Dynamic Privacy Choices

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Motivation



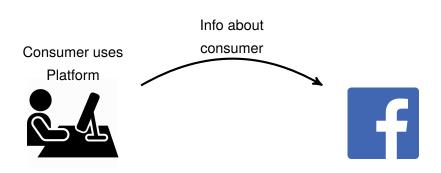


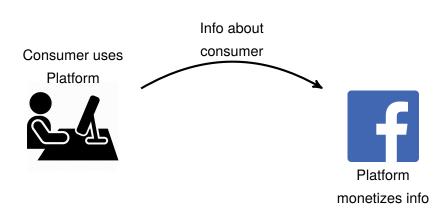


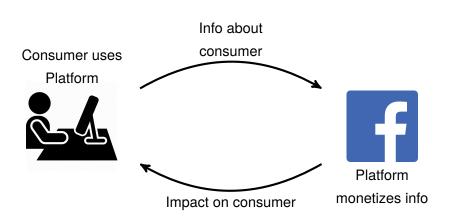
Consumer uses

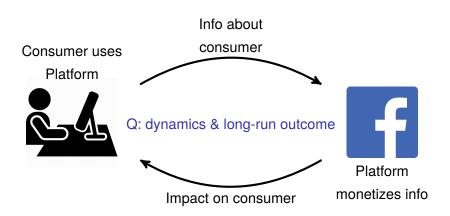














- 1. Model
- 2. Results
- 3. Relaxing Commitment Assumption
- 4. Literature



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Time t = 1, 2, ...

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Consumer

• Type $X \sim \mathcal{N}(0, \sigma_0^2)$, fixed, unobservable¹

¹If privately observable, focus on a "pooling" equilibrium.

Time t = 1, 2, ...

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- Choose an activity level $a_t \in A \subset \mathbb{R}_+$
- A is finite, $\min A = 0$, and $\max A = a_{max} > 0$

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Platform

• Privately observe a signal $X + \varepsilon_t$ with $\varepsilon_t \stackrel{iid}{\sim} \mathcal{N}\left(0, \frac{1}{a_t} + \gamma_t\right)$

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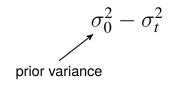
- ▶ Privately observe a signal $X + \varepsilon_t$ with $\varepsilon_t \stackrel{iid}{\sim} \mathcal{N}\left(0, \frac{1}{a_t} + \gamma_t\right)$
- γ_t : level of privacy protection in t

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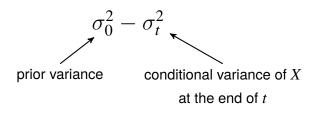
Platform's payoff in period t

$$\sigma_0^2 - \sigma_t^2$$

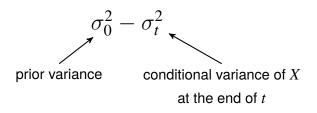
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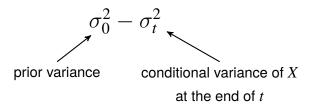


Platform's payoff in period t



More info better

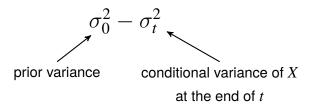
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More info better

• Increasing in (a_1, \ldots, a_t) and decreasing in $(\gamma_1, \ldots, \gamma_t)$

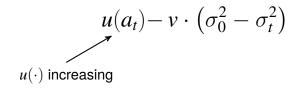
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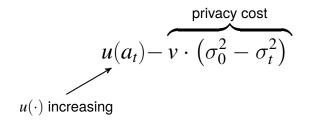


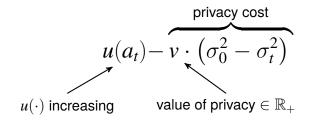
More info better

- Increasing in (a_1, \ldots, a_t) and decreasing in $(\gamma_1, \ldots, \gamma_t)$
- Discount future payoffs

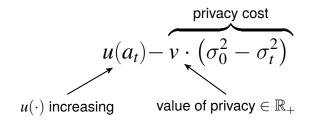
$$u(a_t) - v \cdot \left(\sigma_0^2 - \sigma_t^2\right)$$







Consumer payoff in period t



Discount future payoffs

Timing

1. Platform chooses a privacy policy $(\gamma_1, \gamma_2, \dots) \in \mathbb{R}^{\infty}_+$

• Signal
$$X + \varepsilon_t$$
 with $\varepsilon_t \sim \mathcal{N}\left(0, \frac{1}{a_t} + \gamma_t\right)$

2. Consumer chooses a_1, a_2, \ldots

Solution: SPE



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Flow Payoffs

Consumer's period-t payoff

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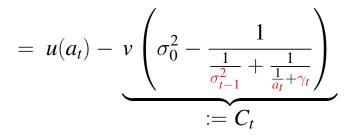
$$u(a_t) - v\left(\sigma_0^2 - \sigma_t^2\right)$$

$$= u(a_t) - v \left(\sigma_0^2 - \frac{1}{\frac{1}{\sigma_{t-1}^2} + \frac{1}{\frac{1}{a_t} + \gamma_t}} \right)$$

Flow Payoffs

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Marginal Privacy Cost

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$$\frac{\partial C_t}{\partial a_t}$$
 is decreasing in γ_t and increasing in σ_{t-1}^2 .

- ► Less privacy (lower σ_{t-1}^2) \rightarrow Lower marginal cost
- Lower payoff \leftrightarrow Higher incentive to raise a_t

- 1. Platform chooses a privacy policy $(\gamma_1, \gamma_2, \dots)$
 - Signal $X + \varepsilon_t$ with $\varepsilon_t \sim \mathcal{N}(0, \frac{1}{a_t} + \gamma)$
- 2. Consumer solves

$$\max_{(a_t)_{t=1}^{\infty}}\sum_{t=1}^{\infty}\delta_C^{t-1}\left[u(a_t)-v\left(\sigma_0^2-\sigma_t^2(\boldsymbol{a}^t,\boldsymbol{\gamma}^t)\right)\right].$$

Reminder: $u(a) - v \cdot (\sigma_0^2 - \sigma_t^2)$

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Theorem

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$$\sigma_t^2 \to 0$$
 and $a_t^* \to a_{max}$ as $t \to \infty$
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• Early: high MC \rightarrow high γ_t to encourage activity

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- Early: high MC \rightarrow high γ_t to encourage activity
- Learning becomes easier over time
- No value of stopping data collection
 - E.g., $\gamma_t = \infty$ after some period?
 - Committing to erode privacy \rightarrow higher activity today

Generalization

Platform's payoff is strictly increasing in $(a_t, -\sigma_t^2)$

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Proposition

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$$\lim_{\delta_P \to 1} \lim_{t \to \infty} \sigma_t^2 = 0 \quad \text{and} \quad \lim_{\delta_P \to 1} \lim_{t \to \infty} a_t^* = a_{max}.$$

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- High activity if high γ_t or low σ_t^2
- Activity-driven platforms benefit from collecting data

Implications

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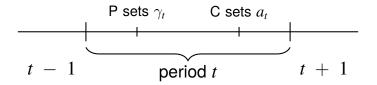
- 1. Privacy paradox (cf. Acquisti et al. 2016)
- 2. Rational addiction (Becker and Murphy, 1988)



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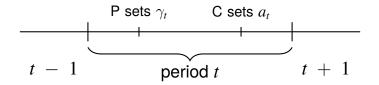
Relaxing Commitment Assumption

Platform with "one-period commitment"



Relaxing Commitment Assumption

Platform with "one-period commitment"



Assumption

Binary activity level: $A = \{0, a_{max}\}.$

Consumer-Worst Outcome

Full characterization in the paper

Proposition (informal)

There is a "consumer-worst" eqm such that:

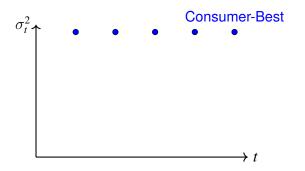
- 1. The outcome is the same as long-run commitment.
- 2. Platform strategy is greedy.
- If σ_0^2 is small, the eqm is unique.

Proposition

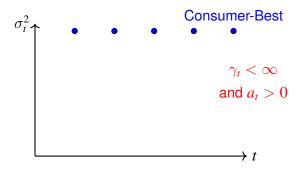
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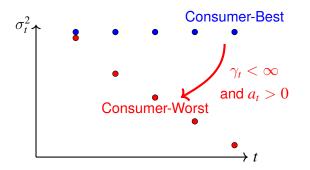
Proposition



Proposition



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Introducing a New Digital Product

Two firms

- Existing firm with a low σ_0^2 (e.g., data from other services)
- New firm with a high σ_0^2

Which firm has higher willingness to launch a new digital service?

New firm faces a higher marginal value of info

But, platform-worst eqm \rightarrow only the existing firm can collect info

Inefficiency: Data go to a firm that already has a lot of data

Literature (not exhaustive!)

Platform data collection: Acemoglu et al. (2019); Bergemann et al. (2019); Choi et al. (2018); Garratt and van Oordt (2019)

Competition with data: Cornière and Taylor (2020); Prufer and Schottmüller (2017); Hagiu and Wright (2020)

Switching cost, barrier to entry: Farrell and Shapiro (1988); Klemperer (1995); Fudenberg and Tirole (2000)

Signal-jamming: Holmstrom (1999)



- A dynamic model of a platform collecting consumer data
- Key: decreasing marginal privacy cost
- Long-run privacy loss with high activity level
- Weaker commitment: optimistic belief prevents data collection
- Data-driven advantage due to lower MC of privacy loss