

# Measuring Conditional Anonymity — A Global Study


Pascal Berrang, **Paul Gerhart**, and Dominique Schröder



UNIVERSITY OF  
BIRMINGHAM



# Prologue: Modern Health App

- Your data is safe
- We do not store personalized data
- EU GDPR
- California Consumer Privacy Act
- Secure by Design
- HTTPS protocol encrypted with TLS (Transport Layer Security)

# Our Contribution



**Conditional  
Anonymity**



**The Dataset**



**Case Studies**



**Visual Anon**

# Starting Point: Matching Attacks

**Anonymized** dataset containing  
**confidential information**

Zip	Age	Sex	Confidential
15XX	70-75	F	...
12XX	25-30	M	...
95XX	65-70	F	...
11XX	15-20	M	...
12XX	45-50	F	...
⋮	⋮	⋮	⋮

**Unanonymized** dataset containing  
**no confidential information**

Identity	Zip	Age	Sex
Alice	1161	19	F
Bob	1234	27	M
Charly	4854	45	F
Dave	1277	28	M
Eve	9584	68	F
⋮	⋮	⋮	⋮

**Adversarial goal: match the databases**

# Open Question

How can we **access the unanonymized dataset?**

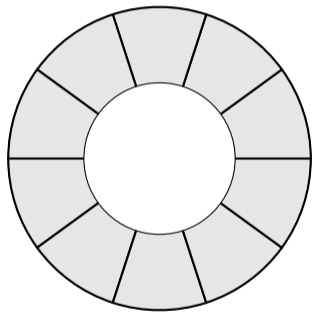
# Idea: Conditional Anonymity

- We gather publicly available statistical data.
- Using **population statistics**, we estimate the anonymity set size  $\psi(\vec{a})$ .
- We refine the set size by each **auxiliary information** we have.
- We define the conditional anonymity set for attributes  $\vec{a}$  and  $\vec{b}$  via

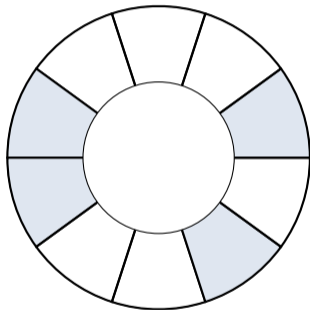
$$A_{\mathcal{P}}(\vec{a} | \vec{b}) = \psi(\vec{a}) \cdot \Pr[\vec{b} | \vec{a}].$$

# Conditional Anonymity Sets

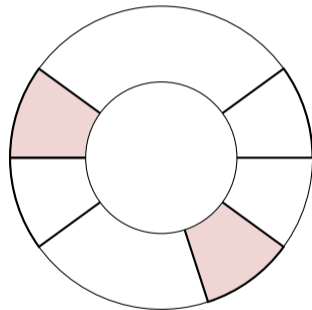
$$A_{\mathcal{P}}(\vec{a} | \vec{b} | \vec{c}) = \psi(\vec{a}) \cdot \Pr[\vec{b} | \vec{a}] \cdot \Pr[\vec{c} | \vec{a} \wedge \vec{b}]$$



$$\psi(\vec{a}) = 10$$



$$\Pr[\vec{b} | \vec{a}] = 0.4$$



$$\Pr[\vec{c} | \vec{a} \wedge \vec{b}] = 0.5$$

$$A_{\mathcal{P}}(\vec{a} | \vec{b} | \vec{c}) = 10 \cdot 0.4 \cdot 0.5 = 2$$

# Comparison To Prior Work

Matching Attacks

Identity	Zip	Age	Sex
Bob	1234	27	M
Dave	1277	28	M
⋮	⋮	⋮	⋮

Use an **exact database**

Counting all matching entries

Conditional Anonymity

Attribute	Value	Perc.	Set
Zip	11**	100 %	<b>2000</b>
Age	25 - 30	15%	<b>300</b>
Sex	M	50%	<b>150</b>

Use of **statistical data**

Compute an abstract group size



# Accuracy of CAS: AnonLand



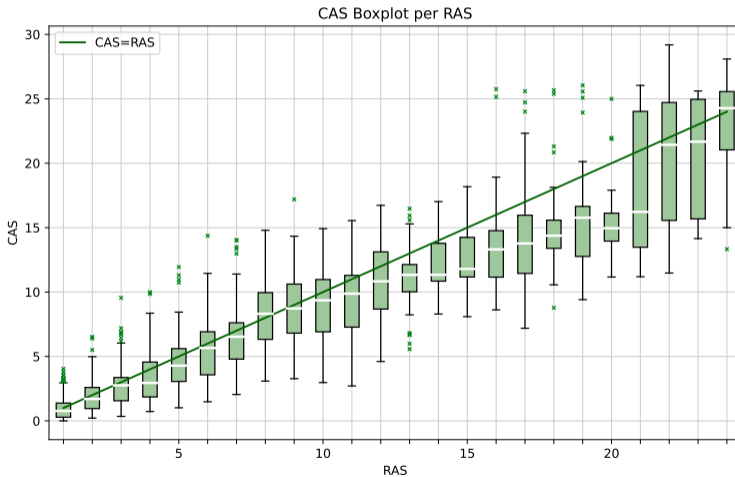
# Accuracy of CAS: AnonLand

AnonLand has 102.5 million citizens

- The age distribution is modeled using a trapezoidal shape
- Gender is distributed uniformly random
- Height and weight follow a normal distribution
  - 5 metropolises of five million inhabitants each
  - 25 cities of one million inhabitants each
  - 250 counties of 100,000 inhabitants each
  - 2,500 areas of 10,000 inhabitants each
  - 2,500 villages of one thousand inhabitants each



# Accuracy of CAS



# Our Contribution



Conditional  
Anonymity



The Dataset



Case Studies



Visual Anon

# Data Request



**Paul Gerhart**

22. November 2021 at 11:56

Consensus-data request for researching purposes

[Hide](#)

Bcc: census.customerservices@ons.gov.uk, Eurostat Helpdesk\_EN, User Information Services Stats SA, leosanni@nigerianstat.gov.ng, STATCAN.infostats-infostats.STATCAN@canada.ca, Atencion a Usuarios, ibge@ibge.gov.br, info, Stat, info@stats.gov.cn, ddu.rgi@nic.in, pbs@pbs.gov.pk, client.services@abs.gov.au, info@stats.govt.nz, nstac-info@nstac.go.jp, statistics@un.org, statistics@afdb.org, nfo@tuik.gov.tr, Dominique Schröder, Pascal Berrang

To whom it may concern,

My name is Paul Gerhart, and I am part of a privacy research team at the chair of applied cryptography of the Friedrich-Alexander-University Erlangen-Nürnberg in cooperation with the University of Birmingham.

My team and I are working on a web app to inform people about the anonymity set they are currently living in. That is the number of people who fit in the same data bucket created by several data points one may provide voluntarily without worries. With our app, we want to create awareness of how sensitive personal data is to help people protect their privacy.

Our work is based on the paper Pandemic Privacy by Berrang and Schröder, but we want to stress the insights to a worldwide dataset.

Therefore, we are interested in census data that gives insights into the population count by postcode separated by age groups and sex. Moreover, we are interested in the distribution of height and weight by age group and sex.

Based on this data alone, we cannot deanonymize people, but we can show how anonymity decreases by the publication of personal data that might seem irrelevant.

Hence we were hoping you could provide the desired data for us.

Best regards

Paul Gerhart

--

Paul Gerhart

[paul.gerhart@fau.de](mailto:paul.gerhart@fau.de)

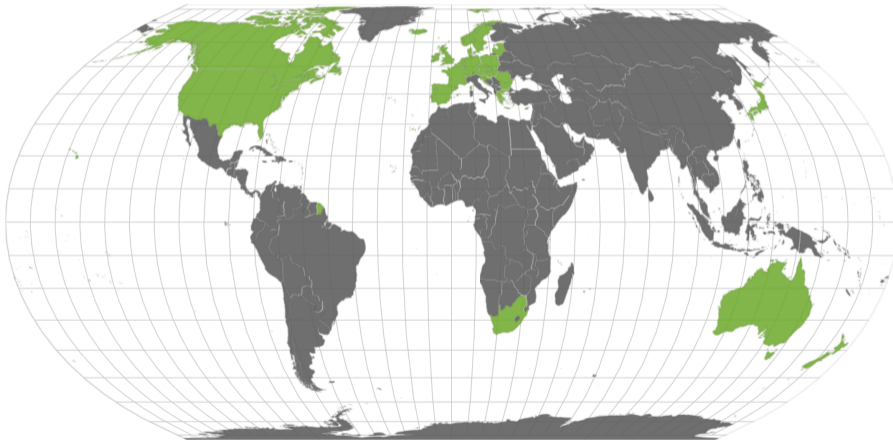
Lehrstuhl für Angewandte Kryptographie

Friedrich-Alexander-Universität Erlangen-Nürnberg



# Our Dataset

Currently, we can calculate anonymity sets for 1 084 230 346 people.



# Data Response I

S

**Stat**

Ответ на обращение

To: Paul Gerhart

VisualAnon 8. December 2021 at 13:33



Данные.xlsx



Подлинник.pdf



# Data Response II



МИНЭКОНОМРАЗВИТИЯ РОССИИ  
ФЕДЕРАЛЬНАЯ СЛУЖБА  
ГОСУДАРСТВЕННОЙ СТАТИСТИКИ  
(РОССТАТ)

Герхарт П.  
paul.gerhart@fau.de

Мясницкая ул., д. 39, стр. 1, г. Москва, 107450  
Тел.: (495) 607-49-02, факс: (495) 607-22-06  
http://www.gks.ru; e-mail: stat@gks.ru

*Dr. P. Gerhart* № 6116/07

на № \_\_\_\_\_ от \_\_\_\_\_

Уважаемый господин Герхарт!

В связи с Вашим обращением направляем имеющуюся официальную статистическую информацию о распространенности роста и веса в разбивке по возрастным группам (в возрасте 15 лет и более) и полу. Данные предоставлены по итогам Выборочного наблюдения состояния здоровья населения 2020 года, материалы и база микроданных которого размещены на официальном сайте Росстата (<https://rosstat.gov.ru/>): Статистика/ Переписи и обследования/ Федеральные статистические наблюдения по социально-демографическим проблемам/ Итоги выборочного наблюдения состояния здоровья населения.



# Our Contribution



**Conditional  
Anonymity**



**The Dataset**



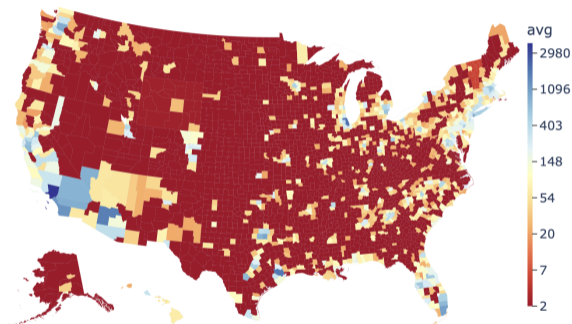
**Case Studies**



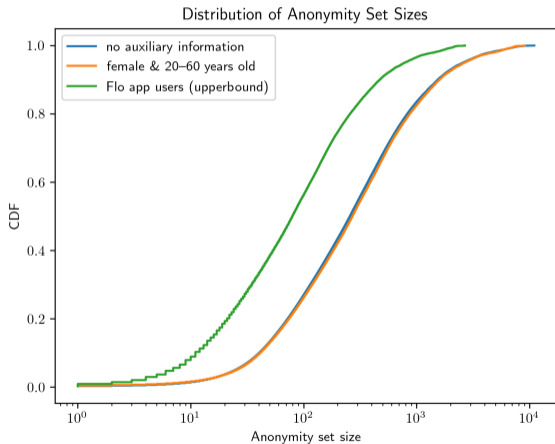
**Visual Anon**

# Case Study: USA

- Avg. CAS: 77
- Avg. CAS in red area: 2
- Avg. CAS below 5 in 97% of the counties



# Case Study: Flo app



- App users avg. CAS: 20
- Avg. CAS **below 5 in 97% of the US counties**

# Our Contribution



**Conditional  
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**The Dataset**



**Case Studies**



**Visual Anon**

# Visual Anon

Visual Anon Check

Country  
Australia ▾

District  
Abbotsford ▾

Sex  
Male ▾

Age  
65 - 69 years ▲

Height  
180 - 184 cm ▲

Weight  
85 - 89 kg ▲

- 26 177 413 people live in Australia
- 7 860 of them in Abbotsford
- 3 981 are male

# Visual Anon (Age)

Visual Anon Check

Country  
Australia

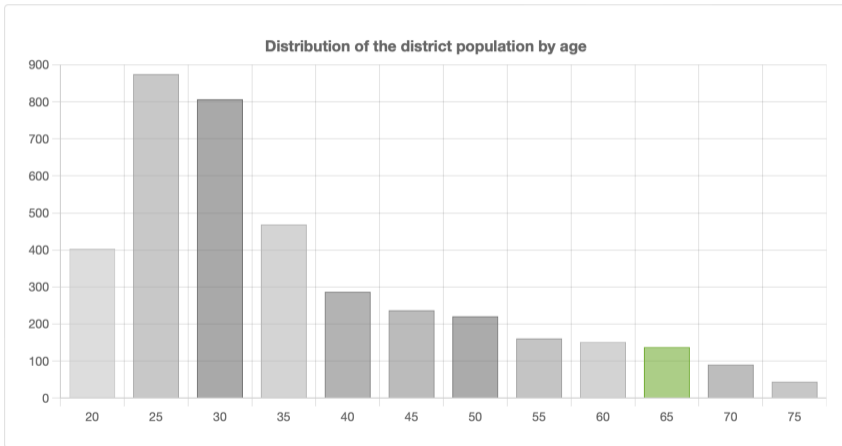
District  
Abbotsford

Sex  
Male

Age  
65 - 69 years

Height  
180 - 184 cm

Weight  
85 - 89 kg



# Visual Anon (Height)

Visual Anon Check

Country  
Australia

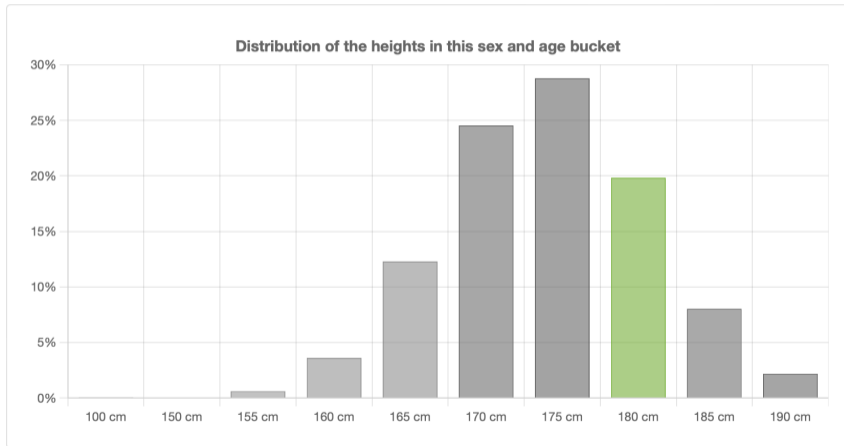
District  
Abbotsford

Sex  
Male

Age  
65 - 69 years

Height  
180 - 184 cm

Weight  
85 - 89 kg



# Visual Anon (Weight)

Visual Anon Check

Country  
Australia

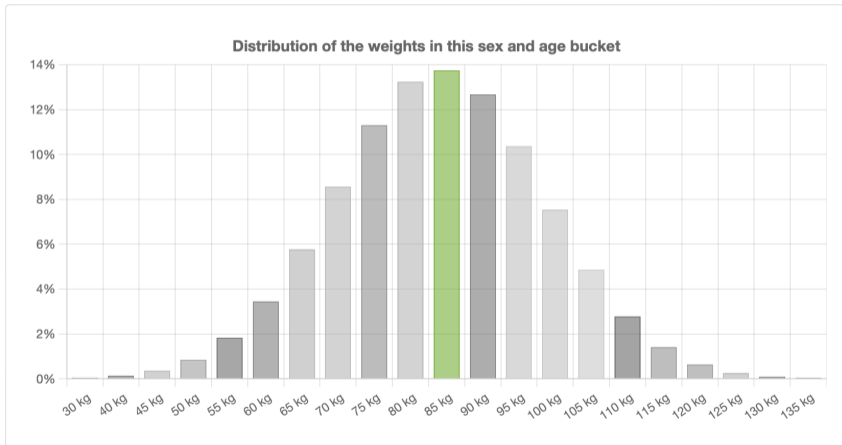
District  
Abbotsford

Sex  
Male

Age  
65 - 69 years

Height  
180 - 184 cm

Weight  
85 - 89 kg





# Visual Anon: Result

Providing this data you can be deanonymized up to **3** Persons.

26 177 413 People live in Australia

---

7 860 of them in Abbotsford

---

3 981 are male

138 are in the ['65'] years bucket

---

27 of them are ['180'] cm high

---

3 of them weight ['85'] kg

# Conclusion



**Conditional  
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<https://eprint.iacr.org/2024/998.pdf>



OurPaper



VisualAnon

# The Smallest CAS Size

Country	District	Minimal Set
Australia	Acton	1 male
	Alps - East	1 female
NZ	Chatham Islands Territory	3 males
	Chatham Islands Territory	3 females
Japan	Nakagusuku-son	19 males
	Nakagusuku-son	34 females
US	Bastrop County, Texas	82 males
	Riley County, Kansas	165 females

# Gender-Specific CAS Differences

Country	CAS (Male)	CAS (Female)	Male/Female
Sweden	13,020.80	12,983.48	1.00
Iceland	4,381.61	4,357.64	1.01
Norway	7,465.53	7,350.08	1.02
Hong Kong	10,551.02	2,672.81	<b>3.95</b>