

# Bilateral Trade with Costly Information Acquisition

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# Introduction

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- Which objectives can be implemented?

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- Principal proposes a trading mechanism,
- Buyer and Seller privately acquire payoff-relevant information,
- Information acquisition is **costly** and **flexible**,
- Information acquisition can be **arbitrarily correlated** across players.

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## What do we do?

- Provide implementability conditions,
- Characterize **info structures** consistent with **allocational efficiency**.
  - Application: **subsidy minimization** for efficient trade.

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## Endogenous information acquisition can help.

- Bikhchandani (2010): FSE  $\Rightarrow$  incentives to **acquire info** about others.
- Bikhchandani and Obara (2017): **“inflexible”** info  $\Rightarrow$  FSE (not always).
  - “inflexible” = finitely many conditionally independent signals.



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## Flexible endogenous information acquisition addresses the challenge.

- Also interesting in its own right.
- Growing literature on flexible info in fixed games.

# Preview of results

## **Tractable characterization of implementability.**

- Finite dimensional system of equations and inequalities.

## **Information structures consistent with allocational efficiency.**

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## Information structures consistent with allocational efficiency.

- Perfectly correlated signals  $\Rightarrow$  allocational efficiency.
- *Allocational efficiency  $\Rightarrow$  (essentially) perfectly correlated signals.*

## Application: subsidy minimization for efficient trade.

- Perfect correlation forces the designer to give up surplus.
  - Compensate for the cost of information acquisition  $\Rightarrow$  *no gross FSE.*
  - Prevent further information acquisition  $\Rightarrow$  *no net FSE.*

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**Principal collects revenue from/subsidizes trade.**



# Model: information

The true quality  $v \in V$  is unknown to anyone at the beginning.

We need a model where players jointly determine info structure:

State $v$	$s_1^s$	$s_2^s$	...	$s_J^s$
$s_1^b$	$\alpha_{11}(v)$	$\alpha_{12}(v)$	...	$\alpha_{1J}(v)$
$s_2^b$	$\alpha_{21}(v)$	$\alpha_{22}(v)$	...	$\alpha_{2J}(v)$
$\vdots$	$\vdots$	$\vdots$	$\ddots$	$\vdots$
$s_I^b$	$\alpha_{I1}(v)$	$\alpha_{I2}(v)$	...	$\alpha_{IJ}(v)$

$\Rightarrow$  player's actions = random variables.

# Model: information acquisition

## Commonly known to everyone at the beginning:

- Probability space  $(X, \mathcal{F}, \mathbb{P})$ , where  $X = [0, 1] \ni x$  and  $\mathbb{P}$  is uniform.
- A random variable  $\mathbf{V} : X \rightarrow V$ , induces a **common prior**  $\mu_0$  on  $V$ .

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## Players acquire info about $v \in V$ by choosing other random var's.

- Players have access to a countably infinite set of **signal realizations**.
- A **signal** of player  $p \in \{b, s\}$  is a pair  $\sigma^p = (S^p, \mathbf{S}^p)$ , where
  - $S^p$  is a **finite non-empty** subset of  $\mathbb{N}$ ,  $\mathbf{S}^p : X \rightarrow S^p$  is a random variable.

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$(\mathbf{V}, \sigma^b, \sigma^s)$  induces a **joint distribution**  $\alpha$  over  $V \times S^b \times S^s$ .

- Any Bayes-plausible (i.e.  $\text{marg}_V \alpha = \mu_0$ )  $\alpha$  can be induced.

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## Signals are costly; $C(\sigma^p)$ is posterior separable, ...

- **Today:**  $C(\sigma^p)$  is proportional to reduction in entropy.

# Model: timing

- ① Nature draws  $x \in X$  uniformly, but nobody observes it.
- ② Principal designs a trading mechanism  $(M, q, t)$ .
  - $M = M^b \times M^s$ ;  $M^p$  is the message space of player  $p$ .
  - $q = (q^b, q^s)$ ;  $q^p : M \rightarrow [0, 1]$  is the allocation function of player  $p$ .
  - $t = (t^b, t^s)$ ;  $t^p : M \rightarrow \mathbb{R}$  is the payment function of player  $p$ .
- ③ Each player  $p$  privately chooses  $\sigma^p = (S^p, \mathbf{S}^p)$ .
- ④ Each player  $p$  privately observes  $s^p = \mathbf{S}^p(x)$  and sends  $m^p \in M^p$ .
- ⑤ Allocations and payments are determined according to  $(q, t)$ ; Quality  $v = \mathbf{V}(x)$  is realized.

# Roadmap

- 1 Revelation principle
- 2 Implementability
- 3 Information structures consistent with efficient trade
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# Revelation principle

**Unlike in standard mechanism design, type space is **endogenous**.**

- Players choose **signals** in response to principal's mechanism.
- Players' **signal realizations** become their "types".

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- Principal selects equilibrium  $\Rightarrow$  correctly anticipates players' choice of signals  $\Rightarrow$  can ask about their signal realizations directly.

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- Principal selects equilibrium  $\Rightarrow$  correctly anticipates players' choice of signals  $\Rightarrow$  can ask about their signal realizations directly.

**Revelation principle: it is w.l.o.g. to consider **direct mechanisms**.**

- Players could report one of their signal realizations or abstain:

$$M^b = S^b \cup \{m_\emptyset^b\}, \quad M^s = S^s \cup \{m_\emptyset^s\},$$

where  $S^b$  and  $S^s$  are endogenously determined.

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# Implementability lemma

## Lemma (Implementability for the buyer)

$(\alpha, q, t)$  is implementable for the buyer iff there are multipliers  $\lambda_j^b(v)$  for all  $s_j^s \in S^s$  and  $\phi_{ij}^b(v)$  for all  $(s_i^b, s_j^s) \in S^b \times S^s$  and all  $v \in V$ :

$$(ST^b) \quad \underbrace{q_{ij}^b u^b(v) - t_{ij}^b}_{\frac{\partial U^b}{\partial \alpha_{ij}(v)}} - \underbrace{\log(\mu_i^b(v))}_{\frac{\partial C^b}{\partial \alpha_{ij}(v)}} - \lambda_j^b(v) + \phi_{ij}^b(v) = 0,$$

$$(DF^b) \quad \phi_{ij}^b(v) \geq 0,$$

$$(CS^b) \quad \alpha_{ij}(v) \phi_{ij}^b(v) = 0,$$

$$(NA^b) \quad \sum_{v \in V} \exp(-\min_j \{\lambda_j^b(v)\}) \leq 1.$$

- Analogous conditions apply to the seller. Seller's implementability

# Buyer's problem

Consider a candidate  $(\alpha, q, t)$  with  $\alpha$  induced by some  $(\sigma^b, \sigma^s)$ .

- Does Buyer have a profitable deviation  $\tilde{\sigma}^b$ ?
- $I + 1$  actions under  $\sigma^b \Rightarrow \tilde{\sigma}^b$  with  $\leq I + 1$  realizations are w.l.o.g.
- $(\tilde{\sigma}^b, \sigma^s)$  will induce an alternative information structure  $\tilde{\alpha}$ .
- Can rewrite the best deviation problem in terms of  $\tilde{\alpha}$ :

$$\mathcal{BD}^b(\alpha, q, t) = \operatorname{argmax}_{\tilde{\alpha}, \tilde{S}^b} \sum_{i=1}^I \sum_{j=1}^J \sum_{v \in V} \tilde{\alpha}_{ij}(v) (q_{ij}^b u^b(v) - t_{ij}^b) - c^b(\tilde{\alpha}),$$

$$(1) \quad \tilde{S}^b = S^b \cup \{s_{\emptyset}^b\}, \quad \tilde{\alpha} \in \Delta(\tilde{S}^b \times S^s \times V);$$

$$(2) \quad \operatorname{marg}_{S^s \times V} \tilde{\alpha} = \operatorname{marg}_{S^s \times V} \alpha.$$

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**Implementability condition for the buyer:**  $(\alpha, S^b) \in \mathcal{BD}^b(\alpha, q, t)$ .

# Solution to Buyer's problem

We split Buyer's deviations into two classes:

- **Class 1:** induce different  $\tilde{\alpha}$ 's over the same signal realizations  $S^b$ .
- **Class 2:** augment  $S^b$  with  $s_{\emptyset}^b$  with positive probability.



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Our approach:

- 1 Solve **Class 1**-problem, characterize solution in terms of  $\lambda$ ,  $\phi$ .
  - Convex problem  $\Rightarrow$  KKT conditions are necessary and sufficient.
- 2 Show the following:

Lemma

If  $\alpha$  solves **Class 1**-problem, then  $(\alpha, S^b)$  solves **Class 2**-problem iff

$$(NA^b) \quad \sum_{v \in V} \exp(-\min_j \{\lambda_j^b(v)\}) \leq 1.$$

# Class 2-lemma

## Lemma

If  $\alpha$  solves **Class 1-problem**, then  $(\alpha, S^b)$  solves **Class 2-problem** iff

$$(NA^b) \quad \sum_{v \in V} \exp(-\min_j \{\lambda_j^b(v)\}) \leq 1.$$

**Proof sketch:** illustrate the proof using a  $2 \times 2 \times 2$  example.

State $\underline{v}$	$s_1^s$	$s_2^s$	State $\bar{v}$	$s_1^s$	$s_2^s$
$s_1^b$	$\underline{\alpha}_{11} - \epsilon \underline{\beta}_{11}$	$\underline{\alpha}_{12} - \epsilon \underline{\beta}_{12}$	$s_1^b$	$\bar{\alpha}_{11} - \epsilon \bar{\beta}_{11}$	$\bar{\alpha}_{12} - \epsilon \bar{\beta}_{12}$
$s_2^b$	$\underline{\alpha}_{21} - \epsilon \underline{\beta}_{21}$	$\underline{\alpha}_{22} - \epsilon \underline{\beta}_{22}$	$s_2^b$	$\bar{\alpha}_{21} - \epsilon \bar{\beta}_{21}$	$\bar{\alpha}_{22} - \epsilon \bar{\beta}_{22}$
$s_\emptyset^b$	$\epsilon(\underline{\beta}_{11} + \underline{\beta}_{21})$	$\epsilon(\underline{\beta}_{12} + \underline{\beta}_{22})$	$s_\emptyset^b$	$\epsilon(\bar{\beta}_{11} + \bar{\beta}_{21})$	$\epsilon(\bar{\beta}_{12} + \bar{\beta}_{22})$

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$G_\alpha(\epsilon\beta)$  is the gain from deviation in direction  $\epsilon\beta$  from  $\alpha$ .

$MG_\alpha(\beta) \equiv \lim_{\epsilon \rightarrow 0} \frac{1}{\epsilon} G_\alpha(\epsilon\beta)$  is the corresponding marginal gain.

# Proof of Class 2-lemma

## We prove the contrapositive statement:

- Suppose there is a deviation with a positive gain  $G_\alpha(\beta) > 0$ .
- Convexity of cost function  $\Rightarrow [G_\alpha(\beta) > 0 \Rightarrow MG_\alpha(\beta) > 0]$ .
- $MG_\alpha(\beta)$  can be computed in closed form:

$$\begin{aligned}
 MG_\alpha(\beta) &= - \sum_{i,j,v} \overbrace{\beta_{ij}(v)}^{0 \text{ if } \alpha_{ij}(v)=0} \times \underbrace{\left[ q_{ij}^b u^b(v) - t_{ij}^b - \log(\mu_i^b(v)) \right]}_{=\lambda_j^b(v) \text{ as long as } \alpha_{ij}(v)>0, \text{ by KKT}} - M\text{Cost}(\beta) \\
 &= - \sum_{i,j,v} \beta_{ij}(v) \lambda_j^b(v) - M\text{Cost}(\beta).
 \end{aligned}$$

- Let  $\beta^* = \operatorname{argmax}_\beta MG_\alpha(\beta; \lambda^b)$ , then  $MG_\alpha(\beta^*; \lambda^b) > 0 \Rightarrow \neg(\text{NA}^b)$ .

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# Information structures consistent with efficient trade

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Proposition (Efficiency  $\Rightarrow$  Essentially perfect correlation)

If  $\alpha$  is consistent with efficient trade, then  $\alpha$  has the following form:

State $v$	$s_1^s$	...	$s_k^s$	...	$s_{l-\ell}^s$	...	$s_l^s$
$s_1^b$	$\alpha_{11}$	...	$\alpha_{1k}$	...	0	...	0
$\vdots$	$\vdots$	$\ddots$	$\vdots$	$\ddots$	$\vdots$	$\ddots$	$\vdots$
$s_k^b$	$\alpha_{k1}$	...	$\alpha_{kk}$	...	0	...	0
$\vdots$	$\vdots$	$\ddots$	$\vdots$	$\ddots$	$\vdots$	$\ddots$	$\vdots$
$s_{l-\ell}^b$	0	...	0	...	$\alpha_{l-\ell, l-\ell}$	...	$\alpha_{l-\ell, l}$
$\vdots$	$\vdots$	$\ddots$	$\vdots$	$\ddots$	$\vdots$	$\ddots$	$\vdots$
$s_l^b$	0	...	0	...	$\alpha_{l, l-\ell}$	...	$\alpha_{ll}$

and the posteriors within each block are equal to each other.



# Efficiency $\Rightarrow$ Essentially perfect correlation

Proof sketch ( $2 \times 2 \times 2$ , distinct posteriors)

Consider the following special case:

State $\underline{v}$	$s_1^s$	$s_2^s$	State $\bar{v}$	$s_1^s$	$s_2^s$
$s_1^b$	$\underline{\alpha}_{11}$	$\underline{\alpha}_{12}$	$s_1^b$	$\bar{\alpha}_{11}$	$\bar{\alpha}_{12}$
$s_2^b$	$\underline{\alpha}_{21}$	$\underline{\alpha}_{22}$	$s_2^b$	$\bar{\alpha}_{21}$	$\bar{\alpha}_{22}$

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$s_2^b$	$\underline{\alpha}_{21}$	$\underline{\alpha}_{22}$	$s_2^b$	$\bar{\alpha}_{21}$	$\bar{\alpha}_{22}$

Our goal is to show perfect correlation:

State $\underline{v}$	$s_1^s$	$s_2^s$	State $\bar{v}$	$s_1^s$	$s_2^s$
$s_1^b$	$\underline{\alpha}_{11}$	0	$s_1^b$	$\bar{\alpha}_{11}$	0
$s_2^b$	0	$\underline{\alpha}_{22}$	$s_2^b$	0	$\bar{\alpha}_{22}$

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$$(ST_{11}^b) \quad \underline{u}^b - t_{11}^b - \log(\underline{\mu}_1^b) - \underline{\lambda}_1^b + \underline{\phi}_{11}^b = 0,$$

$$\bar{u}^b - t_{11}^b - \log(\bar{\mu}_1^b) - \bar{\lambda}_1^b + \bar{\phi}_{11}^b = 0;$$

$$\bar{\lambda}_1^b - \underline{\lambda}_1^b - (\bar{u}^b - \underline{u}^b) = \bar{\phi}_{11}^b - \underline{\phi}_{11}^b - \log \left[ \frac{\bar{\mu}_1^b}{\underline{\mu}_1^b} \right].$$

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 \end{aligned}$$

$$\bar{\lambda}_1^b - \underline{\lambda}_1^b - (\bar{u}^b - \underline{u}^b) = \bar{\phi}_{21}^b - \underline{\phi}_{21}^b - \log \left[ \frac{\bar{\mu}_2^b}{\underline{\mu}_2^b} \right].$$

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$$\bar{\phi}_{11}^b + \underline{\phi}_{21}^b = \bar{\phi}_{21}^b + \underline{\phi}_{11}^b + \underbrace{\log \left[ \frac{\bar{\mu}_1^b}{\underline{\mu}_1^b} \right] - \log \left[ \frac{\bar{\mu}_2^b}{\underline{\mu}_2^b} \right]}_{>0 \text{ by ordering assumption}}.$$

# Efficiency $\Rightarrow$ Essentially perfect correlation

Proof sketch ( $2 \times 2 \times 2$ , distinct posteriors)

$$\bar{\phi}_{11}^b + \underline{\phi}_{21}^b = \bar{\phi}_{21}^b + \underline{\phi}_{11}^b + \underbrace{\log \left[ \frac{\bar{\mu}_1^b}{\underline{\mu}_1^b} \right] - \log \left[ \frac{\bar{\mu}_2^b}{\underline{\mu}_2^b} \right]}_{>0 \text{ by ordering assumption}}.$$

**Consideration of  $(ST_{11}^b)$  and  $(ST_{21}^b)$  therefore implies:**

$$\underbrace{\bar{\phi}_{11}^b + \underline{\phi}_{21}^b}_{\Rightarrow \text{at least one term is } >0} > \bar{\phi}_{21}^b + \underline{\phi}_{11}^b.$$

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$\Rightarrow$  at least one term is  $>0$

**Analogous consideration of  $(ST_{12}^b)$  and  $(ST_{22}^b)$  implies:**

$$\underbrace{\bar{\phi}_{12}^b + \underline{\phi}_{22}^b} > \bar{\phi}_{22}^b + \underline{\phi}_{12}^b.$$

$\Rightarrow$  at least one term is  $>0$



# Efficiency $\Rightarrow$ Essentially perfect correlation

Proof sketch ( $2 \times 2 \times 2$ , distinct posteriors)

Efficiency  $\Rightarrow$  Essentially perfect correlationProof sketch ( $2 \times 2 \times 2$ , distinct posteriors)Suppose  $\bar{\phi}_{11}^b > 0$  and  $\bar{\phi}_{12}^b > 0$ , then CS implies:

State $\underline{v}$	$s_1^s$	$s_2^s$	State $\bar{v}$	$s_1^s$	$s_2^s$
$s_1^b$	$\underline{\alpha}_{11}$	$\underline{\alpha}_{12}$	$s_1^b$	$0$	$0$
$s_2^b$	$\underline{\alpha}_{21}$	$\underline{\alpha}_{22}$	$s_2^b$	$\bar{\alpha}_{21}$	$\bar{\alpha}_{22}$

Efficiency  $\Rightarrow$  Essentially perfect correlationProof sketch ( $2 \times 2 \times 2$ , distinct posteriors)Suppose  $\bar{\phi}_{11}^b > 0$  and  $\bar{\phi}_{12}^b > 0$ , then CS implies:

State $\underline{v}$	$s_1^s$	$s_2^s$
$s_1^b$	$\underline{\alpha}_{11}$	$\underline{\alpha}_{12}$
$s_2^b$	$\underline{\alpha}_{21}$	$\underline{\alpha}_{22}$

State $\bar{v}$	$s_1^s$	$s_2^s$
$s_1^b$	0	0
$s_2^b$	$\bar{\alpha}_{21}$	$\bar{\alpha}_{22}$

Suppose  $\underline{\phi}_{21}^b > 0$  and  $\underline{\phi}_{22}^b > 0$ , then CS implies:

State $\underline{v}$	$s_1^s$	$s_2^s$
$s_1^b$	$\underline{\alpha}_{11}$	$\underline{\alpha}_{12}$
$s_2^b$	0	0

State $\bar{v}$	$s_1^s$	$s_2^s$
$s_1^b$	$\bar{\alpha}_{11}$	$\bar{\alpha}_{12}$
$s_2^b$	$\bar{\alpha}_{21}$	$\bar{\alpha}_{22}$

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State $\underline{v}$	$s_1^s$	$s_2^s$	State $\bar{v}$	$s_1^s$	$s_2^s$
$s_1^b$	$\underline{\alpha}_{11}$	$\underline{\alpha}_{12}$	$s_1^b$	$0$	$\bar{\alpha}_{12}$
$s_2^b$	$\underline{\alpha}_{21}$	$0$	$s_2^b$	$\bar{\alpha}_{21}$	$\bar{\alpha}_{22}$

Efficiency  $\Rightarrow$  Essentially perfect correlationProof sketch ( $2 \times 2 \times 2$ , distinct posteriors)Suppose  $\bar{\phi}_{11}^b > 0$  and  $\underline{\phi}_{22}^b > 0$ , then CS implies:

State $\underline{v}$	$s_1^s$	$s_2^s$	State $\bar{v}$	$s_1^s$	$s_2^s$
$s_1^b$	$\underline{\alpha}_{11}$	$\underline{\alpha}_{12}$	$s_1^b$	$0$	$\bar{\alpha}_{12}$
$s_2^b$	$\underline{\alpha}_{21}$	$0$	$s_2^b$	$\bar{\alpha}_{21}$	$\bar{\alpha}_{22}$

- Bayes-plausibility then implies:

$$\bar{\mu}_0 < \bar{\mu}_1^b = \frac{\bar{\alpha}_{12}}{\underline{\alpha}_{11} + \underline{\alpha}_{12} + \bar{\alpha}_{12}} \leq \frac{\bar{\alpha}_{12}}{\underline{\alpha}_{12} + \bar{\alpha}_{12}},$$

$$\bar{\mu}_0 > \bar{\mu}_2^s = \frac{\bar{\alpha}_{12} + \bar{\alpha}_{22}}{\underline{\alpha}_{12} + \bar{\alpha}_{12} + \bar{\alpha}_{22}} \geq \frac{\bar{\alpha}_{12}}{\underline{\alpha}_{12} + \bar{\alpha}_{12}}.$$

Efficiency  $\Rightarrow$  Essentially perfect correlationProof sketch ( $2 \times 2 \times 2$ , distinct posteriors)

The only remaining possibility is  $\bar{\phi}_{12}^b > 0$  and  $\underline{\phi}_{21}^b > 0$ :

State $\underline{v}$	$s_1^s$	$s_2^s$	State $\bar{v}$	$s_1^s$	$s_2^s$
$s_1^b$	$\underline{\alpha}_{11}$	$\underline{\alpha}_{12}$	$s_1^b$	$\bar{\alpha}_{11}$	0
$s_2^b$	0	$\underline{\alpha}_{22}$	$s_2^b$	$\bar{\alpha}_{21}$	$\bar{\alpha}_{22}$

Efficiency  $\Rightarrow$  Essentially perfect correlationProof sketch ( $2 \times 2 \times 2$ , distinct posteriors)

The only remaining possibility is  $\bar{\phi}_{12}^b > 0$  and  $\underline{\phi}_{21}^b > 0$ :

State $\underline{v}$	$s_1^s$	$s_2^s$	State $\bar{v}$	$s_1^s$	$s_2^s$
$s_1^b$	$\underline{\alpha}_{11}$	$\underline{\alpha}_{12}$	$s_1^b$	$\bar{\alpha}_{11}$	0
$s_2^b$	0	$\underline{\alpha}_{22}$	$s_2^b$	$\bar{\alpha}_{21}$	$\bar{\alpha}_{22}$

Analogous argument for Seller gives  $\underline{\phi}_{12}^s > 0$  and  $\bar{\phi}_{21}^s > 0$ :

State $\underline{v}$	$s_1^s$	$s_2^s$	State $\bar{v}$	$s_1^s$	$s_2^s$
$s_1^b$	$\underline{\alpha}_{11}$	0	$s_1^b$	$\bar{\alpha}_{11}$	$\bar{\alpha}_{12}$
$s_2^b$	$\underline{\alpha}_{12}$	$\underline{\alpha}_{22}$	$s_2^b$	0	$\bar{\alpha}_{22}$

# Roadmap

- 1 Revelation principle
- 2 Implementability
- 3 Information structures consistent with efficient trade
- 4 Application: subsidy minimization for efficient trade**
- 5 Concluding remarks



Subsidy minimization  $\Rightarrow$  perfect correlation is w.l.o.g.

### Corollary (Perfect correlation)

If  $(\alpha', I', J'; t'; \phi', \lambda')$  is feasible in the subsidy minimization problem, then there is  $(\alpha, I, J; t; \phi; \lambda)$ , which is also feasible and achieves the same objective value, but  $I = J$  and  $\underline{\alpha}_{ij} = \bar{\alpha}_{ij} = 0$  for  $i \neq j$ .

State $\underline{v}$	$s_1^s$	$s_2^s$	...	$s_I^s$	State $\bar{v}$	$s_1^s$	$s_2^s$	...	$s_I^s$
$s_1^b$	$\underline{\alpha}_1$	0	...	0	$s_1^b$	$\bar{\alpha}_1$	0	...	0
$s_2^b$	0	$\underline{\alpha}_2$	...	0	$s_2^b$	0	$\bar{\alpha}_2$	...	0
$\vdots$	$\vdots$	$\vdots$	$\ddots$	$\vdots$	$\vdots$	$\vdots$	$\vdots$	$\ddots$	$\vdots$
$s_I^b$	0	0	...	$\underline{\alpha}_I$	$s_I^b$	0	0	...	$\bar{\alpha}_I$

**Proof:** merge signal realizations with equal posteriors.

Subsidy minimization  $\Rightarrow$  perfect correlation is w.l.o.g.

## Corollary (Perfect correlation)

If  $(\alpha', I', J'; t'; \phi', \lambda')$  is feasible in the subsidy minimization problem, then there is  $(\alpha, I, J; t; \phi; \lambda)$ , which is also feasible and achieves the same objective value, but  $I = J$  and  $\underline{\alpha}_{ij} = \bar{\alpha}_{ij} = 0$  for  $i \neq j$ .

State $\underline{v}$	$s_1^s$	$s_2^s$	...	$s_I^s$	State $\bar{v}$	$s_1^s$	$s_2^s$	...	$s_I^s$
$s_1^b$	$\underline{\alpha}_1$	0	...	0	$s_1^b$	$\bar{\alpha}_1$	0	...	0
$s_2^b$	0	$\underline{\alpha}_2$	...	0	$s_2^b$	0	$\bar{\alpha}_2$	...	0
$\vdots$	$\vdots$	$\vdots$	$\ddots$	$\vdots$	$\vdots$	$\vdots$	$\vdots$	$\ddots$	$\vdots$
$s_I^b$	0	0	...	$\underline{\alpha}_I$	$s_I^b$	0	0	...	$\bar{\alpha}_I$

**Proof:** merge signal realizations with equal posteriors.

**Two design concerns for the principal: IC and total cost of info.**

- **IC:** More correlated signals  $\Rightarrow$  easier to incentivize truthful reporting.
- **Total cost:** Less correlated signals  $\Rightarrow$  more info at lower cost.

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### Corollary (Perfect correlation)

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State $\underline{v}$	$s_1^s$	$s_2^s$	...	$s_I^s$	State $\bar{v}$	$s_1^s$	$s_2^s$	...	$s_I^s$
$s_1^b$	$\underline{\alpha}_1$	0	...	0	$s_1^b$	$\bar{\alpha}_1$	0	...	0
$s_2^b$	0	$\underline{\alpha}_2$	...	0	$s_2^b$	0	$\bar{\alpha}_2$	...	0
$\vdots$	$\vdots$	$\vdots$	$\ddots$	$\vdots$	$\vdots$	$\vdots$	$\vdots$	$\ddots$	$\vdots$
$s_I^b$	0	0	...	$\underline{\alpha}_I$	$s_I^b$	0	0	...	$\bar{\alpha}_I$

**Proof:** merge signal realizations with equal posteriors.

**Two design concerns for the principal: IC and total cost of info.**

- **IC:** More correlated signals  $\Rightarrow$  easier to incentivize truthful reporting.
- **Total cost:** Less correlated signals  $\Rightarrow$  more info at lower cost.

**IC overwhelmingly dominates  $\Rightarrow$  pay for the same info twice!**

# Subsidy minimization as Bayesian persuasion

$$\max_{\{\tau, \underline{\mu}; \Lambda\}} \sum_{i=1}^I \tau_i T(\underline{\mu}_i, \bar{\mu}_i; \Lambda^b, \Lambda^s)$$

$$(BP) \quad \sum_{i=1}^I \tau_i \underline{\mu}_i = \underline{\mu}_0, \quad \sum_{i=1}^I \tau_i \bar{\mu}_i = \bar{\mu}_0;$$

$$(NA^b) \quad \exp(-\underline{\Lambda}^b) + \exp(-\bar{\Lambda}^b) = 1,$$

$$(NA^s) \quad \exp(-\underline{\Lambda}^s) + \exp(-\bar{\Lambda}^s) = 1.$$

where  $\tau_i = \underline{\alpha}_i + \bar{\alpha}_i$ , and  $\underline{\Lambda}^p = \min_i \{\underline{\lambda}_i^p\}$  and  $\bar{\Lambda}^p = \min_i \{\bar{\lambda}_i^p\}$ .

**For a fixed  $\Lambda$ , this is a Bayesian persuasion problem**

$\Rightarrow$  **look at concave closure of  $T$ .** Concave closure of  $T$

# Roadmap

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# Concluding remarks

- Bilateral trade problem with **information acquisition**.
- Information acquisition is **costly** and **flexible**.
- Tractable characterization of **implementability**.
- Characterization of info structures consistent with **efficient trade**.
- **Subsidy minimization** for efficient trade.

# Appendix

6 Implementability for the seller

7 Subsidy minimization as Bayesian persuasion: solution

# Implementability for the seller

## Lemma (Implementability for the seller)

$(\alpha, q, t)$  is globally implementable for the seller iff there are multipliers  $\lambda_i^s(v)$  for all  $s_i^b \in S^b$  and  $\phi_{ij}^s(v)$  for all  $(s_i^b, s_j^s) \in S^b \times S^s$  and all  $v \in V$ :

$$(ST^s) \quad \underbrace{t_{ij}^s - q_{ij}^s u^s(v)}_{\frac{\partial U^s}{\partial \alpha_{ij}(v)}} - \underbrace{\log(\mu_j^s(v))}_{\frac{\partial C^s}{\partial \alpha_{ij}(v)}} - \lambda_i^s(v) + \phi_{ij}^s(v) = 0,$$

$$(DF^s) \quad \phi_{ij}^s(v) \geq 0,$$

$$(CS^s) \quad \alpha_{ij}(v) \phi_{ij}^s(v) = 0,$$

$$(NA^s) \quad \sum_{v \in V} \exp(-\min_i \{\lambda_i^s(v)\}) \leq 1.$$

- Analogous conditions apply to the buyer. Buyer's implementability

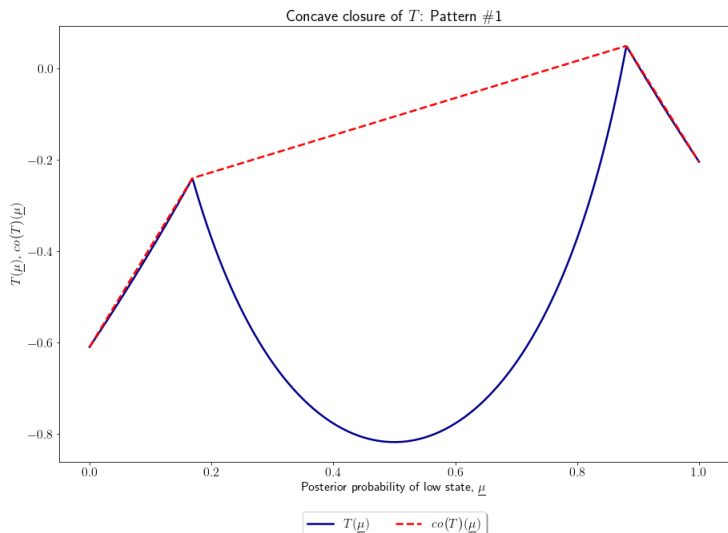


# Appendix

6 Implementability for the seller

7 Subsidy minimization as Bayesian persuasion: solution

# Concave closure of $T(\underline{\mu}, 1 - \underline{\mu}; \Lambda^b, \Lambda^s)$



# Optimality conditions

## Proposition (Optimality conditions)

*If the subsidy minimization problem achieves a minimum, then we can set  $l = 2$  w.l.o.g., and moreover the optimal posteriors satisfy*

$$(\text{Opt}^b) \quad \underline{u}^b - \log(\underline{\mu}_1) - \underline{\Lambda}^b = \bar{u}^b - \log(\bar{\mu}_1) - \bar{\Lambda}^b,$$

$$(\text{Opt}^s) \quad \underline{u}^s + \log(\underline{\mu}_2) + \underline{\Lambda}^s = \bar{u}^s + \log(\bar{\mu}_2) + \bar{\Lambda}^s.$$

**Combine (Opt) with (NA) to solve for  $\Lambda$  and plug into the objective  $\Rightarrow$  unconstrained problem for posteriors.**

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