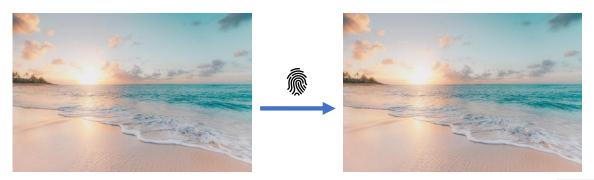




Use case: medical data

- Medical data sensitive and a valuable asset to the hospital
 - The owner wants to share data with researchers
 - The owner wants to have a claim of the ownership and trace the source of an unauthorised publishing
- Idea: hide a piece of information in the data before the distribution

Fingerprinting



| Age | BloodPress. | Diabetes | Age | BloodPress. |
|-----|-------------|----------|-----|-------------|
| 32 | 64 | 1 | 33 | 64 |
| 31 | 66 | 0 | 31 | 68 |
| 50 | 72 | 1 | 50 | 72 |
| 48 | 70 | 0 | 47 | 70 |

Diabetes

0

Fingerprinting

FINGERPRINT - string of bits containing information about the owner and the recipient of the specific data copy

FINGERPRINTING - an information hiding technique that embeds the fingerprint into the data



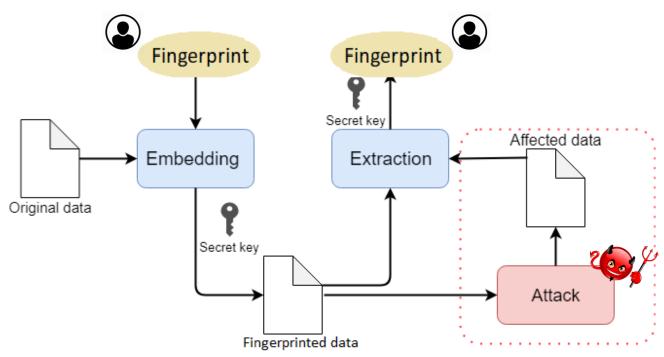
Fingerprinting

Requirements of the fingerprint:

- 1) recognizable by the owner
- not detectable and consequently removable by the recipients
- 3) robust to attacks
- does not change the utility of the data too much



Fingerprinting Schemes: workflow



Fingerprinting Schemes: Numerical data

AK (Agrawal and Kiernan) Scheme:

 Pseudorandom choice of a row, a column and a least significant bit of a value to be marked

| Age | BloodPress. | Diabetes |
|-----|-------------|----------|
| 33 | 64 | 1 |
| 31 | 66 | 0 |
| 50 | 71 | 1 |
| 48 | 70 | 0 |

Block Scheme:

- The data is first divided into blocks
- Pseudorandom choice of the value to be marked within every block

| Age | BloodPress. | Diabetes | | | | |
|-----|-------------|----------|--|--|--|--|
| 32 | 64 | 1 | | | | |
| 33 | 66 | 0 | | | | |
| 49 | 72 | 1 | | | | |
| 48 | 70 | 0 | | | | |

Two-level Scheme:

- 1st layer: Pseudorandomly selects the values to be marked; this pattern identifies the owner
- 2nd layer: Pseudorandomly selects the values to be marked; this pattern identifies the recipient

| Age | BloodPress. | Diabetes |
|-----|-------------|----------|
| 31 | 64 | 1 |
| 31 | 69 | 0 |
| 50 | 72 | 1 |
| 45 | 70 | 0 |

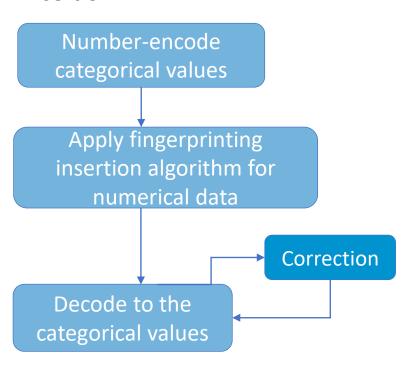
Numerical vs. Categorical Data

• Decimal: 32.3 → 32.7 ⓒ • Integer: 32 → 34 ⓒ

• Non-numerical: France ———— Germany ?

Fingerprinting categorical data: Approach #1

Insertion



- London, Paris, Vienna
- Encoding: 0,1,2 (00,01,**10**)
- Fingerprinting: 10 -> 11
- Decoding: 11 -> ??

- Correction: 11 mod 3 = 00
- Decoding: 00 -> London

Fingerprinting categorical data: Approach #2

Pseudorandomly choose a value to fingerprint - target Find a neighbourhood based on all but the target value Calculate the frequencies of target attribute values Choose randomly a mark value (weighted by the friequencies)

- Addressing the problem of semantic relations between categorical attributes that can be disturbed by fingerprinting
- gender: male, pregnant: yes
- The fingerprinted value will be a value that is likely to occur in a combination with other values from the row

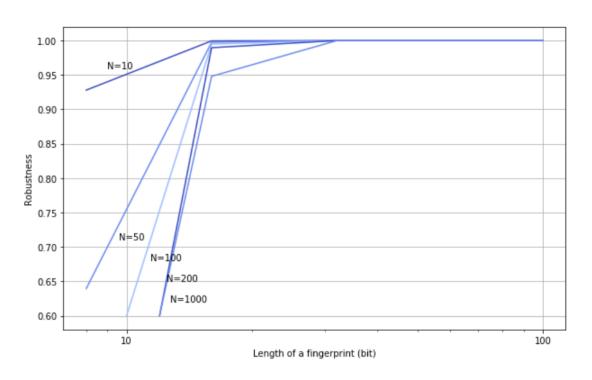
Robustness Evaluation: Attacks

- Malicious operations on the fingerprinted dataset with the goal of
 - disabling the extraction of the correct fingerprint or
 - disabling association of a fingerprint with the correct recipient
- Subset Attack
- Superset Attack
- Bit-flipping Attack
- Additive Attack

 Misdiagnosis False Hit Rate: measures the likelihood of extracting a valid fingerprint from non-fingerprinted data

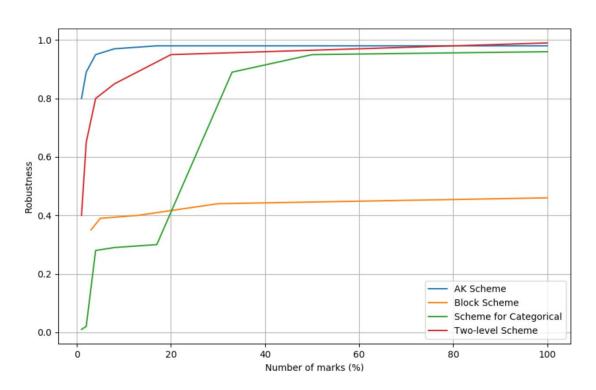
Robustness Evaluation

Misdiagnosis false hit



Robustness Evaluation

Subset Attack



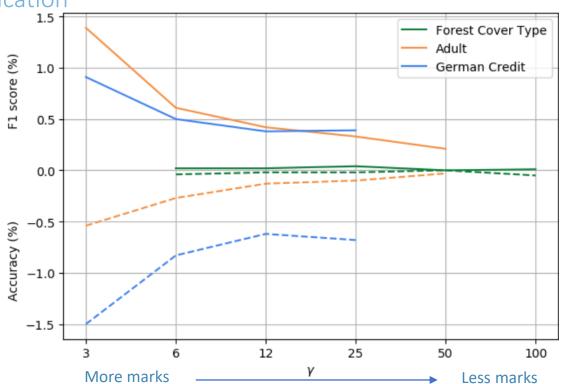
Utility Evaluation

Mean and Variance

| | | Less marks More marks | | | | | | | | | |
|----------------|----------|-----------------------|-----|----|-----|----|-----|----|------|---------|----------------------|
| | β | | 30 | : | 25 | | 15 | | 10 | | |
| | ξ | 4 | 8 | 4 | 8 | 4 | 8 | 4 | (8) | | Number of |
| Attribute | Variance | | | | | | | | | | Least Significant |
| Elevation | 78391 | 0 | +13 | +1 | +15 | +1 | +48 | +1 | +178 | | Bits being |
| Aspect | 12525 | 0 | +7 | 0 | +12 | 0 | +35 | 0 | +127 | | marked |
| Slope | 56 | 0 | +12 | 0 | +18 | 0 | +48 | 0 | 0 | | |
| HD-Hydrology | 45177 | 0 | +6 | +1 | +4 | +1 | +13 | +2 | 0 | | |
| VD-Hydrology | 3398 | 0 | +10 | 0 | +15 | 0 | +38 | 0 | +87 | | |
| HD-Roadways | 2431276 | 0 | +3 | 0 | +3 | 0 | +44 | -2 | 0 | | |
| Hillshade-9am | 717 | 0 | +11 | 0 | +15 | 0 | +41 | 0 | +8 | | |
| Hillshade-noon | 391 | 0 | +11 | 0 | +16 | 0 | +45 | 0 | +200 | | |
| Hillshade-3pm | 1465 | 0 | 0 | 0 | +13 | 0 | +35 | 0 | +160 | | |
| HD-Fire-Points | 1753493 | 0 | 0 | 0 | -4 | 0 | +54 | 0 | +68 | | |

Utility Evaluation

ML Classification



Conclusion and Future Work

Number of marks

| \uparrow | ω | ξ | L |
|------------------------|----------|----------|----------|
| Misdiagnosis false hit | ↑ | | ↑ |
| Subset Attack | ↑ | | + |
| Bit-flipping Attack | ↑ | 1 | + |
| Additive Attack | ↑ | \ | |
| Utility | + | | |

 Trade-off: robustness of the scheme and the utility of fingerprinted data

Future work

- Fingerprinting scheme for categorical data
 - Further analysis on robustnes of the neighbourhood-search approach
 - Blind scheme for fingerprinting relational data with categorical values

