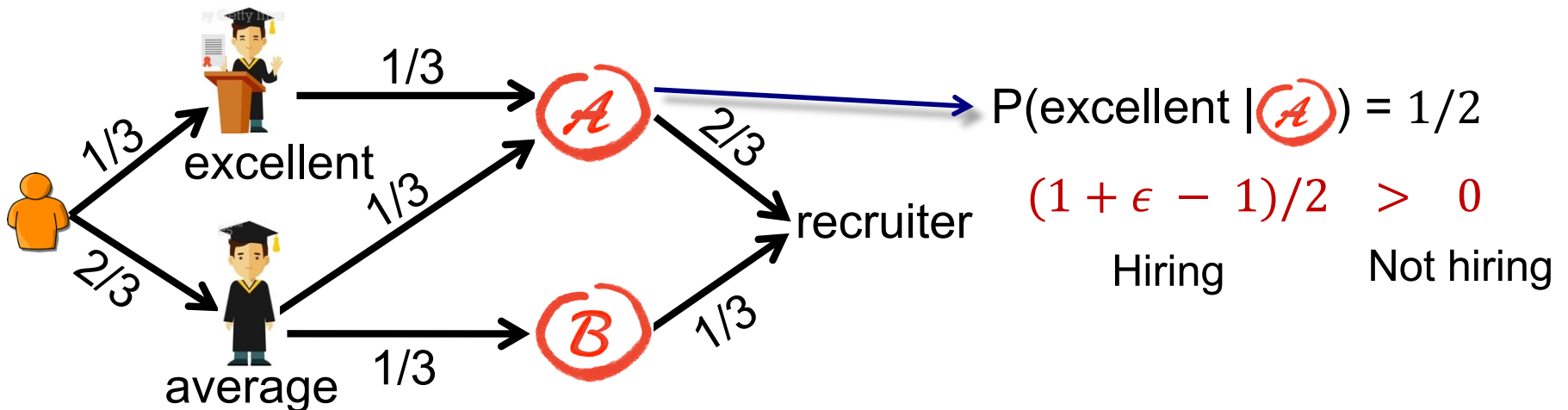


# Example: Recommendation Letters



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  $r(i, \theta)$
  - Sender utility  $s(i, \theta)$

$\theta \in \Theta$  is a random **state of nature**
- Both players know  $\theta \sim \text{prior dist. } \mu$ , but Sender has an **informational advantage** – she can observe realization of  $\theta$
- Sender wants to strategically reveal info about  $\theta$  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

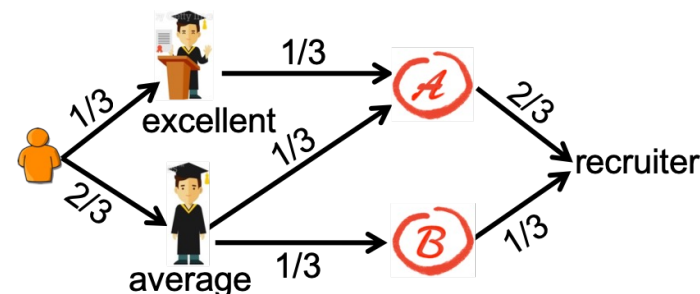
**Definition:** A signaling scheme is a mapping  $\pi: \Theta \rightarrow \Delta_{\Sigma}$  where  $\Sigma$  is the set of all possible signals.

$\pi$  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  $\theta$ )

**Note:** scheme  $\pi$  is always assumed public knowledge, thus known by Receiver

## Example

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



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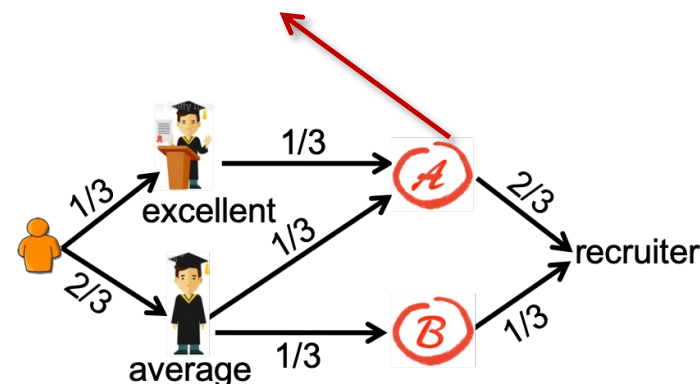
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What can Receiver infer about  $\theta$  after receiving  $\sigma$ ?

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  $\sigma$ :

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

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

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

# Revealing Information via Signaling

**Fact.** Receiver's expected utility (weakly) increases under any signaling scheme  $\pi$ .

Proof:

➤ Expected Receiver utility **without information**:  $\max_i \sum_{\theta \in \Theta} r(i, \theta) \cdot \mu(\theta)$

➤ Expected Receiver utility **under  $\pi$** :  $\sum_{\sigma} \Pr(\sigma) \cdot R(\sigma)$



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$$\begin{aligned} \sum_{\sigma} \Pr(\sigma) \cdot R(\sigma) &= \sum_{\sigma} \Pr(\sigma) \cdot \max_{i \in [n]} \left[ \sum_{\theta \in \Theta} r(i, \theta) \cdot \frac{\pi(\sigma, \theta) \cdot \mu(\theta)}{\Pr(\sigma)} \right] \\ &\geq \max_{i \in [n]} \sum_{\sigma} \Pr(\sigma) \cdot \left[ \sum_{\theta \in \Theta} r(i, \theta) \cdot \frac{\pi(\sigma, \theta) \cdot \mu(\theta)}{\Pr(\sigma)} \right] \end{aligned}$$

By HW2 problem 1

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 &\geq \max_{i \in [n]} \sum_{\sigma} \cancel{\Pr(\sigma)} \cdot \left[ \sum_{\theta \in \Theta} r(i, \theta) \cdot \frac{\pi(\sigma, \theta) \cdot \mu(\theta)}{\cancel{\Pr(\sigma)}} \right] \\
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 &= \max_{i \in [n]} \sum_{\theta \in \Theta} r(i, \theta) \cdot \left( \sum_{\sigma} \pi(\sigma, \theta) \right) \cdot \mu(\theta) \\
 &= \max_{i \in [n]} \sum_{\theta \in \Theta} r(i, \theta) \cdot \mu(\theta)
 \end{aligned}$$

# Revealing Information via Signaling

**Fact.** Receiver's expected utility (weakly) increases under any signaling scheme  $\pi$ .

Remarks:

- Signaling scheme does increase Receiver's utility
- More (even noisy) information always helps a decision maker (DM)
  - Not 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  $\theta$ .

Because any other noisy scheme  $\pi$  can be improved by further revealing  $\theta$  itself

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But this is not Sender's goal...

**Sender Objective:** maximize her own expected utility by picking  $\pi$

# Outline

- Introduction and Bayesian Persuasion
- Algorithms for Bayesian Persuasion
- Persuading Multiple Receivers

**Q:** What are obstacles when designing  $\pi = \{\pi(\theta, \sigma)\}_{\theta \in \Theta, \sigma \in \Sigma}$ ?

- Don't know what is the set of all possible signals  $\Sigma$ ...
  - Too many signals in this world to choose from (think about how many ways Amazon can reveal information to you)
- **Key observation:** a signal is mathematically nothing but a posterior distribution over  $\Theta$ 
  - 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  $n(= \# \text{ receiver actions})$  signals, where signal  $\sigma_i$  induce optimal Receiver action  $i$

➤ Conditioned on any signal  $\sigma$

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

➤ If two signal  $\sigma$  and  $\sigma'$  result in the same best action  $i^*$ , Sender can combine them as a single signal  $\sigma_{i^*} = (\sigma, \sigma')$

- Claim:  $i^*$  is still the optimal action conditioned on  $\sigma_{i^*}$

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

$$\sum_{\theta} \Pr(\theta|\sigma') r(i^*, \theta) \geq \sum_{\theta} \Pr(\theta|\sigma') r(i, \theta), \quad \forall i \quad \times (1 - p)$$

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



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

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

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

# Revelation Principle

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

➤ Conditioned on any signal  $\sigma$

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

➤ If two signal  $\sigma$  and  $\sigma'$  result in the same best action  $i^*$ , Sender can combine them as a single signal  $\sigma_{i^*} = (\sigma, \sigma')$

- Claim:  $i^*$  is still the optimal action conditioned on  $\sigma_{i^*}$
- 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  $\sigma_{i^*}$

# Revelation Principle

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

➤ Each  $\sigma_i$  can be viewed as an **action recommendation** of  $i$

CALVIN KLEIN

Calvin Klein Little Girls' Long Puffer Jacket

★★★★★ ∨ 19 ratings | 4 answered questions

Was: ~~\$48.53~~

Price: **\$33.68** & **FREE Shipping** & **FREE Returns**

You Save: **\$14.85 (31%)**

Fit: **As expected (80%)** ∨

Size:

4 ∨ [Size Chart](#)

Color: White Blackware



**\$33.68**

& **FREE Shipping**  
& **FREE Returns** ∨

**Get it as soon as Nov. 12 - 14**  
when you choose **Standard Shipping** at checkout.

**Only 4 left in stock - order soon.**

Qty: 1 ∨

**\$33.68** + Free Shipping

 **Add to Cart**

 **Buy Now**


Ships from and sold by **La Via**.

# Optimal Persuasion via Linear Program

➤ Input: prior  $\mu$ , 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  $\sigma_i$ )


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

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


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

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

# Optimal Persuasion via Linear Program

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

$\sigma_i$  indeed incentivizes Receiver best action  $i$


$$\begin{aligned} \max \quad & \sum_{\theta \in \Theta} \sum_{i=1}^n s(i, \theta) \cdot \pi(\sigma_i, \theta) \mu(\theta) \\ \text{s.t.} \quad & \boxed{\sum_{\theta \in \Theta} r(i, \theta) \cdot \pi(\sigma_i, \theta) \mu(\theta) \geq \sum_{\theta \in \Theta} r(j, \theta) \cdot \pi(\sigma_i, \theta) \mu(\theta)}, \quad \text{for } i, j \in [n]. \\ & \sum_{i=1}^n \pi(\sigma_i, \theta) = 1, \quad \text{for } \theta \in \Theta. \\ & \pi(\sigma_i, \theta) \geq 0, \quad \text{for } \theta \in \Theta, i \in [n]. \end{aligned}$$

# Optimal Persuasion via Linear Program

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

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$$\sum_{i=1}^n \pi(\sigma_i, \theta) = 1,$$

for  $\theta \in \Theta$ .

$$\pi(\sigma_i, \theta) \geq 0,$$

for  $\theta \in \Theta, i \in [n]$ .



$\pi$  is a valid signaling scheme

# Optimal Persuasion via Linear Program

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

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This should remind you the LP for correlated equilibria

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# Recommendation Letter Example cont'd



- Advisor vs. two fellowship programs
- 1/3 of the advisor's students are **excellent**; 2/3 are **average**
- A fresh graduate is randomly drawn from this population
- Each fellowship:
  - ❖ Utility  $1 + \epsilon$  for awarding excellent student;  $-1$  for average student
  - ❖ Utility  $0$  for no award
  - ❖ A-priori, only knows the advisor's student population
  - ❖ Student can accept both fellowships
- Advisor
  - ❖ Utility 1 if student gets **at least one fellowship**, 0 otherwise
  - ❖ Knows whether the student is excellent or not

# Recommendation Letter Example cont'd



Google



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

Well, we learned the lesson — noisy info!

# Recommendation Letter Example cont'd

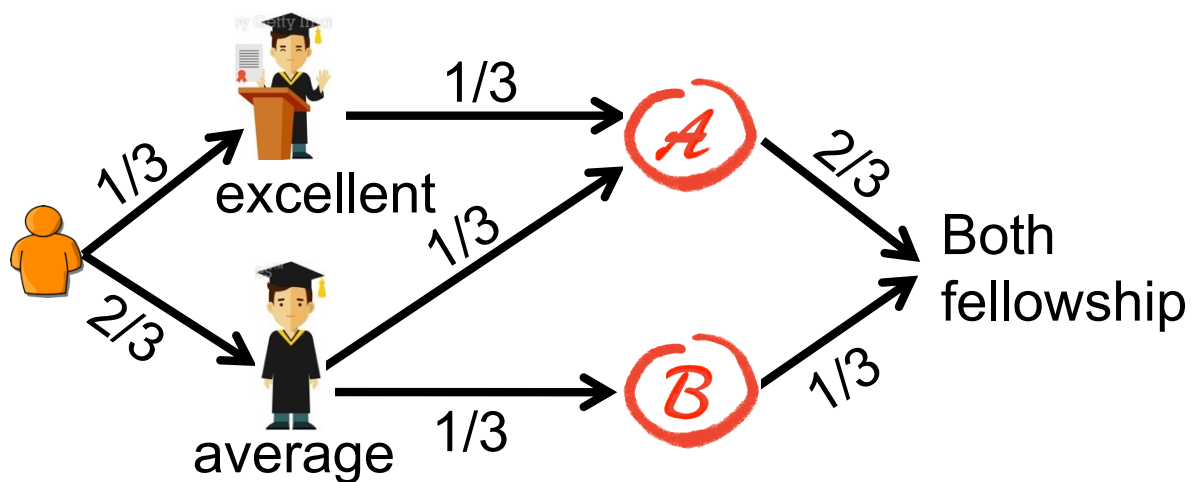


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What is the advisor's optimal "recommendation strategy"?

- Optimal **public** scheme  $\rightarrow$  advisor expected utility  $2/3$



# Recommendation Letter Example cont'd



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

- Optimal **private** scheme → advisor expected utility 1



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excellent

# Recommendation Letter Example cont'd



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

➤ Optimal **private** scheme  $\rightarrow$  advisor expected utility 1

➤ Conditioned on "strong",  
excellent with prob  $\frac{1}{2}$

➤ Always at least one  
fellowship recommended  
"strong"



Google



average

# Recommendation Letter Example cont'd



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Generalize this example to  $n$  fellowships:

advisor utility of optimal **private** scheme

$\geq \frac{n+1}{2}$  advisor utility of optimal **public** scheme

## Conceptual Message

Being able to persuade privately may have a huge advantage

Remark: fellowship programs' utilities did not decrease

# Thank You

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