

Announcements

- HW 3 and proposal due today

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

Selling Information

Instructor: Haifeng Xu

Outline

- Bayesian Persuasion and Information Selling
- Sell to a Single Decision Maker
- Sell to Multiple Decision Makers

Recap: Bayesian Persuasion

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

- One of the two primarily ways to influence agents' behaviors
 - Another way is through designing incentives
- Accounts for a significant share in economic activities
 - Advertising, marketing, security, investment, financial regulation,...



The Bayesian Persuasion Model

- Two players: a sender (she) and a receiver (he)
 - Sender has information, receiver is a decision maker
- Receiver takes an action $i \in [n] = \{1, 2, \dots, n\}$
 - Receiver utility $r(i, \theta)$ and sender utility $s(i, \theta)$
 - $\theta \sim \text{prior dist. } p$ is a random state of nature
- Both players know prior p , but sender additionally observes θ

The Bayesian Persuasion Model

- Two players: a sender (she) and a receiver (he)
 - Sender has information, receiver is a decision maker
- Receiver takes an action $i \in [n] = \{1, 2, \dots, n\}$
 - Receiver utility $r(i, \theta)$ and sender utility $s(i, \theta)$
 - $\theta \sim \text{prior dist. } p$ is a random state of nature
- Both players know prior p , but sender additionally observes θ
- Sender **reveals partial information via a signaling scheme** to influence receiver's decision and maximize her utility

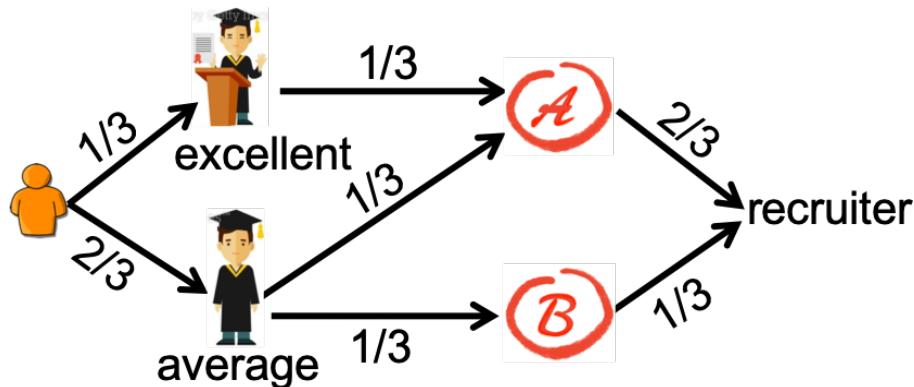
Definition: A signaling scheme is a mapping $\pi: \Theta \rightarrow \Delta_{\Sigma}$ where Σ is the set of all possible signals.

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)

Example: Recommendation Letters



- Sender = advisor, receiver = recruiter
- $\Theta = \{excellent, average\}$, $\mu(excellent) = 1/3$
- Receiver decides Hire or NotHire
 - Results in utilities for receiver and sender
- Optimal strategy is a signaling scheme



Optimal Signaling via Linear Program

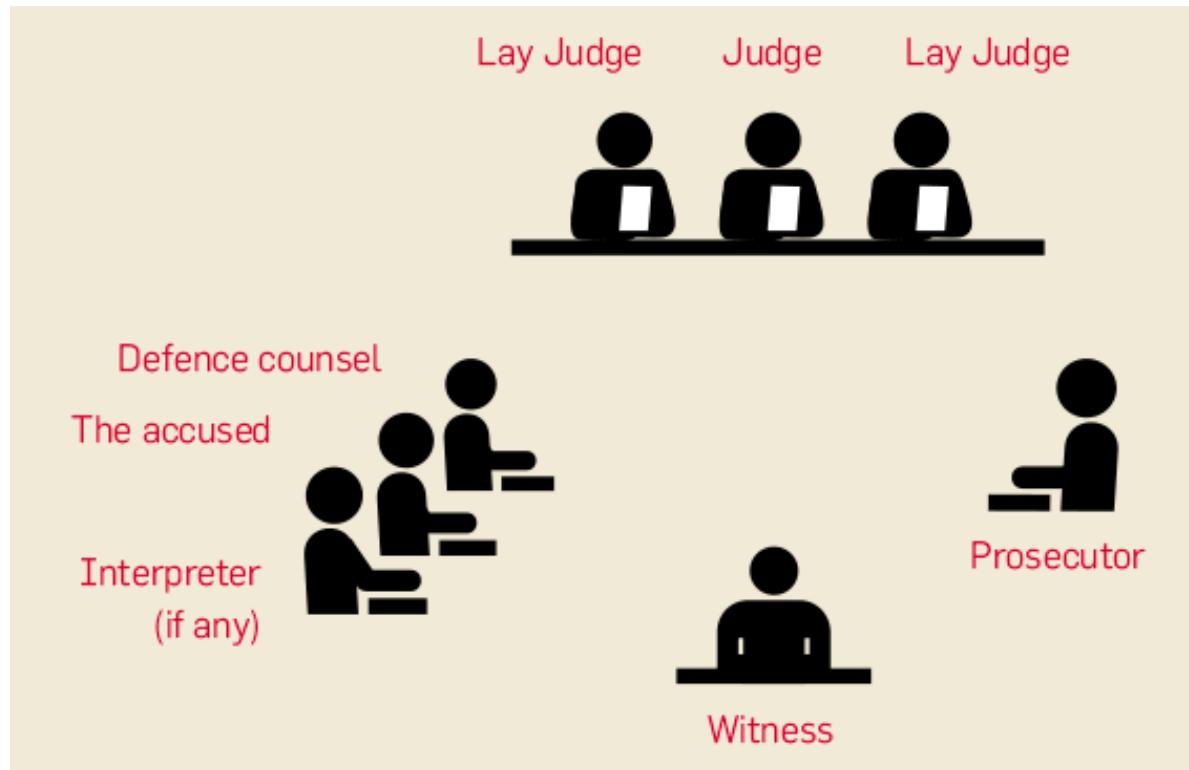
Revelation Principle. There always exists an optimal signaling scheme that uses at most n ($= \#$ receiver actions) signals, where signal σ_i induce optimal receiver action i

- Optimal signaling scheme is computed by an LP
 - Variables: $\pi(\sigma_i, \theta)$ = prob of sending σ_i conditioned on θ
 - Send σ_i = recommend action i

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

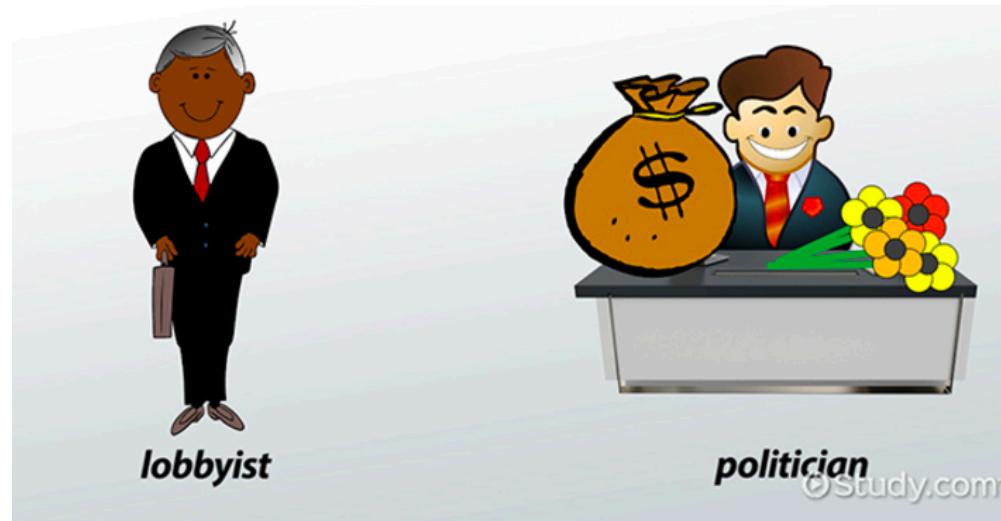
Many Other Examples and Extensions

- Prosecutor persuades judge



Many Other Examples and Extensions

- Prosecutor persuades judge
- Lobbyists persuade politicians



Many Other Examples and Extensions

- Prosecutor persuades judge
- Lobbyists persuade politicians
- Election candidates persuade voters



Many Other Examples and Extensions

- Prosecutor persuades judge
- Lobbyists persuade politicians
- Election candidates persuade voters
- Sellers persuade buyers

6:00am - 10:32am 4h 32m (1 stop) CHO - 49m in ATL - MIA Delta 5405 operated by Endeavor Air DBA Delta C...

▲ Delta Very Good Flight (7.5/10)

[Flight details▼](#)

3 left at \$403 roundtrip [Select](#)

Rules and restrictions apply ▾

Many Other Examples and Extensions

- Prosecutor persuades judge
- Lobbyists persuade politicians
- Election candidates persuade voters
- Sellers persuade buyers

The screenshot shows a product listing for a Calvin Klein Little Girls' Long Puffer Jacket. The product is priced at \$33.68 with free shipping and returns. It has a rating of 4.5 stars from 19 ratings. The price was originally \$48.53, indicating a save of \$14.85 (31%). The item is currently 4 left in stock. Shipping information states it can be delivered by Nov. 12 - 14 via standard shipping. The product is listed as being shipped from and sold by La Via. The page includes options to add the item to a cart or buy now, and a link to deliver to Charlottesville 22901.

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

100% Polyester
Imported
Zipper closure
Machine Wash
Features: this girls heavyweight jacket is water resistant length length

\$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](#).

② Deliver to Charlottesville 22901

Add to List

3 left at **\$403**
roundtrip

Select

Delta C...

13

Many Other Examples and Extensions

- Prosecutor persuades judge
- Lobbyists persuade politicians
- Election candidates persuade voters
- Sellers persuade buyers
- Executives persuade stockholders



Many Other Examples and Extensions

- Prosecutor persuades judge
- Lobbyists persuade politicians
- Election candidates persuade voters
- Sellers persuade buyers
- Executives persuade stockholders
- . . .

Many persuasion models built upon Bayesian persuasion

- Persuading many receivers, voters, attackers, drivers on road network, buyers in auctions, etc..
- Private vs public persuasion
- Selling information is also a variant

Selling Information – the Basic Model

- Sender = seller, Receiver = buyer who is a decision maker
- Buyer takes an action $i \in [n] = \{1, \dots, n\}$
- Buyer has a utility function $u(i, \theta; \omega)$ where
 - $\theta \sim dist.$ p is a random state of nature
 - $\omega \sim dist.$ f captures buyer's (private) utility type

Selling Information – the Basic Model

- Sender = seller, Receiver = buyer who is a decision maker
- Buyer takes an action $i \in [n] = \{1, \dots, n\}$
- Buyer has a utility function $u(i, \theta; \omega)$ where
 - $\theta \sim dist.$ p is a random state of nature
 - $\omega \sim dist.$ f captures buyer's (private) utility type

Remarks:

- u, p, f are public knowledge
- Assume θ, ω are independent
- In mechanism design, seller also does not know buyer's value

Selling Information – the Basic Model

- Sender = seller, Receiver = buyer who is a decision maker
- Buyer takes an action $i \in [n] = \{1, \dots, n\}$
- Buyer has a utility function $u(i, \theta; \omega)$ where
 - $\theta \sim dist.$ p is a random state of nature
 - $\omega \sim dist.$ f captures buyer's (private) utility type

Remarks:

- u, p, f are public knowledge
- Assume θ, ω are independent
- In mechanism design, seller also does not know buyer's value

Q: How to price the item if seller knows buyer's value of it?

Selling Information – the Basic Model

- Sender = seller, Receiver = buyer who is a decision maker
- Buyer takes an action $i \in [n] = \{1, \dots, n\}$
- Buyer has a utility function $u(i, \theta; \omega)$ where
 - $\theta \sim \text{dist. } p$ is a random state of nature
 - $\omega \sim \text{dist. } f$ captures buyer's (private) utility type
- Seller observes the state θ ; Buyer knows his private type ω
- Seller would like to sell her **information about θ** to maximize revenue

Key differences from Bayesian persuasion

- Seller does not have a utility fnc – instead maximize revenue
- Buyer here has private info ω , which is unknown to seller

Outline

- Bayesian Persuasion and Information Selling
- Sell to a Single Decision Maker
- Sell to Multiple Decision Makers

Warm-up: What if Buyer Has no Private Info

- $u(i, \theta; \omega)$ where state $\theta \sim \text{dist. } p$ and buyer type $\omega \sim \text{dist. } f$
- When seller also observes $\omega \dots$

Q: How to sell information optimally?

Warm-up: What if Buyer Has no Private Info

- $u(i, \theta; \omega)$ where state $\theta \sim \text{dist. } p$ and buyer type $\omega \sim \text{dist. } f$
- When seller also observes $\omega \dots$

Q: How to sell information optimally?

- Seller knows exactly how much the buyer values “any amount” of her information → should charge him just that amount

Warm-up: What if Buyer Has no Private Info

- $u(i, \theta; \omega)$ where sate $\theta \sim \text{dist. } p$ and buyer type $\omega \sim \text{dist. } f$
- When seller also observes $\omega \dots$

Q: How to sell information optimally?

- Seller knows exactly how much the buyer values “any amount” of her information → should charge him just that amount
- How to charge the most?
 - Reveal full information helps the buyer the most. Why?
 - So OPT is to charge him following amount and **then** reveal θ directly

$$\text{Payment} = \sum_{\theta \in \Theta} p(\theta) \cdot [\max_i u(i, \theta; \omega)] - \max_i \sum_{\theta \in \Theta} p(\theta) \cdot u(i, \theta; \omega)$$

Warm-up: What if Buyer Has no Private Info

- $u(i, \theta; \omega)$ where sate $\theta \sim dist. p$ and buyer type $\omega \sim dist. f$
- When seller also observes $\omega \dots$

Q: How to sell information optimally?

- Seller knows exactly how much the buyer values “any amount” of her information → should charge him just that amount
- How to charge the most?
 - Reveal full information helps the buyer the most. Why?
 - So OPT is to charge him following amount and **then** reveal θ directly

$$\text{Payment} = \sum_{\theta \in \Theta} p(\theta) \cdot [\max_i u(i, \theta; \omega)] - \max_i \sum_{\theta \in \Theta} p(\theta) \cdot u(i, \theta; \omega)$$



Buyer expected utility if learns θ

Warm-up: What if Buyer Has no Private Info

- $u(i, \theta; \omega)$ where sate $\theta \sim \text{dist. } p$ and buyer type $\omega \sim \text{dist. } f$
- When seller also observes $\omega \dots$

Q: How to sell information optimally?

- Seller knows exactly how much the buyer values “any amount” of her information → should charge him just that amount
- How to charge the most?
 - Reveal full information helps the buyer the most. Why?
 - So OPT is to charge him following amount and **then** reveal θ directly

$$\text{Payment} = \sum_{\theta \in \Theta} p(\theta) \cdot [\max_i u(i, \theta; \omega)] - \boxed{\max_i \sum_{\theta \in \Theta} p(\theta) \cdot u(i, \theta; \omega)}$$

Buyer expected utility
without knowing θ

Warm-up: What if Buyer Has no Private Info

- $u(i, \theta; \omega)$ where sate $\theta \sim dist. p$ and buyer type $\omega \sim dist. f$
- When seller also observes $\omega \dots$

Q: How to sell information optimally?

- Seller knows exactly how much the buyer values “any amount” of her information → should charge him just that amount
- How to charge the most?
 - Reveal full information helps the buyer the most. Why?
 - So OPT is to charge him following amount and **then** reveal θ directly

$$\text{Payment} = \sum_{\theta \in \Theta} p(\theta) \cdot [\max_i u(i, \theta; \omega)] - \max_i \sum_{\theta \in \Theta} p(\theta) \cdot u(i, \theta; \omega)$$

More interesting and realistic is when buyer has private info

Sell Information: Challenge I

The class of mechanisms is too broad

- The mechanism will: (1) elicit private info from buyer; (2) reveal info based on realized θ ; (3) charge buyer
- May interact with buyer for many rounds
- Buyer may misreport his private info of ω

Sell Information: Challenge I

The class of mechanisms is too broad

. . . but, at the end of the day, the buyer of type ω is charged some amount t_ω in expectation and learns a posterior belief about θ

Sell Information: Challenge I

The class of mechanisms is too broad

. . . but, at the end of the day, the buyer of type ω is charged some amount t_ω in expectation and learns a posterior belief about θ

Theorem (Revelation Principle). Any information selling mechanism can be “simulated” by a direct and truthful revelation mechanism:

1. Ask buyer to report ω
2. Charge buyer t_ω and reveal info to buyer via signaling scheme π_ω

- Proof: similar to proof of revelation principle for mechanism design
- Optimal mechanism reduces to an incentive compatible menu $\{t_\omega, \pi_\omega\}_\omega$

Sell Information: Challenge 2

Signaling scheme π_ω is still complicated

- For any fixed buyer type ω , how many signals needed for π_ω ?
 - Still n signals with σ_i recommending action i ?
 - Previous argument of merging all signals with same buyer ω best response is not valid any more – why?

Sell Information: Challenge 2

Signaling scheme π_ω is still complicated

- For any fixed buyer type ω , how many signals needed for π_ω ?
 - Still n signals with σ_i recommending action i ?
 - Previous argument of merging all signals with same buyer ω best response is not valid any more – why?

Incentive compatibility constraint for ω

$$U_\omega(\text{report } \omega) \geq U_\omega(\text{report } \omega')$$

Sell Information: Challenge 2

Signaling scheme π_ω is still complicated

- For any fixed buyer type ω , how many signals needed for π_ω ?
 - Still n signals with σ_i recommending action i ?
 - Previous argument of merging all signals with same buyer ω best response is not valid any more – why?

Incentive compatibility constraint for ω

depends only
on π_ω ,

$$U_\omega(\text{report } \omega) \geq U_\omega(\text{report } \omega')$$

Sell Information: Challenge 2

Signaling scheme π_ω is still complicated

- For any fixed **buyer type ω** , how many signals needed for π_ω ?
 - Still n signals with σ_i recommending action i ?
 - Previous argument of merging all signals with same buyer ω best response is not valid any more – why?

Incentive compatibility constraint for ω

$$U_\omega(\text{report } \omega) \geq U_\omega(\text{report } \omega')$$

depends on π_ω , but will not change due to our way of merging

Sell Information: Challenge 2

Signaling scheme π_ω is still complicated

- For any fixed buyer type ω , how many signals needed for π_ω ?
 - Still n signals with σ_i recommending action i ?
 - Previous argument of merging all signals with same buyer ω best response is not valid any more – why?

Incentive compatibility constraint for ω

$$U_\omega(\text{report } \omega) \geq U_\omega(\text{report } \omega') \checkmark$$

 depends on π_ω , but will not change due to our way of merging

So merging signals in π_ω retains this constraint

Sell Information: Challenge 2

Signaling scheme π_ω is still complicated

- For any fixed buyer type ω , how many signals needed for π_ω ?
 - Still n signals with σ_i recommending action i ?
 - Previous argument of merging all signals with same buyer ω best response is not valid any more – why?

Incentive compatibility constraint for ω

$$U_\omega(\text{report } \omega) \geq U_\omega(\text{report } \omega') \checkmark$$

Incentive compatibility constraint for any ω' ($\neq \omega$)

$$U_{\omega'}(\text{report } \omega') \geq U_{\omega'}(\text{report } \omega)$$

Sell Information: Challenge 2

Signaling scheme π_ω is still complicated

- For any fixed buyer type ω , how many signals needed for π_ω ?
 - Still n signals with σ_i recommending action i ?
 - Previous argument of merging all signals with same buyer ω best response is not valid any more – why?

Incentive compatibility constraint for ω

$$U_\omega(\text{report } \omega) \geq U_\omega(\text{report } \omega') \checkmark$$

Incentive compatibility constraint for any ω' ($\neq \omega$)

$$U_{\omega'}(\text{report } \omega') \geq U_{\omega'}(\text{report } \omega)$$

This will change! Why?

Sell Information: Challenge 2

Signaling scheme π_ω is still complicated

- For any fixed **buyer type ω** , how many signals needed for π_ω ?
 - Still n signals with σ_i recommending action i ?
 - Previous argument of **merging all signals with same buyer ω best response** is not valid any more – why?

Incentive compatibility constraint for ω

$$U_\omega(\text{report } \omega) \geq U_\omega(\text{report } \omega') \checkmark$$

Incentive compatibility constraint for any ω' ($\neq \omega$)

$$U_{\omega'}(\text{report } \omega') \geq U_{\omega'}(\text{report } \omega)$$

This will change! Why?

Sell Information: Challenge 2

Signaling scheme π_ω is still complicated

- For any fixed **buyer type ω** , how many signals needed for π_ω ?
 - Still n signals with σ_i recommending action i ?
 - Previous argument of **merging all signals with same buyer ω best response** is not valid any more – why?

Incentive compatibility constraint for ω

$$U_\omega(\text{report } \omega) \geq U_\omega(\text{report } \omega') \checkmark$$

Incentive compatibility constraint for any ω' ($\neq \omega$)

$$U_{\omega'}(\text{report } \omega') \geq U_{\omega'}(\text{report } \omega) \checkmark$$

Key idea: this term will only decrease since ω' gets less info due to merging of signals

Sell Information: Challenge 2

Signaling scheme π_ω is still complicated

- For any fixed **buyer type ω** , how many signals needed for π_ω ?
 - Still n signals with σ_i recommending action i ?
 - Previous argument of **merging all signals with same buyer ω best response** is not valid any more – why?

Theorem (Simplifying Signaling Schemes). There always exists an optimal incentive compatible menu $\{t_\omega, \pi_\omega\}_\omega$, such that π_ω uses at most n signals with σ_i recommending action i

Such an information-selling mechanism is like consulting – buyer reports type ω , seller charges him t_ω

Sell Information: the Optimal Mechanism

The Consulting Mechanism

1. Elicit buyer type ω
2. Charge buyer t_ω
3. Observe realized state θ and recommend action i to the buyer with probability $\pi_\omega(\sigma_i, \theta)$

- Will be incentive compatible – reporting true ω is optimal
- The recommended action is guaranteed to be the optimal action for buyer ω given his information
- $\{t_\omega, \pi_\omega\}_\omega$ is public knowledge, and computed by LP

Sell Information: the Optimal Mechanism

The Consulting Mechanism

1. Elicit buyer type ω
2. Charge buyer t_ω
3. Observe realized state θ and recommend action i to the buyer with probability $\pi_\omega(\sigma_i, \theta)$

- Will be incentive compatible – reporting true ω is optimal
- The recommended action is guaranteed to be the optimal action for buyer ω given his information
- $\{t_\omega, \pi_\omega\}_\omega$ is public knowledge, and computed by LP

Theorem. Consulting mechanism is optimal with $\{t_\omega, \pi_\omega\}_\omega$ computed by the following program.

Sell Information: the Optimal Mechanism

Optimal $\{r_\omega, \pi_\omega\}_\omega$ can be computed by a convex program

- Variables: $\pi_\omega(\sigma_i, \theta) = \text{prob of sending } \sigma_i \text{ conditioned on } \theta \text{ for } \omega$
- Variable t_ω is the payment from ω

$$\begin{aligned} \max \quad & \sum_\omega f(\omega) \cdot t_\omega \\ \text{s.t.} \quad & \sum_{i=1}^n \sum_{\theta \in \Theta} u(i, \theta; \omega) \cdot \pi_\omega(\sigma_i, \theta) p(\theta) - t_\omega \\ & \geq \sum_{i=1}^n \max_{j \in [n]} \left[\sum_{\theta \in \Theta} u(j, \theta; \omega) \cdot \pi_{\omega'}(\sigma_i, \theta) p(\theta) \right] - t_{\omega'}, \quad \text{for } \omega \neq \omega'. \\ & \sum_{\theta \in \Theta} u(i, \theta; \omega) \cdot \pi_\omega(\sigma_i, \theta) p(\theta) \\ & \geq \sum_{\theta \in \Theta} u(j, \theta; \omega) \cdot \pi_\omega(\sigma_i, \theta) p(\theta), \quad \text{for } i, j \in [n], \omega \in \Omega. \\ & \sum_{i=1}^n \pi_\omega(\sigma_i, \theta) = 1, \quad \text{for } \theta, \omega \in \Omega. \\ & \pi_\omega(\sigma_i, \theta) \geq 0, \quad \text{for } \theta \in \Theta, i \in [n], \omega \in \Omega. \end{aligned}$$

Sell Information: the Optimal Mechanism

Optimal $\{r_\omega, \pi_\omega\}_\omega$ can be computed by a convex program

- Variables: $\pi_\omega(\sigma_i, \theta) = \text{prob of sending } \sigma_i \text{ conditioned on } \theta \text{ for } \omega$
- Variable t_ω is the payment from ω

Expected revenue

$$\begin{aligned}
 & \max \quad \boxed{\sum_\omega f(\omega) \cdot t_\omega} \\
 \text{s.t.} \quad & \sum_{i=1}^n \sum_{\theta \in \Theta} u(i, \theta; \omega) \cdot \pi_\omega(\sigma_i, \theta) p(\theta) - t_\omega \\
 & \geq \sum_{i=1}^n \max_{j \in [n]} \left[\sum_{\theta \in \Theta} u(j, \theta; \omega) \cdot \pi_{\omega'}(\sigma_i, \theta) p(\theta) \right] - t_{\omega'}, \quad \text{for } \omega \neq \omega'. \\
 & \sum_{\theta \in \Theta} u(i, \theta; \omega) \cdot \pi_\omega(\sigma_i, \theta) p(\theta) \\
 & \geq \sum_{\theta \in \Theta} u(j, \theta; \omega) \cdot \pi_\omega(\sigma_i, \theta) p(\theta), \quad \text{for } i, j \in [n], \omega \in \Omega. \\
 & \sum_{i=1}^n \pi_\omega(\sigma_i, \theta) = 1, \quad \text{for } \theta, \omega \in \Omega. \\
 & \pi_\omega(\sigma_i, \theta) \geq 0, \quad \text{for } \theta \in \Theta, i \in [n], \omega \in \Omega.
 \end{aligned}$$

Sell Information: the Optimal Mechanism

Optimal $\{r_\omega, \pi_\omega\}_\omega$ can be computed by a convex program

- Variables: $\pi_\omega(\sigma_i, \theta) = \text{prob of sending } \sigma_i \text{ conditioned on } \theta \text{ for } \omega$
- Variable t_ω is the payment from ω

Reporting true ω is optimal

$$\max \sum_\omega f(\omega) \cdot t_\omega$$

$$\begin{aligned} \text{s.t. } & \sum_{i=1}^n \sum_{\theta \in \Theta} u(i, \theta; \omega) \cdot \pi_\omega(\sigma_i, \theta) p(\theta) - t_\omega \\ & \geq \sum_{i=1}^n \max_{j \in [n]} \left[\sum_{\theta \in \Theta} u(j, \theta; \omega) \cdot \pi_{\omega'}(\sigma_i, \theta) p(\theta) \right] - t_{\omega'}, \quad \text{for } \omega \neq \omega'. \end{aligned}$$

$$\begin{aligned} & \sum_{\theta \in \Theta} u(i, \theta; \omega) \cdot \pi_\omega(\sigma_i, \theta) p(\theta) \\ & \geq \sum_{\theta \in \Theta} u(j, \theta; \omega) \cdot \pi_\omega(\sigma_i, \theta) p(\theta), \quad \text{for } i, j \in [n], \omega \in \Omega. \end{aligned}$$

$$\begin{aligned} & \sum_{i=1}^n \pi_\omega(\sigma_i, \theta) = 1, \quad \text{for } \theta, \omega \in \Omega. \\ & \pi_\omega(\sigma_i, \theta) \geq 0, \quad \text{for } \theta \in \Theta, i \in [n], \omega \in \Omega. \end{aligned}$$

Sell Information: the Optimal Mechanism

Optimal $\{r_\omega, \pi_\omega\}_\omega$ can be computed by a convex program

- Variables: $\pi_\omega(\sigma_i, \theta) = \text{prob of sending } \sigma_i \text{ conditioned on } \theta \text{ for } \omega$
- Variable t_ω is the payment from ω

$$\max \quad \sum_\omega f(\omega) \cdot t_\omega$$

$$\begin{aligned} \text{s.t.} \quad & \sum_{i=1}^n \sum_{\theta \in \Theta} u(i, \theta; \omega) \cdot \pi_\omega(\sigma_i, \theta) p(\theta) - t_\omega \\ & \geq \sum_{i=1}^n \max_{j \in [n]} \left[\sum_{\theta \in \Theta} u(j, \theta; \omega) \cdot \pi_{\omega'}(\sigma_i, \theta) p(\theta) \right] - t_{\omega'}, \quad \text{for } \omega \neq \omega'. \end{aligned}$$

$\begin{aligned} & \sum_{\theta \in \Theta} u(i, \theta; \omega) \cdot \pi_\omega(\sigma_i, \theta) p(\theta) \\ & \geq \sum_{\theta \in \Theta} u(j, \theta; \omega) \cdot \pi_\omega(\sigma_i, \theta) p(\theta), \end{aligned}$	for $i, j \in [n], \omega \in \Omega$.
$\sum_{i=1}^n \pi_\omega(\sigma_i, \theta) = 1,$	for $\theta, \omega \in \Omega$.
$\pi_\omega(\sigma_i, \theta) \geq 0,$	for $\theta \in \Theta, i \in [n], \omega \in \Omega$.

Similar to constraints in persuasion

Sell Information: the Optimal Mechanism

Optimal $\{r_\omega, \pi_\omega\}_\omega$ can be computed by a convex program

- Variables: $\pi_\omega(\sigma_i, \theta) = \text{prob of sending } \sigma_i \text{ conditioned on } \theta \text{ for } \omega$
- Variable t_ω is the payment from ω

- A convex fnc of variables
- Can be converted to an LP

$$\begin{aligned}
 \max \quad & \sum_\omega f(\omega) \cdot t_\omega \\
 \text{s.t.} \quad & \sum_{i=1}^n \sum_{\theta \in \Theta} u(i, \theta; \omega) \cdot \pi_\omega(\sigma_i, \theta) p(\theta) - t_\omega \\
 & \geq \sum_{i=1}^n \max_{j \in [n]} \left[\sum_{\theta \in \Theta} u(j, \theta; \omega) \cdot \pi_{\omega'}(\sigma_i, \theta) p(\theta) \right] - t_{\omega'}, \quad \text{for } \omega \neq \omega'.
 \end{aligned}$$

$\sum_{\theta \in \Theta} u(i, \theta; \omega) \cdot \pi_\omega(\sigma_i, \theta) p(\theta)$
 $\geq \sum_{\theta \in \Theta} u(j, \theta; \omega) \cdot \pi_\omega(\sigma_i, \theta) p(\theta), \quad \text{for } i, j \in [n], \omega \in \Omega.$

$\sum_{i=1}^n \pi_\omega(\sigma_i, \theta) = 1, \quad \text{for } \theta, \omega \in \Omega.$
 $\pi_\omega(\sigma_i, \theta) \geq 0, \quad \text{for } \theta \in \Theta, i \in [n], \omega \in \Omega.$

Outline

- Bayesian Persuasion and Information Selling
- Sell to a Single Decision Maker
- Sell to Multiple Decision Makers

Challenges

- For single decision maker, more information always helps
 - Recall in persuasion, receiver always benefits from signaling scheme
- A fundamental challenge for selling to multiple buyers is that information does not necessarily help them

Example: More Information Hurts Buyers

- Insurance industry: *insurance company* and *customer*
 - Both are potential information buyers
- Two types of customers: *Healthy* and *Unhealthy*
 - Publicly know, $\Pr(\text{Healthy}) = 0.9$
- Seller is an information holder, who knows whether any customer is healthy or not

Insurance company		
customer	Sell	Not Sell
Buy	(-10, 10)	(-0, 0)
Not Buy	(0 , 0)	(0 , 0)

Healthy customer

Insurance company		
customer	Sell	Not Sell
Buy	(-10, -50)	(-110, 0)
Not Buy	(-111 , 0)	(-111 , 0)

Unhealthy customer

Example: More Information Hurts Buyers

Insurance company		
customer	Sell	Not Sell
Buy	(-10, 10)	(-0, 0)
Not Buy	(0 , 0)	(0 , 0)

Healthy customer, prob = 0.9

Insurance company		
customer	Sell	Not Sell
Buy	(-10, -50)	(-110, 0)
Not Buy	(-111 , 0)	(-111 , 0)

Unhealthy customer

Q: What happens without seller's information ?

- Customer and insurance company will look at expectation
 - Dominant strategy equilibrium is (Buy, Sell)

	Sell	Not Sell
Buy	(-10, 4)	(-11 , 0)
Not Buy	(-11.1, 0)	(-11.1, 0)

Example: More Information Hurts Buyers

Insurance company		
customer	Sell	Not Sell
Buy	(-10, 10)	(-0, 0)
Not Buy	(0 , 0)	(0 , 0)

Healthy customer, prob = 0.9

Insurance company		
customer	Sell	Not Sell
Buy	(-10, -50)	(-110, 0)
Not Buy	(-111 , 0)	(-111 , 0)

Unhealthy customer

Q: What if seller tells (only) customer her health status ?

E.g., customer wants to buy info from seller to decide whether he should buy insurance or not

Example: More Information Hurts Buyers

Insurance company

customer	Sell	Not Sell
Buy	(-10, 10)	(-0, 0)
Not Buy	(0 , 0)	(0 , 0)

Healthy customer, prob = 0.9

Insurance company

customer	Sell	Not Sell
Buy	(-10, -50)	(-110, 0)
Not Buy	(-111 , 0)	(-111 , 0)

Unhealthy customer

Q: What if seller tells (only) customer her health status ?

- If Healthy, customer will not buy
- If Unhealthy, customer will buy
- Customer's reaction reveals his healthy status

Example: More Information Hurts Buyers

Insurance company		
customer	Sell	Not Sell
Buy	(-10, 10)	(-0, 0)
Not Buy	(0 , 0)	(0 , 0)

Healthy customer, prob = 0.9

Insurance company		
customer	Sell	Not Sell
Buy	(-10, -50)	(-110, 0)
Not Buy	(-111 , 0)	(-111 , 0)

Unhealthy customer

Q: What if seller tells (only) customer her health status ?

- If Healthy, customer will not buy → utility (0,0) for both
- If Unhealthy, customer will buy → Will not sell, utility (-110,0)
- Customer's reaction reveals his healthy status

Example: More Information Hurts Buyers

Insurance company		
customer	Sell	Not Sell
Buy	(-10, 10)	(-0, 0)
Not Buy	(0 , 0)	(0 , 0)

Healthy customer, prob = 0.9

Insurance company		
customer	Sell	Not Sell
Buy	(-10, -50)	(-110, 0)
Not Buy	(-111 , 0)	(-111 , 0)

Unhealthy customer

Q: What if seller tells (only) customer her health status ?

- If Healthy, customer will not buy → utility (0,0) for both
- If Unhealthy, customer will buy → Will not sell, utility (-110,0)
- Customer's reaction reveals his healthy status
- In expectation (-11, 0)

Recall previously (-10,4)

Thank You

Haifeng Xu

University of Virginia

hx4ad@virginia.edu