

# Assessing deanonymisation risk in unstructured text

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Crucial issue for **privacy-preserving data publishing** (PPDP, see Fung et al. 2010)

Quantitative measures

- **re-identification risk** = #linked records / #records (Manzanares-Salor et al. 2024: 4044)
- **k-anonymity** = candidate set cannot be reduced to less than  $k$  entities
- flip side: **information loss** due to privacy models

Most research focuses on **statistical databases**

- full data set and background knowledge assumed to be known → can compute  $k$ -anonymity etc.
- approaches often based on perturbation of numerical data
- but not applicable to **unstructured text** (except artificial tasks such as Wikipedia pages of well-known 20<sup>th</sup>-century actors, cf. Manzanares-Salor et al. 2024: 4060 f.; Manzanares-Salor & Sánchez 2025)

Anonymisation of **unstructured text** poses entirely different challenges:

- may contain **direct identifiers** (personally identifying information = **PII**)
- contains wealth of further information that might contribute to deanonymisation = **quasi-identifiers**
- pseudo-identifiers are not formally recognisable (unlike attributes in a database, see Lison et al. 2021)
- background knowledge cannot easily be quantified → impossible to estimate  $k$ -anonymity etc.
- often about “ordinary people” with little information available online (our use case: **court verdicts**)
- information loss difficult to quantify and depends on application (e.g. *DeLorean* irrelevant for case)

On the late afternoon of **Wednesday, Oct 29<sup>th</sup>** **retired scientist** **Dr Emmett Brown** came driving down from **Beachy Head** in his **silver DeLorean** at high speed. Missing a turn, he crashed through the garden fence at **25 Baslow Rd** causing massive damage to ...

- **redaction** (critical spans deleted) → text difficult to read, maximal information loss
- **initials** (surprisingly common) → leaks information, PII spans become quasi-identifiers
- **randomised initials** → much better protection, but potential inconsistencies (e.g. random dates)
- **realistic surrogates** → natural text, suitable as LLM input, dates shifted to remain consistent (not pseudonymisation if mapping is discarded after anonymisation)

On the late afternoon of Wednesday, Nov 12<sup>th</sup>, retired scientist Prof John Cage came driving down from Capitol Hill in his silver BMW at high speed. Missing a turn, he crashed through the garden fence at 9 High St, causing massive damage to ...

## 1) PII text span **not detected**

- assumption: always leads to full reidentification of entity
- deanonymisation risk = recall of PII detection (more precisely: % of texts with  $\geq 1$  FN)

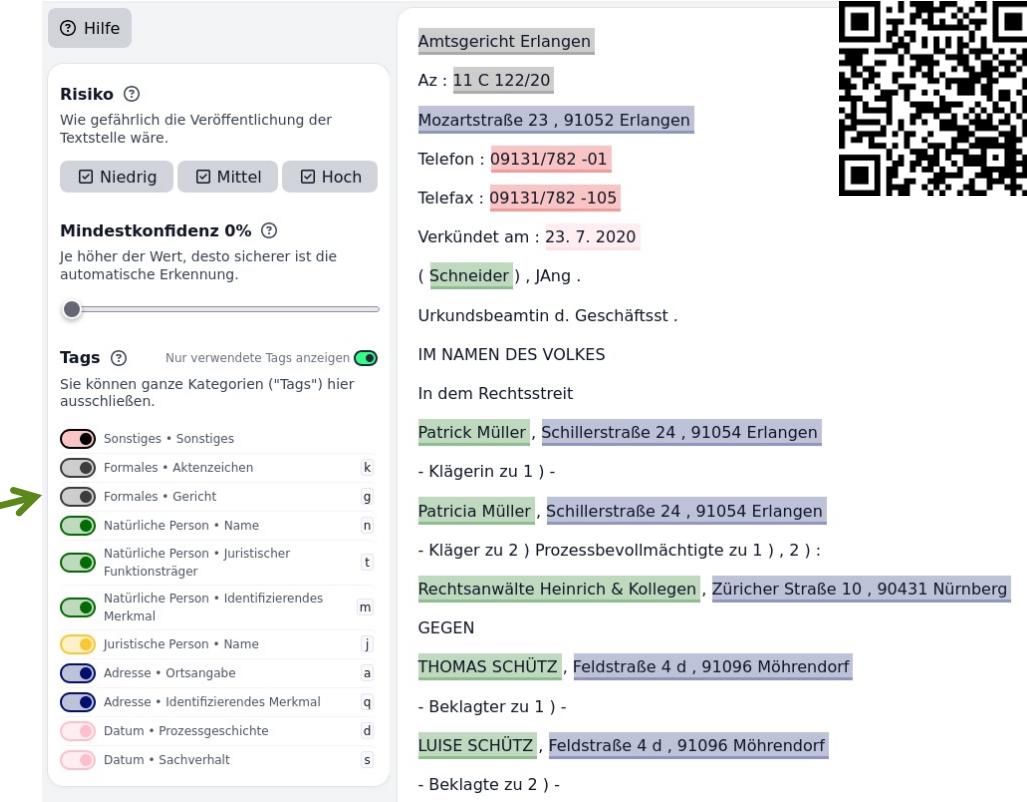
## 2) Quasi-identifiers not masked

- combination of multiple quasi-identifiers with background knowledge may enable reidentification
- difficult to quantify: empirical success rate of human adversaries using Web searches (or recently LLMs)
- controlled experiments are almost impossible

## 3) Masking techniques

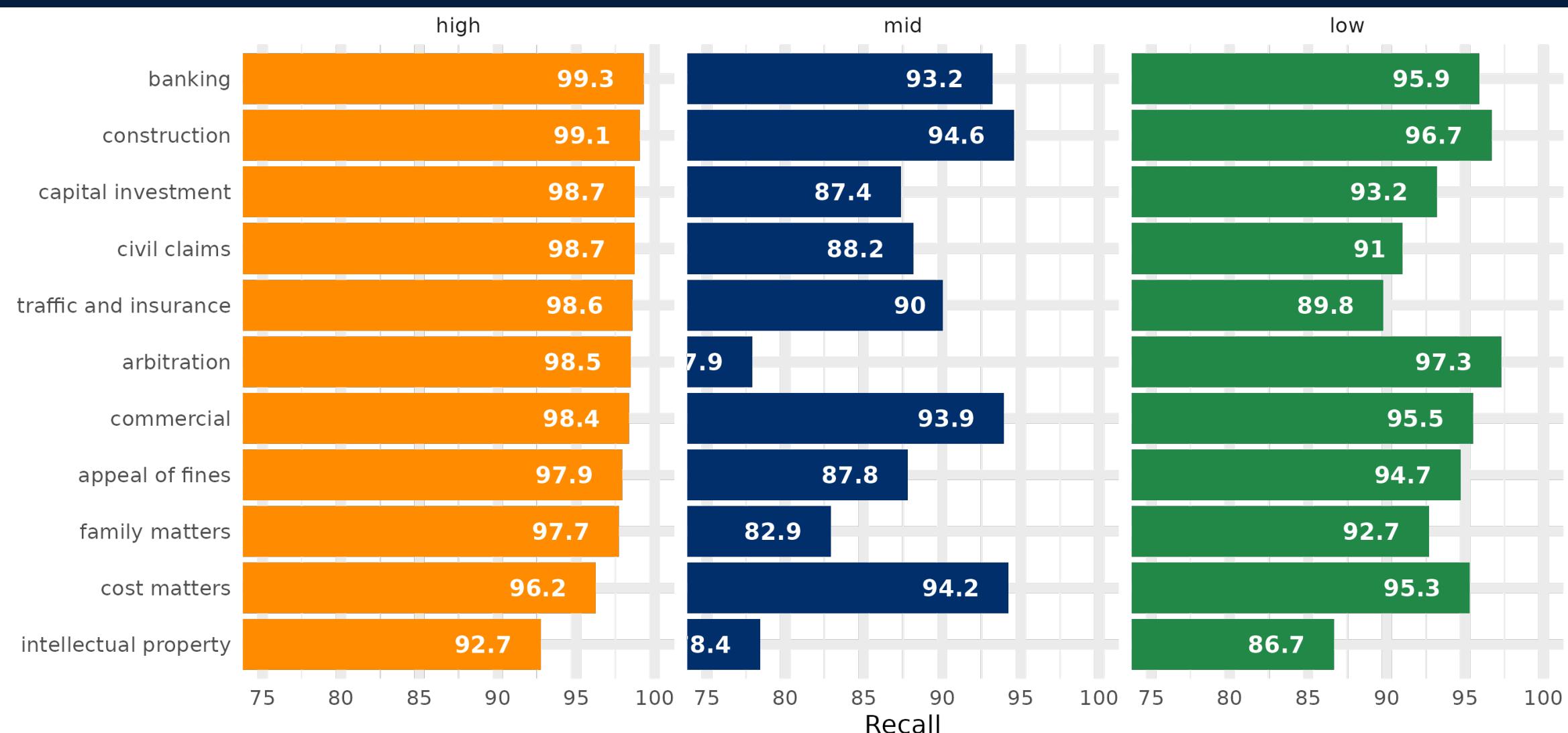
- simple techniques such as initials leak information, but also randomised initials (*the D&B shop*)
- realistic surrogates are considered safe, but challenging for quasi-identifiers

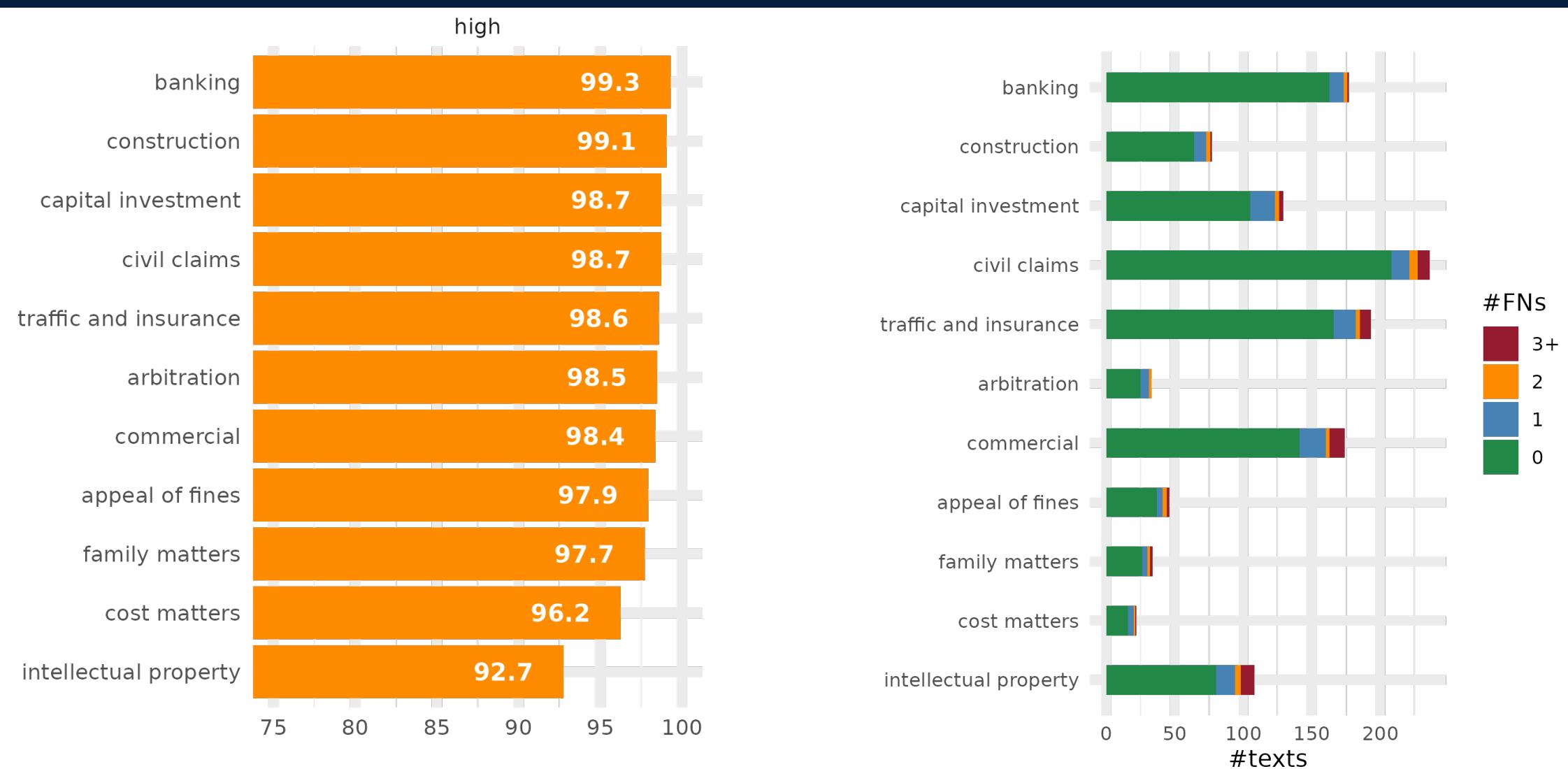
- **Goal:** fully automatic anonymisation → only way to publish as many as 1.5 million German court verdicts / year
- **Challenge:** extremely high recall (> 99%) for PII mandatory, deep text anonymisation also has to cover quasi-identifiers
- **Approach:** fine-tuning of pretrained LLMs for span identification and categorisation (multi-task)
- **Gold standard:** unusually high quality essential for evaluation and training (6 annotators / adjudicators for each text)
- **Result:** automatic anonymisation is **feasible**, but needs training or adaptation for each legal domain & court type
- **Team:** Bao Minh Doan Dang, Philipp Heinrich, Michael Keuchen, Mahdi Mantash, Daniel Odorfer, Melanie Rosa, Pei-Yu Shen, Naveed Unjum, Julian Werner, Leonardo Zilio + student assistants as annotators



The screenshot shows a web-based tool for anonymizing court documents. At the top, it displays the court's name (Amtsgericht Erlangen), address (Mozartstraße 23, 91052 Erlangen), and contact information (Telefon: 09131/782-01, Telefax: 09131/782-105). It also shows the date the document was published (23. 7. 2020) and the names of the parties involved: Schneider and Jang. The interface includes a 'Risiko' (Risk) section with a slider and a 'Mindestkonfidenz' (Minimum Confidence) section with a slider. On the left, there is a 'Tags' (Tags) section where users can filter results by category. A green arrow points from the 'Result' section of the list to the 'Tags' section, highlighting the tool's functionality for handling legal domain-specific data.

Domain (AG)	Precision	Recall	Recall PII
tenancy law	97.04%	96.05%	<b>98.90%</b>
traffic law	97.41%	97.38%	<b>99.11%</b>





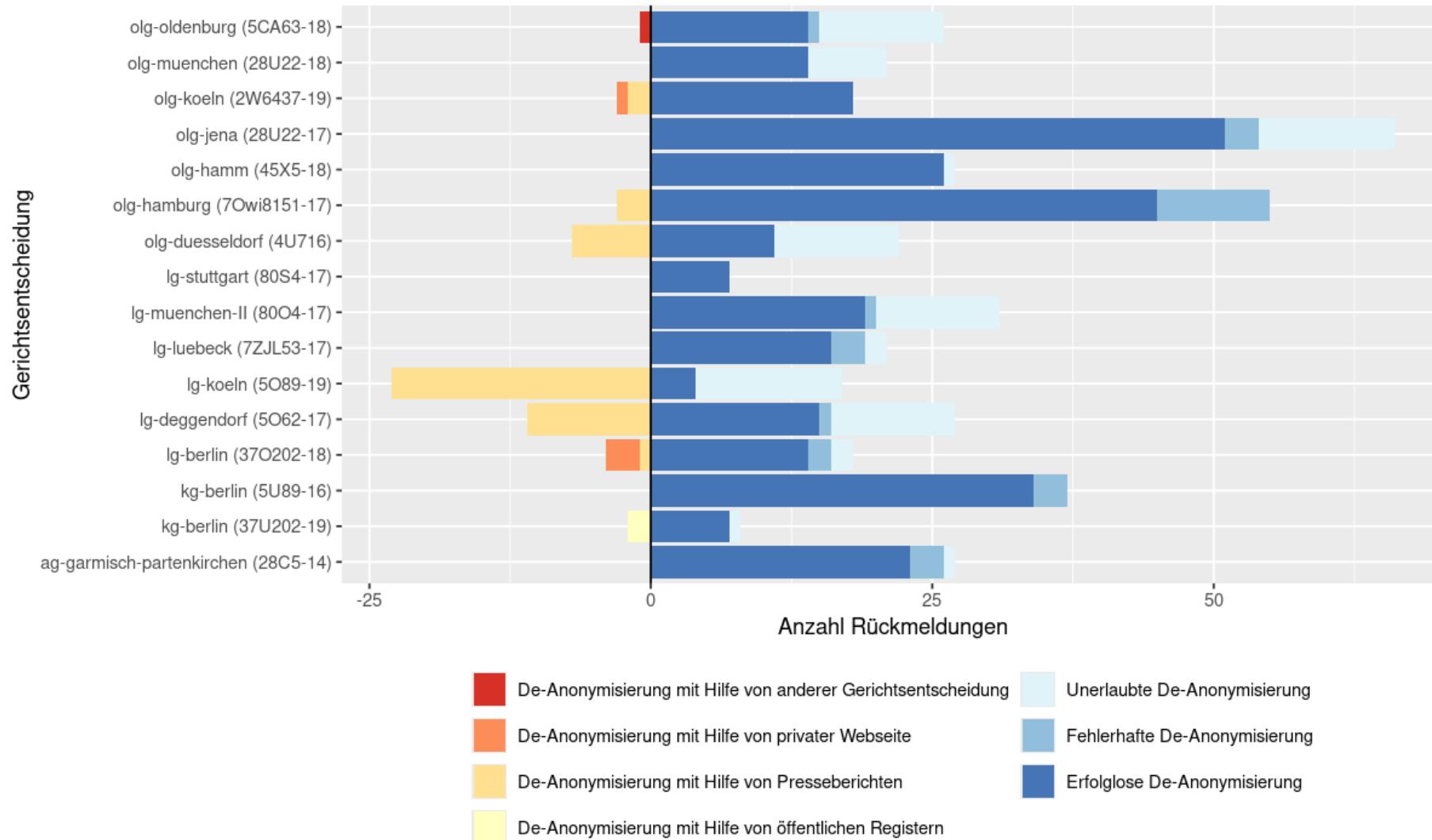
Experiment: **empirical deanonymisation** by human adversaries

- 10 verdicts from district court (AG) attacked by 6 human adversaries (time limit: 35 minutes / verdict)
- verdicts from AG gold standard chosen to contain large number of quasi-identifiers with elevated risk

Results:

- no successful reidentification of a natural person
- many spurious deanonymisation results (existing family name, existing street name or address, *I+D-Versicherung* → *R+V-Versicherung*, press reports of similar accident exactly one month off)
- success: format of case reference → **insurance company**
- success: index of rents (average cost, area type "red") → **specific city**

- 17 published verdicts of more prominent cases at regional courts
- known to offer attack vectors (previous work)
- deep anonymisation via LeAK/Anger guidelines with realistic surrogates (manual annotation)
- small number of texts with successful attacks
- mostly due to press coverage of the case



# Deanonymisation risk due to masking technique

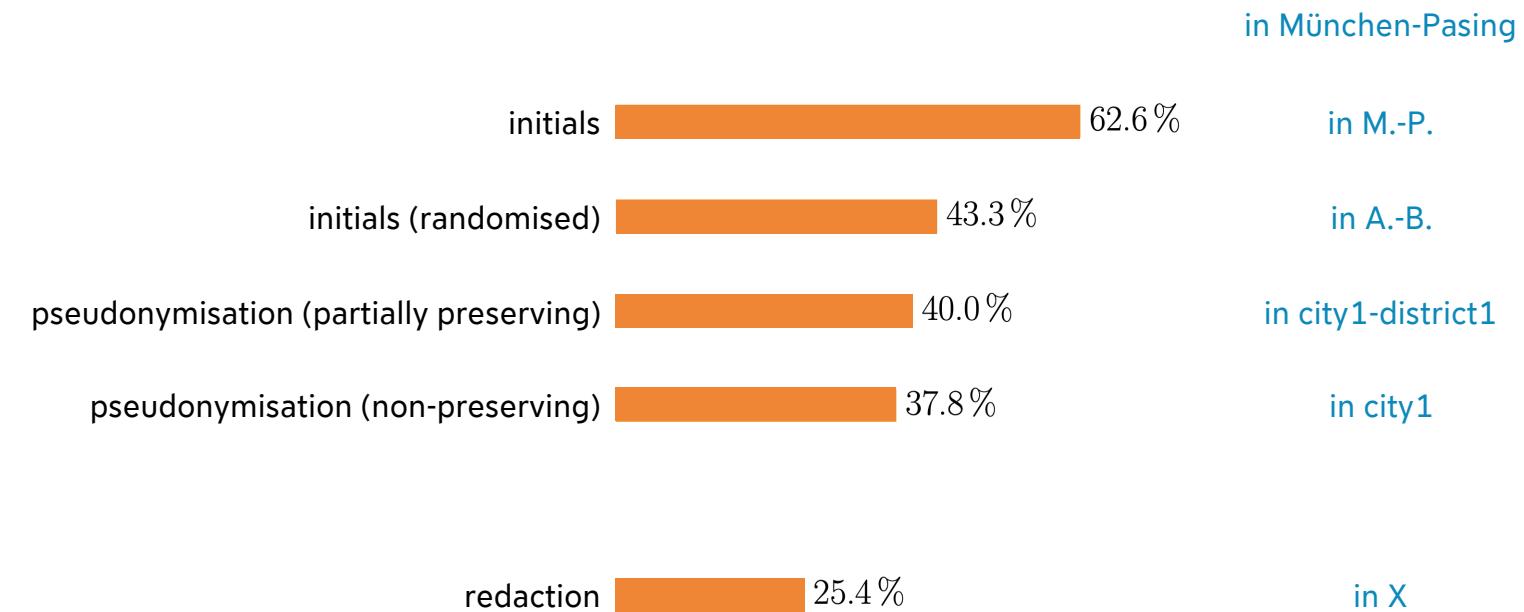
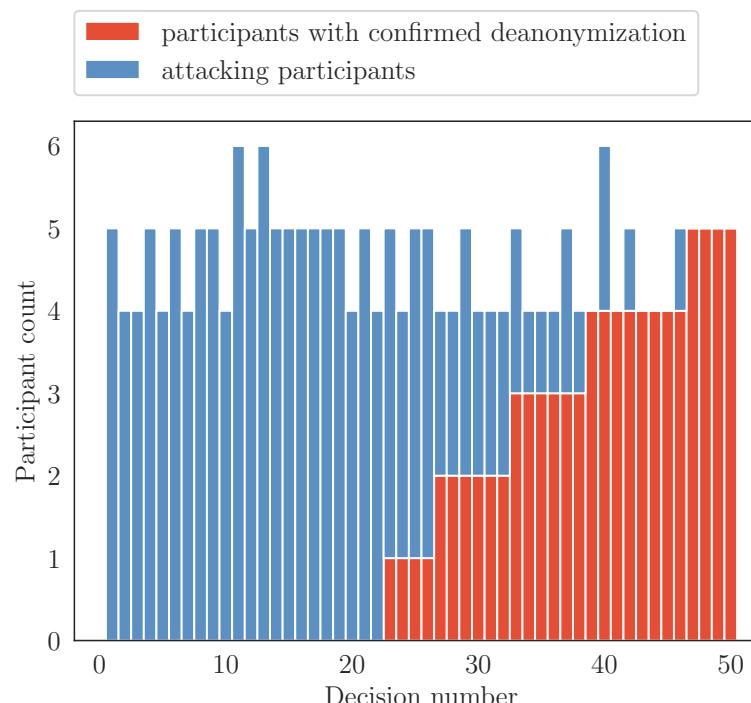
(Deuber et al. 2023)



FAU

## Experiment: **empirical deanonymisation** by human adversaries

- 50 published verdicts using different masking techniques (identified with heuristics)
- strong publication bias towards higher courts (due to random selection of eligible verdicts)
- human adversaries: 54 law students in online experiment | time limit: 35 minutes per verdict



# Thanks for listening!



<https://www.linguistik.phil.fau.de/projects/leak-anger/>

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