

The Economics of Privacy: An Agenda

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Abstract

This paper boldly attempts to set forth an agenda of topics that seem important to study in the economics of privacy in the future.

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1 The Challenge of Privacy For Economics

The Economics of Privacy is a challenging field in which to be an economist. There are two reasons for this:

The first challenge stems from the definition of privacy is. What is privacy? My favorite definition is ‘freedom from unwarranted intrusion.’ This definition stems from (Warren and Brandeis, 1890) who defined privacy as the ‘right to be left alone’. Warren and Brandeis (1890) were famously inspired to write their influential essay by the rise of gossip columnists capturing photos with the new technology of portable cameras. This is important as I will argue in this essay that our conceptions of privacy, and therefore what is important to study as economists, are inextricably tied up with progress in technology. Concepts of privacy are constantly challenged by new technologies that parse personal information in new and unexpected ways. Therefore unlike a field such as health economics where the definition of what health is - is relatively unchanging - our ideas of what privacy is and should be are in constant flux.

The second challenge stems from our need as economists to at essence conceptualize any question in economics through the lens of a utility function. Farrell (2012) describes the issue very usefully. Typically in the theoretical literature in economics we tend to think about intermediate preferences for privacy - these reflect the anticipation that if we share our data with a firm it can be used potentially for things like price discrimination which harm us. By contrast, the vast majority of the literature outside of economics think about privacy as a right or something where people should just have a fixed intrinsic taste for keeping certain types of information private. Indeed, often the descriptions of tastes for privacy outside of economics suggest a distaste for creepiness (Richards and Hartzog, 2015), or a taste for data being only used in the same context (Nissenbaum, 2004). While of course a taste for anything can be included in a utility function, it is unsatisfactory for a discipline that has

tried to always model utility functions based on first principles.

2 The Focus of the Economics of Privacy Literature So Far

When trying to write an essay on the economics of privacy, it is important to highlight that this has already been done stupendously well by Professor Alessandro Acquisti of Carnegie Mellon University and coauthors, culminating in an essay published in the Journal of Economics Literature (Acquisti et al., 2016). What is attractive about this conception of the history of privacy is that he defines a variety of decades of schools of thought and how this has progressed over time.

The first wave identified by Acquisti et al. (2016) is that of the Chicago School in the 1970s, led by theorists such as Stigler and Posner. In this literature, privacy was defined as a propensity towards secrecy - and in a world where information is generally beneficial to welfare, these models evaluated how tastes for privacy itself could lead to harm to welfare (Posner, 1978, 1981; Stigler, 1980). Of course the wave of information economics that characterized theory in the 1980s in economics, itself questioned the idea that more information is always beneficial initiated by some of the idea in theories of signaling and information cascades (Spence, 1978; Hirshleifer, 1978).

The second wave identified by Acquisti et al. (2016) is also a theoretical literature but this time led by information economists who were interested in questions of technology. Varian (2002) shifted the question of privacy from being one of simply secrecy in what information is shared with other people, to being one firmly about data. This led to new questions such as what should be secondary use rights associated with data. As such it gave rise to what Acquisti et al. (2016) describe as the third wave of theoretical literature which is interested in questions such as price discrimination (though the use of cookies) (Acquisti and Varian, 2005) and targeting effects in online advertising (Johnson, 2013; Bergemann and Bonatti, 2011, 2015).

The other large shift in the last decade of research has been a proliferation of empirical work in privacy. As described by Goldfarb and Tucker (2012a), much of this work has tried to quantify the effects of privacy regulation on the economy, much of the literature asking questions advertising markets (Goldfarb and Tucker, 2011; Chiou and Tucker, 2012; Johnson et al., 0; Jia et al., 2018; Peukert et al., 2020; Johnson et al., 2022; Godinho de Matos and Adjerid, 2022), though some of the literature also asking about online behavior (Zhao et al., 2021), financial markets (Kim and Wagman, 2015), and health markets (Miller and Tucker, 2009, 2011; Adjerid et al., 2016; Miller and Tucker, 2017).

These few brief paragraphs do not of course do justice to the literature on the economics of privacy. However, it is fair to say that as yet the number of researchers and number of publications are relatively small given its potential importance in the digital economy. Recently recognizing this the NBER, and with support from the Sloan Foundation, has instituted a one-off conference on the economics of privacy and also a PhD tutorial to try and inspire more work in this area. This chapter of the handbook that reflects this work tries to offer some suggestions about how economists might be able to deepen and broaden this current literature.

3 Outstanding Questions

This handbook is aimed at young researchers who are starting off their careers. Therefore it makes sense to focus on some of the big questions that researchers in economics have not yet tackled (or have only tackled in part).

3.1 The Value of Privacy

3.1.1 Measuring Positive Consequences of Privacy Regulation

Much of the empirical wave of research on privacy has focused on the question of how privacy regulation hurts economic outcomes - by restricting advertising effectiveness (Goldfarb and Tucker, 2011; Johnson et al., 0), leading to market concentration (Peukert et al.,

2020; Johnson et al., 2022), exacerbating inequality (Kim and Wagman, 2015) or hurting health outcomes (Miller and Tucker, 2009; Adjerid et al., 2016; Miller and Tucker, 2017; ?). However, given the large literature on how privacy regulation has large negative economic consequences the paucity of literature on the benefits of privacy regulation is surprising.

Therefore, it may make sense for researchers to also think about situations or contexts where privacy rights and regulations might have clear positive consequences for individuals. Some I have thought of include:

- Data concerning reproductive health
- Data concerning mental illness
- Data concerning disability that might be used to disqualify potential employees from jobs they could do well
- Data concerning past crimes that are orthogonal to a current question that requires judgment

One thing which all these things have in common is that they concern questions where a stigma exists that is unrelated to potential economic output or the economic quality of a match. In such cases, if privacy regulation tempers data diffusion about something that has a irrational stigma, then it must be the case that privacy regulation benefits individuals. If this is the case, privacy regulation or privacy protections should have positive effects on consumer welfare.

Other occasions where it should be straightforward to document benefits from privacy regulation include instances where data itself might be used for coercion: This might include

- Targeting those who suffer from addictive behavior to pursue their addiction
- Targeting those who have struggled managing their credit in the past, with further unwise credit offerings

- Targeting those who suffer eating disorders, with weight loss products

These examples share the theme that if prompted an individual might pursue a course that is not ultimately utility-maximizing for them. As such a restriction of data that means they are not likely to be targeted with prompts may benefit them.

3.1.2 Measuring Tastes for Privacy

If we are to truly understand though whether privacy regulation has benefits to consumers we have to return though to measurement of key parameters in the utility function. If we assume that a taste for privacy is built into a consumers' utility function, than by definition any regulation that caters to this taste improves consumer welfare. However, this implies we have to actually measure relative intrinsic tastes for privacy. One of the first empirical papers that has attempted to do is Lin (2022). This paper finds that in general there is a lot of heterogeneity in intrinsic tastes for privacy and perhaps the magnitudes are smaller than might be expected given the privacy literature.

It is clear that the more than can be inferred about underlying tastes from privacy, given observed consumer choices over privacy decisions regarding their data the more informed this debate can be.

One issue which has thwarted attempts at measuring preferences for privacy is something know as the privacy paradox (Athey et al., 2017). This reflects the observed phenomenon that often while consumers express a desire for privacy when asked about it, it appears they are willing to share their data very readily in a way which seems to contradict this. Of course, in economics tension between stated preferences and revealed preferences are not new, and economists by disposition tend to trust more revealed preferences. But privacy is a domain where trying to unpack this tension appears worthwhile. How is the privacy paradox moderated by the knowledge of consumers? Does the privacy paradox ever reflect consumers engaging in some type of behavioral distortion which means their stated preferences are closer

to the truth? In what domains is the privacy paradox most important, and how does that affect our attempts to evaluate privacy regulation.

In general, what is clear is that from an economics perspective the more we can examine and model actual consumer behavior regarding privacy data using individual-level decisions the more we will be able to model and parse individual privacy preferences. In particular, the more we could have data on individual's decision making regarding the privacy of their data across different domains the more informative this may be.

3.1.3 Privacy Preferences and Context

An appealing theory for understanding some apparent disconnects in privacy preferences is that of the idea of contextual integrity (Nissenbaum, 2010). This states that privacy preferences or intrinsic tastes for privacy can be understood as reflecting five contextual parameters that help shape the view of privacy of the individual. These span who the sender of the data is, who the subject of the data is, who the recipient of the data is, the type of data that is sent, and what is referred to as the 'transmission' principle which reflects whether the data was obtained via consent, coercion, or by sale, or by law. Dr Nissenbaum is a philosopher meaning much of the work is conceptual. As such it reflects a potentially rich testing ground for different theories of tastes for privacy (Bleier et al., 2020): For example:

1. How much does the same person have different tastes for privacy depending on the recipient? Or the type of data?
2. How does the original context in which data was given affect privacy principles?
3. Do we have similar privacy preferences over our associates data as we do our own?

Mapping out all these parameters that affect privacy preferences conceptually, seems a very useful exercise for empirical analysis.

3.1.4 Time-Inconsistency and Privacy Preferences

One important question which appears to have been neglected in the literature is the question of how privacy preferences evolve over time. This is particularly striking because the drop in costs of storing and parsing digital data means that it is virtually costless to store an individual's history of actions over time, rather than periodically deleting it. In other words, there is no reason to think that the . However, there is evidence that (Goldfarb and Tucker, 2012b) people's privacy preferences evolve as they grow older, that is as people get older they get more privacy conscious. This means that data that young people create today, may not reflect their privacy preferences when they get older which may have negative consequences. In addition, it is of course possible that there are technology shocks which mean that there are unanticipated consequences of sharing data. For example, I might have made decisions about sharing video footage of myself without predicting that advances in machine learning could lead such data to be decomposed in a manner which allows seamless prediction.

3.1.5 Inferential Privacy

Much of the privacy debate has focused on issues of data. And indeed this reflects law-making - most laws regarding privacy do not mention the use of algorithms or how that might affect privacy considerations. However, this could be an artifact of laws generally being backwards-looking rather than a prediction of the future.

If I was to speculate I would argue that in the future we may see a realization that many potential privacy concerns are not a result of the data itself being transferred, but instead a result of predictions that are made using this data. For example, though I might be happy to share my photos publicly, and the photos themselves not cause me any privacy concerns, if algorithms were able to make predictions from these photos about my health, my financial status, my fertility or other domains of data that I considered private then I might want to object to using the data in that manner.

If this prediction comes to pass then, this opens up multiple different avenues for research. Indeed, there are already theory papers that are exploring these topics (Acemoglu et al., 2019, 2022; Bergemann et al., 2020; Goldfarb et al., 2020).

3.2 Markets and Privacy

3.2.1 Property Rights and Privacy

Whenever economists who have not studied the economics of privacy give interviews about privacy they tend to immediately and instinctively talk about property rights. After all, one of the central tenets of economics is the coase theorem, (Coase, 1960) which suggests that many instances of inefficiencies in information markets can be solved by simply clarifying property rights (Farrell, 1987). And the idea that all the tensions involved by trying to optimize privacy protections can be best solved by property rights is superficially an attractive one. And indeed my own research has shown that giving controls to people over their privacy - perhaps akin to property rights helps address privacy concerns (Tucker, 2014; Miller and Tucker, 2017).

However, there are ultimately obvious flaws in thinking that property rights alone can address privacy concerns, which are themselves worth exploring as potential research topics:

- The idea that data is neatly binary does not fit current data markets. Instead, it makes sense in a world of spreadsheets where each person's data is neatly encased in a single row of data. Take for example, a photo I take of myself in a shopping center. This photo might - through facial recognition technologies - also place other individuals at that shopping center. However, even though I might have taken the data and therefore own the data, it is not clear that I have property rights over anyone else's image that might appear in the photo.
- When I take a genetic test, and create data, I am creating data that might affect my ancestors and my descendants. Though, I might be able to sell my genetic data to an

interested firm, what should be done about the spillovers this has and inferences that are created for my family members?

- Often my data is not particularly valuable, however, inferences from it may be (?). Let us suppose I liked curly fries on Facebook and researchers were able to infer that this implied I was clever. Do I as the owner of the data also own rights to this inference - or to property rights to that inference belong to the researchers? As an aside, this correlation is based on real-life research (Kosinski et al., 2013).

Therefore, perhaps a way for research in this area to succeed is to study the differences between data where there is a clear property right, and data where there is not. And understand the economic implications of both. This is an area where it seems to be that the talents of theorists would be particularly helpful.

3.2.2 Individual Data Markets

Though it is possible to think of all the ways that property rights being fuzzy when it comes to data as being a potential explanation for why rights of property right approaches to privacy have failed, it is also possible to think of more traditional sources of market failure such as moral hazard and adverse selection also being at play. A useful place to study this is in current efforts to build up individual data markets. There are plenty of firms who have sought to set up businesses which would allow individuals to own their data and trade it for monetary value. For example, firms like <https://www.citizenme.com/>, <https://www.streamlytics.co/> and <https://www.clture.io/> have tried to establish individual data markets along these lines. Firms, like brave offer to pay people for their attention and data.¹. However, as of yet none of these efforts have thrived.

There is a fledgling literature that tries to understand some of the limitations from a privacy perspective of these markets (Spiekermann et al., 2015). There is also a theoretical

¹<https://brave.com/compare/chrome/earning/>

literature that explores the consequences of these markets not existing (Jones and Tonetti, 2020), being distorted by regulation (Fainmesser et al., 2022) or being plagued by externalities (Ichihashi, 2021). But it seems clear that more papers are needed that tries to study the diffusion of these attempts to create data markets and issues of adverse selection and moral hazard that might intuitively plague attempts to create such markets.

Another explanation that may be worth exploring is also that the ubiquity of data and non-rivalry of data has also hampered the successful monetization of an individual's data.

3.2.3 Competitive Dynamics and Privacy

It is also useful to think about privacy regulation or tastes for privacy might affect market dynamics and competition as a whole. Early theoretical work such as Campbell et al. (2015) sketched out theoretical reasons why privacy regulation might lead to concentration. Since then, a variety of work has appeared to confirm this (Miller and Tucker, 2014; Peukert et al., 2020; Johnson et al., 2022; Marthews and Tucker, 2019a). However, this doesn't mean that the topic is closed to new research. Instead, it means it is time to broaden the number of contexts that such studies are conducted in - for example extending the insights to less studied industries where privacy matters - such as educational technology.

It is also possible to take this type of research and ask questions that illuminate competitive strategy. For example, it would be useful to study where an differentiation on the privacy dimension is a successful strategy, or whether as appears to have been the case so far, that it ultimately a niche strategy. What types of privacy regulation might be most successful and curtailing the market power of firms, where their market power stems from data? For example, in the fledgling genetic and genomic health industry can privacy regulations be designed in a way which will not cement market power for an incumbent?

3.2.4 The Market for Privacy Enhancing Technologies

Just as technology has led to an increase in privacy concerns, there has also been an increase in the use of technologies to help individuals and firms institute privacy protections. In general the work on economics that has considered the spread and importance of these technologies has focused on ad-blocking software (Shiller et al., 2018; Gritkevich et al., 2022). However, this vastly understates the breath and depth of these technologies - especially the extent to which they are used by firms. The new suite or stack of technologies are often referred to by the label of 'privacy-enhancing' technologies. As the Office of Science and Technology Policy recently said²:

Privacy-Enhancing Technologies (PETs) present a key opportunity to harness the power of data and data analysis techniques in a secure, privacy-protecting manner. This can enable more collaboration across entities, sectors, and borders to help tackle shared challenges, such as health care, climate change, financial crime, human trafficking, and pandemic response. PETs can also help promote continued innovation in emerging technologies in a manner that supports human rights and shared values of democratic nations, as highlighted during the Summit for Democracy in December 2021, which included an announcement that the United States and the United Kingdom are collaborating to develop bilateral innovation prize challenges focused on advancing PETs.

Such statements make it clear that policy makers believe that these technologies may help unravel the traditional tradeoff between privacy regulation and economic efficiency documented by economists. Therefore, it makes sense for economists to both explore the extent to which such privacy-enhancing technologies are successful at achieving these aims

²<https://www.federalregister.gov/documents/2022/06/09/2022-12432/request-for-information-on-advancing-privacy-enhancing-technologies>

and also any barriers that exist to their adoption. Indeed there are a whole set of technologies designed to help firms manage their privacy internally such as ‘consent managers’ ‘privacy assessment managers’ and ‘de-identification tools.’³ Economists are well placed to assess the extent to which these tools actually enhance privacy of customer, grounding such research on the insights of both organizational economics and enterprise-level diffusion of technology. It is also worth exploring the extent to which such tools reflect the deadweight welfare loss of document privacy regulation compliance relative to actual enhancements of privacy protection for consumers.

3.2.5 Decentralized Markets

The twin popular waves of cryptoeconomics and web 3 emphasize the emergence of decentralized markets. As such any discussion of markets and data should at least consider the potential consequences of decentralization of markets on privacy.

In general, I have expressed some skepticism about the extent to which many of the underpinning technologies or principles of blockchain technologies are naturally privacy enhancing (Marthews and Tucker, 2019b, 2022). In particular the qualities of verifiability of data and immutability of data that are inherent in the blockchain, appear to both restrict users’ ability to control their data or privacy principles such as the right to be forgotten.

However, it is certainly the case that firms and individuals within this community are hopeful that there are potential ways of using these technologies to enhance privacy. For example, firms like <https://www.meeco.me/platform> and <https://github.com/solid/solid> are both hoping to establish privacy-compliant data markets and data ownership structures. These new technologies and these new settings themselves present opportunities for researchers - given the promise that they themselves will generate data which allow us to study privacy-related behaviors and underlying preferences.

³<https://www.trustradius.com/data-privacy-management>

3.3 The Broader Economy and Privacy

3.3.1 Privacy and Inequality

There appears to be a positive correlation between privacy concerns, the enactment of privacy regulation and GDP. We also know that when we use proxy measures for privacy concerns such as sign ups to the do not call list - an anti telemarketing innovation - that this proxy for privacy preferences correlates with household income and is negatively correlated with demographic group indicators that have been historically disadvantaged in the USA (Varian et al., 2005).

However, despite these striking correlations there is little work that tries to understand why there is this relationship between economic prosperity and privacy concerns. Correspondingly, there is little work which investigates whether there are distributional consequences of privacy protections or privacy regulations. For example, one provocative way of thinking about the advertising-supported internet is that it is hugely redistributional. Rich people's data is valued by advertisers, and it is these high valuations which allow advertising-supported platforms to supply their services to free to many low income households both domestically and internationally. Privacy regulation might restrict this redistribution. Whether or not the reader agrees with this rather provocative characterization, it does suggest that the question of whether privacy regulation or protection has distributional consequences is an important one to answer. After all in economics we are interested in both studying phenomenon that affect efficiency but also equity.

Some initial research in this area has tried to at least establish some facts about how the scale and accuracy of data collection (Neumann et al., 2022). This suggests that low income households, less-educated households and renters are far less likely to have demographic information accurately filed and also actually have data available to be collected about them.

3.3.2 Privacy and Discrimination

In general, as discussed in this agenda, it has been difficult to measure and calibrate privacy harms. This may explain the shift in the policy debate towards questions of algorithmic bias or discrimination. Just by terminology alone, the potential for such phenomenon is alarming and also may reflect the untrammelled use of individual data by organizations and corporations. The growing algorithmic fairness literature is beyond the scope of this article - see (Cowgill and Tucker, 2019) for an overview of the topic from an economics perspective. However, it is useful to think about how and whether privacy regulation reduces, doesn't affect or augments the potential for algorithmic discrimination. It would seem from a theoretical perspective that any of these outcomes are possible. Privacy regulation might demand that firms reduce the amount personally identifiable information that is available - this might hinder firms and government's ability to audit their algorithms and identify instances of bias. Privacy regulation could also restrict the use of data by algorithms which give rise to algorithmic discrimination. Since the direction of the interaction between privacy regulation and algorithmic bias is unclear, this makes it an important area for empirical research.

3.3.3 Political Economy: Surveillance and Privacy

A clear gap in the focus of the current economics literature is that it has virtually all been focused on the question of commercial surveillance rather than government surveillance. However, the consequences of these two types of surveillance are not equal - unlike firms, governments can put you in jail and confiscate your property. As such understanding how the digital revolution has affected our conclusions about government and privacy is important.

There are two exceptions to this gap which are instructive. The first, is a paper that explores the extent to which government surveillance of commercial searches associated with the PRISM scandal had chilling effects on customer behavior (Marthews and Tucker, 2017).

This area of work is important as it suggests that knowledge and fear of government surveillance actions can shape the commercial landscape putting this topic squarely in the realm of microeconomics and industrial organization. Recent work by Beraja et al. (2020) examined the relationship between government surveillance and economic success but taking more of a macroeconomics perspective. This is useful as it suggests that understanding government surveillance can help shape our understanding of important macroeconomic questions such as growth and trade.

Indeed, there appear to be many questions to uncover when it comes to privacy and trade. In the US there have been multiple attempts to try and bridge trade barriers with the EU caused by among other things government surveillance. Indeed, the Safe Harbor Framework ended up having to be replaced by the Privacy Shield Framework due to the inadequacy of the current regime in complying with EU privacy regulations. ⁴. This uncertainty over trade and compliance has almost certainly had consequences that are important to study but have not yet been evaluated by economists.

4 The Future

This article has been an attempt to set an agenda in privacy. However, it is written by a researcher who has been working on these topics for two decades. She anticipates both that she has missed things that are important and also made many wrong predictions about what is important. As a result this article concludes by expressing excitement about what the research that young researchers who read this paper will do in the future.

⁴<https://www.ftc.gov/business-guidance/privacy-security/us-eu-safe-harbor-framework>

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Appendix

The FTC recently shared an Advance Notice of Proposed Rulemaking. Here are some illustrative questions. Answering these questions would indeed move the field forward

1. Which practices do companies use to surveil consumers?
2. Which measures do companies use to protect consumer data?
3. Which of these measures or practices are prevalent?
4. Are some practices more prevalent in some sectors than in others?
5. How, if at all, do these commercial surveillance practices harm consumers or increase the risk of harm to consumers?
6. Are there some harms that consumers may not easily discern or identify? Which are they?
7. Are there some harms that consumers may not easily quantify or measure? Which are they?
8. How should the Commission identify and evaluate these commercial surveillance harms or potential harms? On which evidence or measures should the Commission rely to substantiate its claims of harm or risk of harm?
9. Which areas or kinds of harm, if any, has the Commission failed to address through its enforcement actions?
10. Has the Commission adequately addressed indirect pecuniary harms, including potential physical harms, psychological harms, reputational injuries, and unwanted intrusions?

11. Which kinds of data should be subject to a potential trade regulation rule? Should it be limited to, for example, personally identifiable data, sensitive data, data about protected categories and their proxies, data that is linkable to a device, or non-aggregated data? Or should a potential rule be agnostic about kinds of data?
12. Which, if any, commercial incentives and business models lead to lax data security measures or harmful commercial surveillance practices? Are some commercial incentives and business models more likely to protect consumers than others? On which checks, if any, do companies rely to ensure that they do not cause harm to consumers?
13. Lax data security measures and harmful commercial surveillance injure different kinds of consumers (e.g., young people, workers, franchisees, small businesses, women, victims of stalking or domestic violence, racial minorities, the elderly) in different sectors (e.g., health, finance, employment) or in different segments or “stacks” of the internet economy. For example, harms arising from data security breaches in finance or healthcare may be different from those concerning discriminatory advertising on social media which may be different from those involving education technology. How, if at all, should potential new trade regulation rules address harms to different consumers across different sectors? Which commercial surveillance practices, if any, are unlawful such that new trade regulation rules should set out clear limitations or prohibitions on them? To what extent, if any, is a comprehensive regulatory approach better than a sectoral one for any given harm