# Data and Privacy: Tools from Information Design

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Personal data has become key input of the modern economy

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- ▶ Implications of collecting, trading, using personal data on economics outcomes, e.g., consumer welfare, market power, etc.
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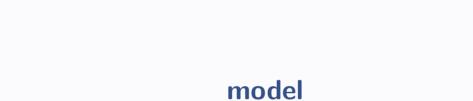
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#### $\textbf{Today} \Rightarrow \textbf{A} \ \textbf{Methodological Point}$

How/why tools from **information design** constitute a building block to study questions concerning the **economics data** (plus some future directions)



#### The typical **Information Design** setting:

- A finite set  $\Omega$  of payoff states  $\qquad \qquad \text{(e.g., demand condition)}$
- A common prior belief of about  $\omega$ , denoted  $q \in \Delta(\Omega)$
- $\ \, {\rm One} \ \, {\rm ``designer,''} \ \, N \ \, {\rm agents} \qquad \qquad ({\rm e.g., \ e-com \ platform \ and \ merchants})$
- Finite action sets  $A_0 \times A_1 \times \ldots \times A_N =: A$  (e.g., price, feature, quality)
- Payoffs:  $v:A\times\Omega\to\mathbb{R}$  for the designer;  $\pi_i:A\times\Omega\to\mathbb{R}$  for each agent

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  - Base Game:  $G = (\Omega, q, N, (A_i, u_i)_{i \in N})$
  - An information structure is a pair  $(S,\pi)$  s.t.  $S=S_1\times\ldots\times S_N$  (finite) and  $\pi:\Omega\to\Delta(S)$

- Base game + information structure:  $(G,(S,\pi))\Rightarrow$  a Bayesian game
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The **information-design** problem:

$$V(q) = \max_{(S,\pi)} \ \max_{\sigma \in \mathsf{BNE}(G,(S,\pi))} \sum_{\omega,s,a} v(a,\omega) \sigma(a|s) \pi(s|\omega) q(\omega)$$

# **Applications**

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Limiting assumptions of the standard model:

Commitment power, equilibrium selection, inherently static

## **Obedient Recommendation Mechanisms**

The ID problem can be equivalently formulated as choosing an obedient direct recommendation mechanism  $x:\Omega\to\Delta(A)$ : Bergemann and Morris '16, TE

$$\begin{split} V(q) &= \max_{x:\Omega \to \Delta(A)} \sum_{\omega,a} v(a,\omega) x(a|\omega) q(\omega) \\ &\text{such that, for all } i \in N, a_i, a_i' \in A_i \\ &\sum_{\omega,a_{-i}} \Big( \pi_i(a_i,a_{-i},\omega) - \pi_i(a_i',a_{-i},\omega) \Big) x(a_i,a_{-i}|\omega) q(\omega) \geq 0 \end{split}$$

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A finite-dimensional linear program ⇒ Quite tractable

The typical object of interest in the ID literature: Characterize the optimal  $\boldsymbol{x}$ 

# a different perspective

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A database of data records:  $q = (q(\omega))_{\omega \in \Omega} \in \mathbb{R}^{\Omega}_+$ 

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#### Examples of Questions

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- What is the value for the intermediary of each single data record in q?
  - A benchmark for compensating individuals for their data
- What are the properties of the "demand function" for data

Law of demand — as  $q(\omega)$   $\nearrow$ , value of  $\omega$ -records  $\searrow$ 

Indifference curves — When are records complement/substitute?

Merging databases: When is V(q + q') > V(q) + V(q')?

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Galperti, Liu, Perego '24 (For more details on this, come to my talk!)

Sunday @8am, Session title: Competitive Implications of Data Sharing

conclusions

## **Conclusions**

- ► A large literature in **information design** provides natural framework and powerful tools to study questions concerning the **economics of data**
- ► Information design as a production problem:

*Input:* Personal Data → *Output:* Optimal Information

- ► A new perspective: How does changing inputs affects economic outcomes (through changes in optimal information)?
- ► This is key for studying: The value of data; The effects of privacy protection policies; The role of data unions; etc.
- ► In my view, this is a natural direction for this literature: lots of open questions and high demand for better theory

an example

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- $\omega_H$  reveals consumer has valuation 2

Platform's database contains:

3 million such records

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Platform is an **intermediary** that provides the merchant with **information** about each consumer, and thus can influence the price it charges to them

Merchant chooses price  $a \in A$  given information received. He maximizes profits:

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Suppose platform choose information to maximizes consumer's surplus

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**Question:** How much value does platform derive from each record in q?

The (unique) optimal recommendation mechanism  $(x_q:\Omega\to\Delta(A))$  is:

Type-
$$\omega_L$$
 records  $a=1$ 
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Thus, platform's expected **payoff** from each record is  $u_q^*(\omega) = \begin{cases} 0 & \text{if } \omega_L \\ \frac{1}{2} & \text{if } \omega_H \end{cases}$ 

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Are  $\omega_L$ -records worthless? No, value of data records is  $\psi_q^*(\omega) = \left\{ egin{array}{ll} 1 & \mbox{if } \omega_L \\ 0 & \mbox{if } \omega_H \end{array} \right.$ 

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— **Approach.** Platform uses inputs (data records) to produce outputs (recommendations). GLP '23 use LP duality to **characterize** the **values** of these inputs, namely  $\psi_q^*(\omega)$  – platform's willingnes to pay for an additional  $\omega$ -record