

Article

Location privacy in the wake of the GDPR

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Abstract: The General Data Protection Regulation (GDPR) protects the personal data of natural persons and at the same time allows the free movement of such data within the European Union (EU). Hailed as majestic by admirers and dismissed as protectionist by critics, the Regulation is expected to have a profound impact around the world, including in the African Union (AU). For European–African consortia conducting research that may affect the privacy of African citizens, the question is ‘*how to protect personal data of data subjects while at the same time ensuring a just distribution of the benefits of a global digital ecosystem?*’ We use location privacy as a point of departure, because information about an individual’s location is different from other kinds of personally identifiable information. We analyse privacy at two levels, individual and cultural. Our perspective is interdisciplinary: we draw from computer science to describe three scenarios of transformation of volunteered/observed information to inferred information about a natural person and from cultural theory to distinguish four privacy cultures emerging within the EU in the wake of GDPR. We highlight recent data protection legislation in the AU and discuss factors that may accelerate or inhibit the alignment of data protection legislation in the AU with the GDPR.

Keywords: Location privacy; GDPR; European Union; inference; privacy cultures; African Union

1. Introduction

On May 25, 2018, two years after its enactment into law, the General Data Protection Regulation (GDPR), “*the most contested law in the E.U.’s history, the product of years of intense negotiation and thousands of proposed amendments,*” became enforceable in the European Union [1]. The GDPR protects the processing of personal data of natural persons and allows the free movement of such data within the European Union (EU). Unlike the Data Protection Directive (DPD) of 1995, which GDPR repealed, ‘Regulations’ are directly applicable as statutory law across the Union, and are not amenable to the opportunistic transposition of ‘Directives’ within individual Member States. A blatant example of a Member State transposing the DPD of 1995 and at the same time courting non-EU tech companies with weak enforcement and advantageous tax schemes was the Republic of Ireland [2]. Ireland’s transposition allowed tech companies to develop an EU site with such favorable conditions that Facebook shifted its headquarters for all of its global business outside North America to Ireland. Only in late April 2018, just before the GDPR took effect, did Facebook move more than 1.5 billion users out of Ireland and out of reach of the new law, despite Mark Zuckerberg’s promise to apply the spirit of the GDPR globally [3].

International flows of personal data are becoming more significant than ever. The EU trades with the US some \$260 billion worth of digital services annually, much of which involves personal data [4]. Data flows increased the global GDP, of which an estimated \$2.8 trillion represents data

34 flows, by \$7.8 trillion in 2014 [5]. These economic facts coupled with the data protection asymmetry
35 between countries explain much of the global interest in the GDPR, ever since its enactment in 2016.
36 International commentators' views of the GDPR are oscillating between ecstatic and dismissive. Daniel
37 Solove, a leading American scholar of privacy law, lauded the GDPR as the "*most profound privacy*
38 *law of our generation,*" "*majestic in its scope and ambition.*" He even professed his love for the GDPR,
39 because of the law's broad definition of personal data and its attention-grabbing penalties, among
40 other things [6]. Other non-EU analysts consider GDPR the "*most consequential regulatory development*
41 *in information policy in a generation*" [2], "*a new paradigm in data privacy*" [7], "*a real chance to renegotiate*
42 *the terms of engagement between people, their data, and the companies*" [8], and the thing that will bring
43 surveillance capitalism [9] to its knees [10]. On the other hand, American critics dismiss GDPR as
44 'EU protectionism' and for its ability to achieve important European geopolitical goals including
45 "(1) *solidifying legitimacy for Brussels during a period of deep skepticism among voters,* and (2) *strengthening*
46 *European political power against the real or perceived threat of American digital prowess*" [11, p. 236]. Most
47 importantly, they are unconvinced by the EU's independent data protection authority billing itself the
48 "*global gold standard*" [12] in data protection and call attention to the substantial historical and cultural
49 differences across nations that may inhibit exporting the GDPR as a one-size-fits-all approach to other
50 countries.

51 The latter critique is warranted. Like all human societies, the EU and the US have completely
52 different and deeply entrenched cultures of privacy. European privacy law protects human *dignity*,
53 a right rooted in the devastating experience with Fascism and Nazism, and in the ruling of
54 the German Federal Constitutional Court in its celebrated Census case of 1983 for informational
55 self-determination [13]. In the EU, the constitutional protection of dignity is anchored in Article 8(1)
56 "*Everyone has the right to the protection of personal data concerning him or her*" of the Charter of fundamental
57 rights of the European Union (2000/C 364/01). Privacy law in the US protects *freedom*, especially
58 freedom from intrusions by the state—not by private corporations—in the sanctity of one's own
59 home [14]. The word privacy is not mentioned in the US Constitution, except indirectly in the Fourth
60 Amendment, which prohibits the violation of "*the right of the people to be secure in their persons, houses,*
61 *papers, and effects, against unreasonable searches and seizures.*" One of the most significant constitutional
62 safeguards for information in the US concerns the free flow of data in the First Amendment's free
63 speech clause. Thus EU privacy legislation engages in a rights-based discourse centered on the dignity
64 of the 'data subject'—the individual citizen whose data is processed and whose information is at stake.
65 The US situates the individual squarely in marketplace relations trading her personal information in
66 free exchanges as a 'consumer,' exercising a free speech of sorts [4]. These fundamental differences in
67 privacy cultures between two ostensibly similar western polities suggest how difficult it may be for
68 the rest of the world to free ride on Europe's GDPR rules as the enthusiasts claim [15]. The differences
69 also suggest that a more complete account of privacy warrants tracking it at two levels, individual
70 and cultural [16,17]. To distinguish democratic from authoritarian societies, Westin [18]—probably the
71 most influential privacy scholar of the last century—also examines a third, political level of privacy,
72 which is beyond the scope of this study.

73 In this paper, we distinguish two levels of privacy—individual and cultural. We focus on
74 information privacy, and in particular on location privacy, or geoprivacy. The compound word location
75 privacy suggests that while control of location information is the central issue, location can be inferred
76 from people's interests, activities, and socio-demographics, and not only from 'traditional' location
77 information, e.g. geographic coordinates [19]. Focusing on location privacy is necessary because
78 "*information about an individual's location is substantially different from other kinds of personally identifiable*
79 *information*" [19, p. 5]. We define (location) privacy as the positive right to control the collection,
80 access, recording, and usage of an individual's (location) information and determine when, how, and
81 to what extent it is processed by others [20]. We distinguish between privacy as a negative right
82 (freedom from interference) and privacy as a positive right (freedom to control). This is because old,
83 pre-digital technologies—such as the instantaneous photographs and newspaper tabloids in Brandeis

84 and Warren’s time – restricted individuals to claiming privacy only as a negative right, as freedom from
85 interference, or ‘*the right to be left alone*’ [21]. New, digital technologies can reduce but also significantly
86 enhance privacy as a positive right, the freedom to control [20], but often in combination with social,
87 organizational and/or legal measures/strategies [22]. In the following sections, we first take the point
88 of view of Alice, an individual ‘data subject’ or ‘consumer.’ We show how Alice can control (part of)
89 the transformation process of volunteered and/or observed to inferred information and safeguard
90 her own as well as the privacy of her research subjects from attackers. We then discuss privacy at the
91 cultural level, starting from the basic premise that Alice’s (privacy) preferences and commitments are
92 shaped by and shape the culture of her community and society. Privacy preferences oscillate between
93 two extremes:

94 *“A high-privacy position assigns primary value to privacy claims, has high organizational distrust,*
95 *and advocates comprehensive privacy interventions through legal rules and enforcement. A*
96 *limited-privacy position views privacy claims as usually less worthy than business efficiency and*
97 *societal-protection interests, is generally trustful of organizations, and opposes most new regulatory*
98 *interventions as unnecessary and costly” [16, p. 434].*

99 We contribute to (location) privacy scholarship in two ways. We complement recent studies
100 (e.g. [19,23,24]) by analyzing (location) privacy at two levels—individual and cultural. We also
101 take a genuine interdisciplinary perspective. We draw from the field of (geo)computing to
102 describe the transformation of volunteered and observed to inferred information and to suggest
103 privacy-safeguarding measures. We draw from organization studies to dissect privacy into ideal types
104 of social relationships and strategies, and from cultural theory to distinguish ideal types of privacy
105 cultures. In the concluding section, we turn our gaze to (location) privacy in the African Union, a
106 polity where currently ongoing, intense legislative activity for personal data protection is matched by
107 an equally intense activity of data extraction from African-based organisations for expert analysis in
108 advanced economies [25,26]. Such an exploration, however tentative, is important to us, because of our
109 long-term engagement with African academia and government in joint research projects, involving
110 African ‘data subjects.’

111 **2. Privacy at the individual level**

112 *2.1. The individual data subject or consumer*

113 A privacy typology focused on the individual is both problematic and useful. It is problematic
114 at the most basic level of personal data emission, because an individual careless with her personal
115 data exposes information about herself as well as about others. If an algorithm knows her location
116 at a given time, it may predict the location of her spouse or friend. A child posting something about
117 her heart disease on social media may increase her parents’ health insurance premium [27]. It is also
118 problematic at the social level because it is the individual’s social environment that influences what
119 is deemed personal (data). If a society considers a given mode of personal behavior—e.g. political
120 opinion, sexual orientation, religious or philosophical beliefs, trade union membership (see Article 9(1)
121 of GDPR)—to be socially legitimate, only then is related data deemed personal [16]. On the other hand,
122 a privacy typology focused on the individual is useful. This is because the individual—as ‘data subject’
123 or ‘consumer’—is the subject of privacy theory and the bearer of the fundamental rights of dignity or
124 freedom. Privacy is a dynamic, ever-changing relationship, or a ‘negotiated relationship’ [28], between
125 an individual and her environment. It is present in all human cultures; what is culturally specific are
126 the strategies individuals and groups use to negotiate social interaction [29].

127 *2.2. A typology of privacy at the individual level*

128 At the heart of the privacy typology is Alice, a fictitious (geo)computing scientist, who adheres
129 to the ACM Code of Ethics and the rules of a GDPR-compliant European university. Alice values

130 (location) privacy as her positive right to control the collection, access, recording, and usage of her
 131 (location) information and determine when, how, and to what extent it is processed by others [20].
 132 At its simplest, Alice is related to her social environment in four ways—to another individual, to a
 133 group of individuals, to a private corporation and to a government institution—arranged in four cells
 134 of relations in Table 1. The incongruity of privacy goals between the related parties can be low or
 135 high and furnishes the horizontal dimension of the typology. The vertical dimension refers to Alice’s
 136 ability to control the transformation process of volunteered or observed personal data to inferred
 137 data referring to her or to her research subjects. Her ability is high when she can control the entire
 138 transformation process—the behavior of humans (incl. herself), of digital machines and of outputs. It
 139 is low when she can control some or none of these [30,31].

Table 1. A typology of (location) privacy relations

		Goal incongruity	
		<i>Low(er)</i>	<i>High(er)</i>
(Alice’s) Ability to control human behavior, machine behavior, outputs	<i>Low(er)</i>	Cell (4) Alice – Government institution Privacy strategy: Compliance; lodge complaint to DPA in case of violation of GDPR; anti-surveillance resistance	Cell (3) Alice – Private corporation Privacy strategy: Control behavior of corporation (via GDPR); lodge complaint to DPA in case of violation of GDPR
	<i>High(er)</i>	Cell (1) Alice – Bob Privacy strategy: Right and duty of partial display	Cell (2) Alice – (Bob – Carol – Dan – etc.) Privacy strategy: Geoprivacy by design

140 Before discussing each cell in detail, we draw attention to two obvious simplifications in Table 1.
 141 First, in Cell (3), Alice’s interaction with a private corporation, e.g. a location-based service (LBS)
 142 provider, involves not just the LBS provider but twelve other parties. These are the mobile device,
 143 the hardware manufacturer, the operating system, the operating system manufacturer, the mobile
 144 application, the mobile application developer, the core application, the third-party software, the
 145 third-party software developer, the LBS and the network operator and government [32]. Second, the
 146 boundary between Cell (3) and (4) is fuzzy. Government institutions often cooperate with private
 147 corporations. The US National Security Agency (NSA) obtained direct access to the systems of Google,
 148 Facebook, Apple and other big tech companies, as part of the Prism program, which allowed NSA
 149 officials to collect material including search history, the content of emails, file transfers and live
 150 chats [33]. Nevertheless, the four ideal types of relations help us draw a rough grid into which finer
 151 resolution grids may be inserted in future iterations.

152 In Cell (1), two humans (Alice and Bob) are interacting face to face in a private or public space.
 153 This is the archetypal human-to-human interaction. Both Alice and Bob are conscious of being observed
 154 by each other and other humans, and have similar privacy goals—to uphold a tacit social code, the
 155 ‘right and duty of partial display.’ The sociologist Erving Goffman [34] described how all humans
 156 reveal personal information selectively to uphold this code, while constructing their public personae.
 157 Hence, the low incongruity between Alice’s and Bob’s goals to protect their privacy—both strive to
 158 uphold this tacit social code, to protect (or curate) their public personae, but also modulate it gradually
 159 over time, as the relation expands or shrinks. As Fried [35] explains, Alice may not mind that Bob
 160 knows a general fact about her, and yet feel her privacy invaded if he knows the details. For instance,
 161 Bob may comfortably know that Alice is sick, but it would violate her privacy if he knew the nature of
 162 the illness. Or, if Bob is a good friend he may know what particular illness Alice is suffering from, but
 163 it would violate her privacy if he were actually to witness her suffering. Both control their behavior
 164 and the knowledge they share (outputs) about each other and may choose to modulate them over

165 time. Goffman's theory applies in settings where participants can see one another face to face, but
 166 it has implications for technology-mediated interactions, e.g. in email security [28]. When emailing
 167 each other, Alice and Bob may choose from a continuum of strategies to safeguard their privacy
 168 depending on context. They may refrain from emailing, they may email each other but self-censor,
 169 they may delegate privacy protection to mail encryption and firewalls, or they can work socially and
 170 organizationally to make certain that members of their community understand and police norms about
 171 privacy [36].

Table 2. Examples of measures controlling the transformation process

	Measures controlling human/machine behavior and outputs
Prior to start of campaign	human behavior (participation agreement, informed consent, institutional approval); outputs (define criteria of access to restricted data)
Security and safe settings	human behavior (assign privacy manager, train data collectors); machine behavior (ensure secure sensing devices, ensure secure IT system)
Processing and analysis	outputs (delete data from sensing devices, remove identifiers from data set)
Safe disclosure	outputs (reduce spatial and temporal precision, consider alternatives to point maps) human behavior (provide contact information, use disclaimers, avoid the release of multiple versions of anonymized data, avoid the disclosure of anonymization metadata, plan a mandatory licensing agreement, authenticate data requestors)

172 Cell (2) describes the interaction of a human, e.g. Alice, the research leader of a participatory
 173 sensing campaign, with a group of campaign participants (Bob, Carol, Dan, Eric, etc.). Goal incongruity
 174 between Alice and the group may be high, if the group members are not aware of possible breaches to
 175 their privacy and their implications. As campaign leader, Alice has a high ability to control outputs,
 176 behaviors of group members, as well as of machines, and takes a series of privacy-safeguarding
 177 measures for the entire group before, during and after the campaign, a strategy Kounadi and Resch [37]
 178 call 'geoprivacy by design.' They propose detailed privacy-preserving measures in four categories,
 179 namely, 6 measures prior to the start of a research survey, 4 measures for ensuring secure and safe
 180 settings, 9 measures for processing and analysis of collected data, and 24 measures for safe disclosure of
 181 datasets and research deliverables. Table 2 provides illustrative examples in each category. Interestingly,
 182 measures to control human behavior include two subtypes: outreach measures, e.g. participation
 183 agreement, and measures of self-restraint, e.g. use of disclaimers, avoiding release.

184 Cell (3) describes the interaction of Alice with a private corporation, as user of a location-based
 185 service (LBS), of which Google Maps is the most popular and commonly used. Alice volunteers her
 186 location to the LBS to get directions to a desired destination [32]. In this case, goal incongruity between
 187 Google and Alice is high, as we can see by comparing Alice's commitment to (location) privacy to
 188 Google's former executive chair Eric Schmidt. "If you have something that you don't want anyone to
 189 know, maybe you shouldn't be doing it in the first place." [38]. On the other hand, Alice's ability to control
 190 how her location information is used by the LBS to infer other information about her is low. As a EU
 191 citizen, she can rely on GDPR to (partly) control the behavior of the LBS provider. Another strategy is
 192 lodging a complaint to her national Data Protection Authority (DPA). DPAs are independent public
 193 authorities in each EU state that supervise the application of GDPR and handle complaints lodged
 194 against violations of GDPR. Max Schrems, the Austrian lawyer and privacy activist, is the face of
 195 GDPR complaint-lodging. His non-profit None of Your Business (NOYB) lodged four complaints
 196 about the take-it-or-leave-it practices of Google, Instagram, WhatsApp, and Facebook, the day GDPR
 197 became enforceable. He claimed that the platforms force users' consent to terms of use and demanded
 198 damages of \$8.8 billion. The French advocacy group La Quadrature du Net (LQDN) similarly filed

199 19 complaints. On January 21, 2019, the French National Data Protection Commission (CNIL) imposed
200 a financial penalty of €50 million against the company Google LLC, in accordance with the General
201 Data Protection Regulation (GDPR), for lack of transparency, inadequate information and lack of valid
202 consent regarding the ads personalization.

203 Cell (4) describes the interaction of Alice with government institutions. Alice trusts that her
204 government will respect her right to information privacy (thus goal incongruity is low) but may be
205 in the dark regarding the transformation process, unless a whistleblower leaks a secret surveillance
206 program (e.g. [33]) or the abuse of private data [39]. Further, if the public organization, where Alice
207 works, engages in processing likely to result in a high risk to the rights and freedoms of individuals,
208 Alice may lodge a complaint to the DPA and request a Data Protection Impact Assessment (DPIA). Such
209 processing may include the systematic and extensive evaluation of personal aspects of an individual,
210 including profiling, the processing of sensitive data on a large scale; or, the systematic monitoring
211 of public areas on a large scale. Or she may apply more covert techniques. The surveillance scholar
212 Gary Marx [40], a student of Erving Goffman, outlined 11 behavioral strategies intended to privately
213 subvert the collection of personal information and resist surveillance. He claims that

214 *“in spite of doomsday scenarios about the death of privacy, in societies with liberal democratic,*
215 *economic and political systems, the initial advantages offered by technological developments may be*
216 *weakened by their own ironic vulnerabilities and [by] ‘human ingenuity’” [40, p. 388].*

217 Another strategy for Alice is collective, e.g. participating in popular resistance to unpopular
218 government action. When the government of the Federal Republic of Germany announced a national
219 Census on April 27, 1983, German citizens protested so strongly that a dismayed German government
220 had to comply with the Federal Constitutional Court’s order to stop the process and take into account
221 several restrictions imposed by the Court in future censuses. Apparently, asking the public for personal
222 information in 1983, the fiftieth anniversary of the National Socialists’ ascent to power, was bad
223 timing, to say the least [41]. When the Census was finally conducted in 1987, thousands of citizens
224 either boycotted (overt resistance) or sabotaged (covert resistance) what they perceived as Orwellian
225 state-surveillance [42]. We should bear in mind that these remarkable events took place in an era
226 where the government was the only legitimate collector of data at such a massive, nationwide scale
227 and at a great cost (approx. a billion German marks). Nowadays, state and corporate surveillance are
228 deeply entangled. In response, technologically savvy digital rights activists have been influential in
229 several venues, including the Internet Engineering Task Force (IETF) and the Internet Corporation for
230 Assigned Names and Numbers (ICANN), through the Non-commercial User Constituency (NCUC)
231 caucus. Yet their efforts have largely remained within a community of technical experts (‘tech justice’)
232 with little integration so far with ‘social justice’ activists [43].

233 3. The transformation processes of our data in the context of location privacy

234 3.1. Data types in play

235 Having a *personal understanding* of one’s location privacy requires an understanding of regulations
236 in place, trust in regulation maintenance processes, as well as an understanding of the personal data
237 that is in play, and the inferencing capabilities that others may have once they have access to such data.
238 The regulations and their maintenance are important because they secure the boundary conditions,
239 under which a person can attempt to understand her information vulnerability. Below, we define
240 ‘personal data’ and provide a typology that helps to unravel the data in play. With that understanding,
241 we also address the mechanisms of inferencing over personal data, and possible countermeasures
242 available to the various actors: those that volunteer the data and those that receive such.

243 According to the GDPR, Article 4(1) [44], personal data is *“any information relating to an identified*
244 *or identifiable natural person.”* One can distinguish between three types of personal data: *volunteered,*
245 *observed* and *inferred*. The first type is typically explicitly handed over by the natural person ‘as part of

246 the deal'; the second type is captured by monitoring that natural person's actions and is much more
 247 stealthy in nature. The third type is created beyond the natural person's cognitive horizon. We remark
 248 that in the context of location privacy, these three types may involve both spatial and non-spatial data.
 249 We define spatial data as explicitly including location, i.e. information interpretable in an open and
 250 well-known system with the interpretation leading to a location. Examples are map or GPS coordinates,
 251 postal codes and street addresses.

252 One needs to agree that we are running an urgent agenda in addressing location privacy concerns,
 253 because in this information age, and in this infoconomy, data sources are rapidly expanding and
 254 third-party inference capabilities show substantial growth. First of all, public domain geospatial base
 255 layers have drastically increased in volume, quality and geographic coverage, and thus, information
 256 once known as the yellow pages, the road infrastructure, or the land administration registry are
 257 now often online [45–47]. Such sources provide the background against which location intelligence
 258 gathering and interpretation is made fruitful.

259 Next, we are making use of many more online applications (smartphone, lap- and desktop) now
 260 than we were, say, five years ago, and that trend is not tailing off yet. Younger generations consist of
 261 more intensive producers also. Moreover, the net of 'natural person satellites' around us is densifying
 262 rapidly. Such satellites are entities in our vicinity with which, with some regularity, we share the
 263 same location: family members, colleagues and friends come to mind first. The majority of them
 264 are already in the 'data play' anyway. Developments in smart cities and the Internet-of-Things, are
 265 bringing inanimate satellites to the scene such as our refrigerator, our bike, car and car keys, our home
 266 thermostat, and our garbage can. These will all have an online presence, and their (sometimes static)
 267 location and operational state may inform third-parties of our own location, presence *and* absence, as
 268 well. Data security on these devices at present is alarmingly low [48].

269 Central to location privacy are data sources that associate with the person or with the location.
 270 Looking first at personal data, we recognize the following types:

271 **unique identifier** This is a data element that is associated with just one entity of interest;
 272 that entity may be a data subject or something else. This is a wide class of identifiers;
 273 **key identifier** A data element that can be exploited with minimal effort to identify a
 274 (privacy-sensitive) data subject;
 275 **quasi-identifier** A data element that almost discloses the identity of a data subject, due
 276 to its semi-unique value, and that will allow full disclosure when combined with other
 277 quasi-identifiers;
 278 **private attribute** The remainder class of privacy-relevant but non-identifying data
 279 elements.

280 Key identifiers such as person names, phone numbers, email addresses, to some extent license
 281 plates and certain other device identity numbers, and also social media account names are all in this
 282 category. As the name suggests, quasi-identifiers do not immediately allow the identification of a
 283 data subject, and they require work. Key to quasi-identifiers is the aspect of data combinatorics. For
 284 instance, a combination of personal traits of athletes in a sports team may allow unique identification
 285 of some athlete. A private attribute represents information about a data subject that is exploitable in
 286 inferencing about her. Alice may be known for her fondness of chai latte.

287 A useful typology regarding location data is one that splits out on the basis of data structure
 288 complexity. We recognize the following types ordered by increasing complexity:

289 **location** This is the base case and it provides the whereabouts of an entity, a data subject,
 290 or that of a data subject's activity;
 291 **location with co-variates** A slight extension of the first case, in which the provided
 292 location is augmented with quasi-identifiers or private data elements. Such augmentation
 293 allows contextual inferencing about the function of the location to the data subjects;
 294 **timed location** This is a next step up the ladder, and associates with any provided location
 295 also some form of time stamp. The combination of these data elements allows inferences

296 towards what we can generically call trajectories. Presence and absence information also
 297 falls in this category;
 298 **timed location with co-variates** The final case, which allows inferencing over entity or
 299 data subject activity trails and life cycles.

300 This typology is important because it implies levels of data richness that determine the caloric
 301 value of the data that fuels possible inference processes over them. The types hint at an important
 302 distinction in what can be inferred, and they warrant different levels of awareness with the data subject
 303 who volunteers the data.

304 3.2. Inferencing over personal and location data

305 The exchange of personal and location data with third-parties commonly takes place in a service
 306 provision scenario. We use that term in a wide sense: local and national governments provide services
 307 to the citizens, software applications running on mobile and stationary devices (such as sensors and
 308 computing devices) provide services to the owners/holders. Adequate service delivery requires
 309 adequate data, and thus data processing, to serve appropriately. Alice understands that a driver's
 310 license renewal requires her to hand over identity and address details.

311 It is sensible to discriminate between controller and processor roles, as the GDPR does in its
 312 Article 4(7–8). The first determines purpose and means of processing, and the second 'just processes'.
 313 The key role of the controller is to define what is appropriate and adequate data in the context of some
 314 service delivery. In Figure 1, we sketch three common scenarios of service delivery. They differ in what
 315 happens to the personal and location data submitted by Alice in each scenario; they also indicate what
 316 Alice should know about data processing in each scenario.

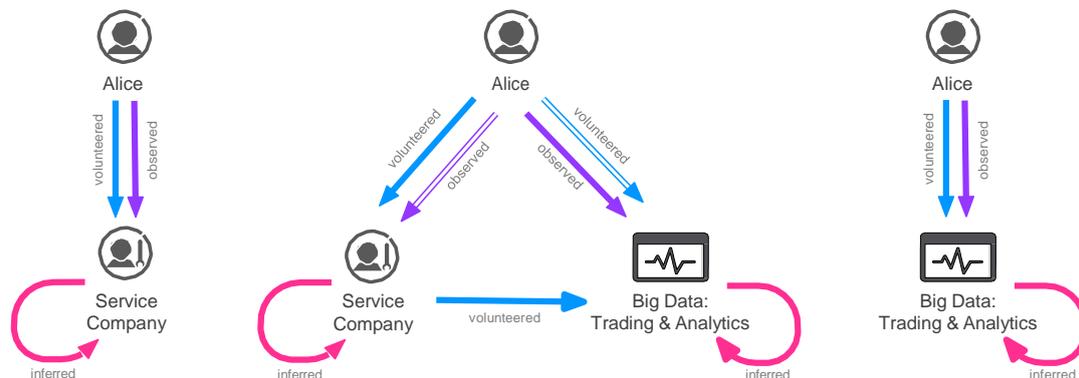


Figure 1. Three common scenarios in which Alice volunteers data with another party that provides some service: (a) An isolated service provision. (b) A service provision with 'unknown friend benefits'. (c) A 'full monty' service provision.

317 In the plain scenario of Figure 1(a), the service is provided in an isolated setting, and the controller
 318 and processor are often the same entity. Data is processed to serve properly and optimally. Alice is
 319 advised to do a few things, in the context of her concern over location privacy when she considers
 320 making use of the service. First, she will want to know whether a DPIA has been carried out over this
 321 service provision and provider. She will also want to know whether alternative services exist. While
 322 some services are monopolistic, such as the driver's license renewal, others, such as a smartphone
 323 navigation application, are not. Navigation applications do exist in this first scenario category. A third
 324 sanity check that Alice should carry out is to understand which personal data she is committing in the
 325 scenario, whether that data is needed for the functional purpose, and of the appropriate detail, and
 326 whether the service provider is building up user history and profile with the committed data. In this
 327 scenario, such is only legitimate when it aims to improve service quality levels to Alice. In principle,
 328 the user history, when purposeful, could be stored on Alice's device only.

329 The typical business model for the plain scenario depends on the controller type. When
330 governmental, usually a small, fixed fee applies that can be seen as transaction cost; often government
331 agencies have succeeded to economize their own processes and the service provision is seen as a
332 win-win. When commercial, different modes exist, with one being a license fee for using the service
333 that is either one-time and fixed, or that is subscription fee-based. Obviously, models also exist where
334 Alice is offered to accept receiving advertisements against a lower, possibly zero, fee.

335 Many of the awareness questions that Alice must pose under scenario (a) remain valid in
336 scenario (b). Much of what we wrote on scenario (a), applies here equally well. Yet, scenario (b)
337 is more complex and typically also has a form of data hand-over by the service provider to the big data
338 industry, consisting of data traders and data analytics companies. This is commonly part of the service
339 provider's business model, and so Alice should think twice when receiving great service at low cost.
340 She should also scrutinize the end-user agreement on what happens with her data after the service
341 provider has handed over her data to a third party. This scenario has become much more common
342 in recent years, due to the exponential growth of the big data markets. The situation is aggravated
343 by Alice's potential engagement with big data market companies directly, for instance in using social
344 media, or with providers of other services, which may also be enjoying a data hand-over agreement
345 with the big data industry. The laws of data combinatorics imply that Alice is likely deeply personally
346 profiled in such cases. She must remain on her guard for questionable business ethics.

347 The scenario of Figure 1(c) sketches a case where Alice has decided to use services directly from
348 a big data entity. She may not be less disconcerted, but at least understands with which entity she
349 is involved, and knows its reputation as a service provider and controller/processor. Were she less
350 informed, she might perceive the service as the only of its type, and so consider its use inevitable.
351 Luckily, Alice knows of alternative internet search engines, navigation apps, mail service providers,
352 and so forth, for those situations when she cares about her location privacy. It is worth observing that
353 where service providers operate in a competitive market, such as is the case in the telecom industry,
354 big data companies are often known to be careful in location privacy handling.

355 One can justifiably pose the question whether Alice can distinguish between the sketched
356 scenarios (a) to (c), let alone understand where her data ends up and which parties use it and to
357 what end. At present, her information position is sometimes dire indeed, and if she finds it hard
358 already, clearly others with a less strong background in information processing will be similarly
359 uncertain. At present, they will need to rely on emerging regulations that bring transparency and on
360 enforcing DPAs as well as consumer organisations that aim to keep parties honest, transparent and
361 informed.

362 4. Privacy at the cultural level

363 In Sections 2 and 3, we focused on Alice's privacy, her goals and capability to safeguard it, and
364 on the types of data that she is volunteering. We also discussed the possibly invading mechanisms of
365 observation and inference that service providers and third parties may be applying. In so doing, we
366 discussed privacy at the individual level, the first of Alan Westin's [16,18] three levels—individual,
367 cultural and political. We placed Alice at the center of a typology linking her with her social
368 environment. Alice safeguards the privacy of her research subjects as well as her own from attackers.
369 She is capable of making use of the opportunities her legal environment provides to lodge formal
370 complaints, when GDPR rules are broken. She is capable of overt and covert political resistance, when
371 all other options seem futile. In this section, we discuss privacy at the cultural level, starting from
372 a basic premise in social theory [49]: Alice's (privacy) preferences and commitments are shaped by
373 and shape the culture of her community and society. Her individual preferences and the culture—i.e.
374 the shared beliefs, attitudes, or way of life, or world view—of the community or society in which
375 she is socialized are deeply enmeshed and mutually reinforcing, with no way to decide which is the
376 dependent and which the independent variable.

377 4.1. Cultural theory: a typology of cultures

378 Social scientists have shown that the various ways in which individuals around the world express
 379 their preferences and commitments to human values, such as social justice, equality and privacy
 380 among others, are associated with four alternative ways of organizing human relations into cultures.
 381 Pepperday [50] offers a concise summary of these theories, each starting from different premises but
 382 arriving to similar ideal types of culture. The particular theory we use here goes back to the social
 383 anthropologist Mary Douglas and her fieldwork in the 1950s with the Lele people of the colonial
 384 Belgian Congo, now the Democratic Republic of Congo. Douglas' cultural theory distinguishes four
 385 ideal types of culture: egalitarianism, hierarchy, individualism and fatalism [51, (1978)]. For instance,
 386 egalitarians view social justice as just outcomes for all, hierarchists view justice as a just, rights-based
 387 process, while individualists view justice as just deserts—to each individual his due. Fatalists are
 388 resigned to a world of injustice they cannot control.

389 None of these four cultures is sustainable in pure form. Each culture needs more or less elements
 390 of the other three to become a viable hybrid. For instance, the individualist rule of 'notice and consent'
 391 fails to protect individuals' privacy. Neither do people read privacy notices, nor do they turn off the
 392 location tracking function of their cell phone. This leads to the "privacy paradox" – disclosing personal
 393 information, despite an expressed commitment to privacy [52]. Only a hierarchically reinforced form
 394 of consent can protect individualists from themselves. Article 7 of GDPR accomplishes this purpose. It

395 *"requires affirmative consent, which must be freely given, specific, informed, and unambiguous.*
 396 *Consent can't be assumed from inaction. Pre-ticked boxes aren't sufficient to constitute consent" [6].*

397 The mix of individualism and hierarchy is an improvement over the choice to opt-out, that infers
 398 consent from inaction.

399 The dimensions of Douglas' typology of cultures are 'grid' and 'group' [53]. Grid measures the
 400 extent to which role differentiation constrains the behavior of individuals: where roles are primarily
 401 ascribed, grid constraints are high; where roles are primarily a matter of choice, grid constraints are
 402 low. Group, by contrast, measures the extent to which an overriding commitment to a social unit
 403 constrains the thought and action of individuals.

Table 3. A typology of cultures [51, (1978)]

		Grid	
		Weak	Strong
Group	Strong	Egalitarianism	Hierarchy
	Weak	Individualism	Fatalism

404 4.2. A typology of privacy cultures

405 In the wake of the GDPR, all privacy cultures are readily observable. Fatalism, the most passive of
 406 the four, has been aggressively promoted by big tech companies for at least two decades. In 1999, Scott
 407 McNealy, the founder and CEO of Sun Microsystems, declared "you have zero privacy . . . get over
 408 it," a statement some in the privacy industry took as tantamount to a declaration of war [54]. In 2009,
 409 when asked whether users considered Google a 'trusted friend,' former Google CEO Eric Schmidt
 410 responded, "If you have something that you don't want anyone to know, maybe you shouldn't be doing it in
 411 the first place." In 2010, Mark Zuckerberg claimed that the privacy social norm "is just something that
 412 has evolved over time" as people tend to share more personal information with more people (Johnson
 413 2010). Privacy fatalism among data subjects or consumers enables big tech companies to naturalize the
 414 massive extraction of personal data – conveniently labelled 'data exhaust' – and to duly appropriate it,
 415 much like the former colonial powers, who appropriated terra nullius or 'no man's land', and exploited
 416 it without legal interference [55]. The tech companies' extractivism is premised on formal indifference
 417 to the populations that comprise both their data sources and their ultimate targets for behavioral

418 prediction, modification and monetization [9], as the Cambridge Analytica scandal [39] has shown.
 419 Besides fatalism, the three active privacy cultures – egalitarianism, individualism and hierarchy – are
 420 clearly discernible in the EU. They share a common core – a commitment to data distributism, a more
 421 just distribution of benefits from data extraction (Table 3). We discuss each in turn using their most
 422 recent empirical manifestations in France, Germany and the United Kingdom.

423 Individualists frame data privacy as a product that can be exchanged in the market place for a
 424 fair price (to each his due). An excellent example of this approach is the advocacy of the French think
 425 tank GenerationLibre [56] to extend the private property paradigm to personal data. GenerationLibre
 426 aspires to change the way the digital ecosystem works, by giving user-producers:

- 427 1. *“The possibility for e-citizens to negotiate and conclude contracts with the platforms (possibly*
 428 *via intermediaries) regarding the use of their personal data, so that they can decide for themselves*
 429 *which use they wish to make of them;*
- 430 2. *The ability to monetise these data (or not) according to the terms of the contract (which could*
 431 *include licensing, leasing, etc.);*
- 432 3. *The ability, conversely, to pay the price of the service provided by the platforms without giving*
 433 *away our data (the price of privacy?)” (p. 7).*

434 Hierarchists may be willing to surrender some of their privacy to a legal/rational authority (e.g.
 435 government) they trust, in exchange for another public good they value, e.g. security or economic
 436 growth. Andrea Nahles [57], the Chairperson of the German Social Democratic Party, framed the
 437 problem thus:

438 *“Empires like Google and Amazon cannot be beaten from below. No start-up can compete with their*
 439 *data power and cash. If you are lucky, one of the big Internet whales will swallow your company. If*
 440 *you are unlucky, your ideas will be copied.”*

441 Her solution is a Data-for-all law:

442 *“The dividends of the digital economy must benefit the whole society. An important step in this*
 443 *direction: we [the state] must set limits to the internet giants if they violate the principles of our social*
 444 *market economy. [...] A new data-for-all law could offer decisive leverage: As soon as an Internet*
 445 *Company achieves a market share above a fixed threshold for a certain time period, it will be required*
 446 *to share a representative, anonymized part of their data sets with the public. With this data other*
 447 *companies or start-ups can develop their own ideas and bring their own products to the market place.*
 448 *In this setting the data are not “owned” exclusively by e.g. Google, but belong to the general public.”*

449 Yet, as Morozov (2018) argues, Nahles’ agenda *“needs to overcome a great obstacle: citizens’ failing trust in*
 450 *the state as a vehicle of advancing their interests,”* especially in a country like Germany, with a long history
 451 of data privacy activism.

Table 4. A typology of privacy cultures

		Grid	
		<i>Weak</i>	<i>Strong</i>
Group	<i>Strong</i>	Data distributism (egalitarianism) Slogan: We produce and manage our personal data Privacy: Personal data as unalienable, constituting the self	Data distributism (hierarchy) Slogan: Data-for-all law Privacy: Personal data as a good that may be traded with a public good
	<i>Weak</i>	Data distributism (individualism) Slogan: My data are mine, but sell them for a fair price Privacy: Personal data as tradeable product.	Data extractivism (fatalism) Slogan: You have zero privacy, get over it Privacy: Zero

452 Instead Morozov [58] argues for an egalitarian approach to privacy as constitutive of who we are
453 and as radical citizen empowerment.

454 *"We should not balk at proposing ambitious political reforms to go along with their new data ownership*
455 *regime. These must openly acknowledge that the most meaningful scale at which a radical change in*
456 *democratic political culture can occur today is not the nation state, as some on the left and the right*
457 *are prone to believe, but, rather the city. The city is a symbol of outward-looking cosmopolitanism – a*
458 *potent answer to the homogeneity and insularity of the nation state. Today it is the only place where*
459 *the idea of exerting meaningful democratic control over one's life, however trivial the problem, is still*
460 *viable."*

461 Similarly, the Oxford-based Digital Rights to the City group, proposes a deeper meaning to the right
462 to information, a declaration that *"we will no longer let our information be produced and managed for us*
463 *[presumably by the state or corporations], we will produce and manage our information ourselves"* [59].

464 We saw that in the wake of the GDPR, the 'new global digital gold standard' [12], privacy cultures
465 are emerging that value privacy differently:

- 466 1. as a tradeable private good in return for another private good,
- 467 2. as something that constitutes who we are, and therefore is unalienable,
- 468 3. as something to be delegated to a trusted father-state and traded with a public good, and
- 469 4. as something that does not exist anymore and we should get over with.

470 As cultural theory predicts, eventually hybrid privacy cultures will prevail in specific European
471 communities and societies.

472 5. Personal data protection in the African Union: An outlook

473 Throughout this study, we argued that the fundamentally different privacy cultures of EU and
474 the US can be attributed to a clash of two core values, which has resulted in what some analysts
475 call a *"transatlantic data war"* [4, p. 117]. On the one hand, we have a European interest in personal
476 dignity, on the other hand, an American interest in freedom. *"On both sides of the Atlantic, these values*
477 *are founded on deeply felt sociopolitical ideals, whose histories reach back to the revolutionary era of the later*
478 *eighteenth century"* [14, p. 1219, emphasis added]). This motivated us to explore (location) privacy and
479 related safeguarding measures and strategies at two levels. First, at an individual level, with Alice
480 as protagonist, a fictitious (geo)computing scientist safeguarding the privacy of her research subjects
481 as well as her own from attackers; and second at a cultural level, by describing four data privacy
482 cultures – egalitarian, hierarchist, individualist and fatalist – all of recent vintage and emerging within
483 the EU in the wake of GDPR. We also noted that historically entrenched privacy cultures may inhibit
484 exporting the GDPR as a one-size-fits-all approach to other countries, e.g. to Member States of the
485 African Union (AU).

486 In this final section, we speculate about the challenges a group of African and European
487 collaborators, working on a joint research project, should tackle, when its activities are likely to
488 affect the (location) privacy of African 'data subjects' and/or their territorial assets, in a Member State
489 of the AU. We use the word 'speculate' purposefully to emphasize that any such exploration is meant
490 only to provoke deliberation, with other (geo)computing scientists who, like us, are routinely engaged
491 with African academia and government officials in long-term collaborative research. Three issues
492 merit attention in this regard: (1) African data protection legislation, (2) (data) privacy cultures within
493 the AU, (3) the social construction of personal data.

494 First, we note that the broad territorial scope of GDPR implies that its articles are applicable to
495 every non-EU organisation that processes the personal data or monitors the online activities of EU
496 citizens. Article 45(1) of GDPR regulates that the

497 *"transfer of personal data to a third country or an international organisation may take place where the*
498 *Commission has decided that the third country, a territory or one or more specified sectors within that*
499 *third country, or the international organisation in question ensures an adequate level of protection."*

500 If the European Commission decides that an African country is ensuring an adequate level of data
501 protection in processing data of Europeans, we may assume that the citizens of that country enjoy
502 the same level of data protection. About 40% of African countries have enacted data protection
503 legislation, which abides either to OECD standards (1st generation), or the EU DPD 1995 standards
504 (2nd generation), or even features a few GDPR elements (3rd generation), according to Greenleaf
505 and Cottier [60]. The latter refers to Mauritius, one of Africa's dynamic but small economies, which
506 updated its 2004 law in 2017, with a new Data Protection Act 2017 featuring some GDPR elements. In
507 June 2014, the African Union adopted the Convention on Cyber-security and Personal Data Protection,
508 known as the Malabo Convention [61], the first treaty outside the EU to regulate the protection of
509 personal data at a continental level [62]. The Convention aims to establish regional and national legal
510 frameworks for cyber-security, electronic transactions and personal data protection, but its actual
511 impact depends on ratifications, of which there were none by early 2016 [5]. In 2018, the AU created
512 data protection guidelines, broadly aligned with the GDPR, for its Member States, with contributions
513 from regional and global privacy experts, including industry privacy specialists, academics and civil
514 society groups [61].

515 What could we expect from those AU Member States without any data protection legislation in
516 place? As Makulilo [63], a Tanzanian privacy scholar, wryly observes,

517 *"the major legal systems in Africa namely common and civil law legal systems which are Western*
518 *in origin, create fertile grounds for adaptability of European law. While these systems were forcibly*
519 *imposed on Africa by European countries during colonial rule as part of the colonial superstructure*
520 *and an instrument of coercing Africans to participate in the colonial economy, they were inherited*
521 *by African countries on independence. [...] Thus, the attitude to view these systems as colonial has*
522 *diminished significantly as more customisation continues to take place. It is arguable that African*
523 *countries are no strangers to the adaptation of 'foreign law'." (p. 451).*

524 Obviously, a law in place does not necessarily imply its enforcement. Only Kenya and South Africa
525 have tested data protection rights in courts so far, an indicator of willingness to enforce the law [62].

526 Second, much has been made of Ubuntu, the famed African egalitarian culture, whose core
527 definition 'people are people through other people' leaves little room for personal privacy, and is at
528 odds with the strong western emphasis on individual rights. The South African information scientists
529 Olinger, Britza and Olivier [64] submit that Ubuntu, while still inspirational in many spheres of life
530 including African politics and business, has little purchase as a philosophical foundation of African data
531 protection legislation. Instead, they recommend alignment with European data protection legislation
532 on pragmatic grounds, especially for African countries for which the EU is a major data trading partner.
533 Arguing differently, Makulilo [62] makes a similar claim. He suggests that urbanization, the influence
534 of modern technologies and globalization have destroyed the social cohesion of communities, while
535 individualism is the order of the day in urban Africa. He recommends that data privacy regulations
536 should ensure the right to privacy of individual African citizens' is secured, much like it ensures the
537 right of EU citizens. Further, we take for granted that the other three privacy cultures, and associated
538 hybrids must exist as well (after all, Mary Douglas' cultural theory originated in her ethnographic
539 work in the Democratic Republic of Congo). However, African data privacy cultures and surveillance
540 mechanisms are undocumented, apart from few exceptions [65]. For instance, we do not know to what
541 extent, data privacy fatalism, aggressively promoted by big tech companies, has taken root in African
542 societies and communities, as it has in the EU. What data justice scholars (e.g. [25,26]) have documented
543 is the massive personal data extraction from African-based organisations, as well as the lobbying
544 of multinational corporations and their advocates for greater data emission, personalization and
545 centralization for expert analysis in advanced economies. It appears that corporations are becoming
546 the *de facto* custodians of African personal data, while local governments are hollowed-out, their
547 capacities depleted and local livelihoods are harmed.

548 Third, an individual's social environment influences what is deemed her personal (data). If
549 a society considers a given mode of personal behavior—e.g. political opinion, sexual orientation,

550 religious or philosophical beliefs, trade union membership—to be socially legitimate, only then is
 551 related data deemed personal. This social fact will affect efforts to harmonize data protection legislation
 552 across nations. Finally, as Alan Westin [16, p. 433] noted

553 “debates over privacy are never-ending, for they are tied to changes in the norms of society as to what
 554 kinds of personal conduct are regarded as beneficial, neutral, or harmful to the public good. In short,
 555 privacy is an arena of democratic politics. It involves the proper roles of government, the degree of
 556 privacy to afford sectors such as business, science, education, and the professions, and the role of
 557 privacy claims in struggles over rights, such as equality, due process, and consumerism.”

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