

Introduction to Data Management Practical Aspects

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Practical Aspects

HW5 is due on Friday

- HW6 has two parts:
 - Part 1 due 5/17. No late days (for quick feedback)
 - Part 2 due 5/24. Much more work than part 1

Data Privacy Laws

Some data is protected by law:

- HIPPA
- GDPR
- FERPA

Health Information Portability and Accountability Act

- Mandatory for healthcare and health insurance institutions
- Privacy Rule to protect Protected Health Information
- Security Rule to ensure administrative, physical, and technical safeguards

GDPR

General Data Protection Regulation (GDPR)

- European Union
- Corporate disclosure of what user data is stored
- Only recently implemented (a few years ago)

FERPA

Family Education Rights and Privacy Act

- Mandatory for education institutions
 - Requires written consent to disclose academic info
 - Allows the release of directory information
- Allows institutions to disclose "directory information" without consent (institution policies can be stronger)
 - Name
 - Email
 - Photographs
 - Phone Number

Privacy Leaks via Linking

Anonymity

- Common practice for making a dataset private: remove Personal Identifiable Information (PII)
- But by linking data from distinct datasets one can reveal private information
- In her PhD thesis* (2001) Latanya Sweeney described a famous example

^{*} https://dspace.mit.edu/handle/1721.1/8589

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- 6 people had same **dob**
- 3 had also sex='M'

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- Sweeney paid \$20 and bought voter registration list for Cambridge, MA
- William Weld** lived in Cambridge: in VOTER
- 6 people had same dob
- 3 had also sex='M'
- Weld only one in that zip

GIC(zip, dob, sex,
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 diagnosis, procedure,...)

VOTER (**name**, party, ..., zip, dob, sex)

Sweeney learned Weld's medical records !

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The best common practice is still to remove PII

Law specifies which attributes are considered PII

Privacy Leaks via Aggregates

Implicit Disclosure

FERPA says:

- These might be public*
 - Name
 - Email
 - Photographs
 - Phone Number
- Grades are private;
- Grade averages from larger groups are OK

* Each university may impose further restrictions

Student(sid, name, email) Takes(sid, cid, grade) Course(cid, ...)

Student(sid, name, email) Takes(sid, cid, grade) Course(cid, ...)

Alice's grade in cse414:

SELECT T.grade
FROM Students S, Takes T
WHERE S.sid = T.cid
and T.cid = `cse414'
and S.name = `Alice'

No

Student(sid, name, email) Takes(sid, cid, grade) Course(cid, ...)

Alice's grade in cse414:

SELECT T.grade	9
FROM Students	S, Takes T
WHERE S.sid =	T.cid
and T.cid =	'cse414'
and S .name =	'Alice'

No

Student(sid, name, email) Takes(sid, cid, grade) Course(cid, ...)

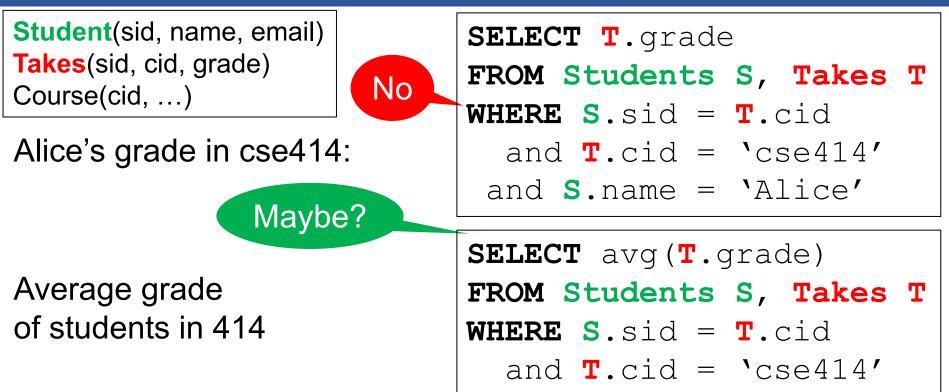
Alice's grade in cse414:

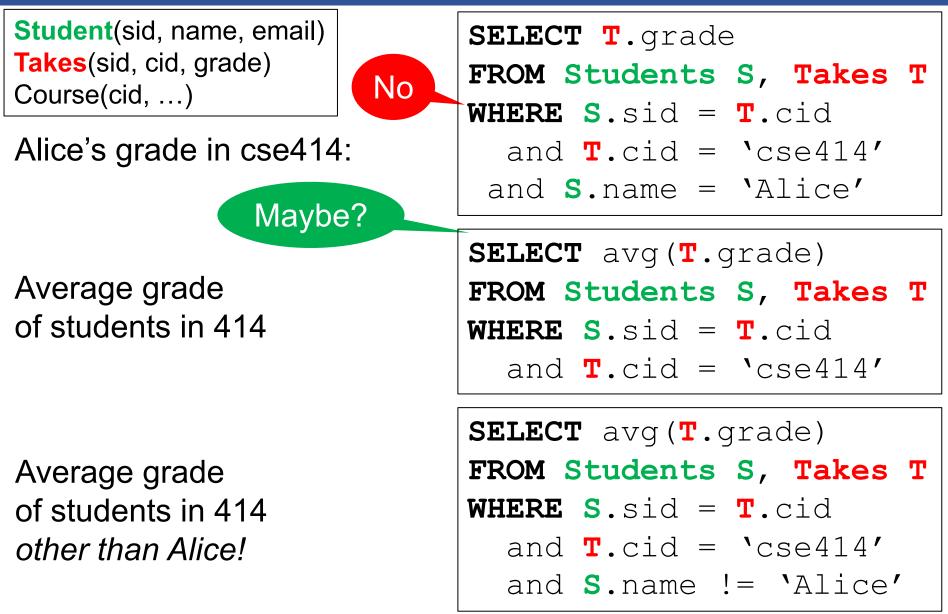
Average grade of students in 414 SELECT T.grade
FROM Students S, Takes T
WHERE S.sid = T.cid
and T.cid = `cse414'
and S.name = `Alice'
SELECT avg(T.grade)

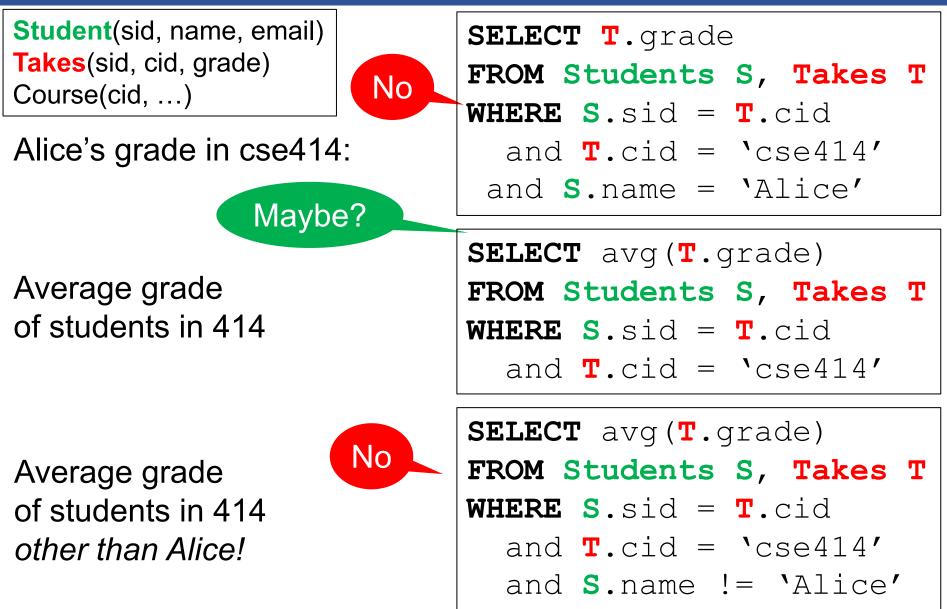
FROM Students S, Takes T

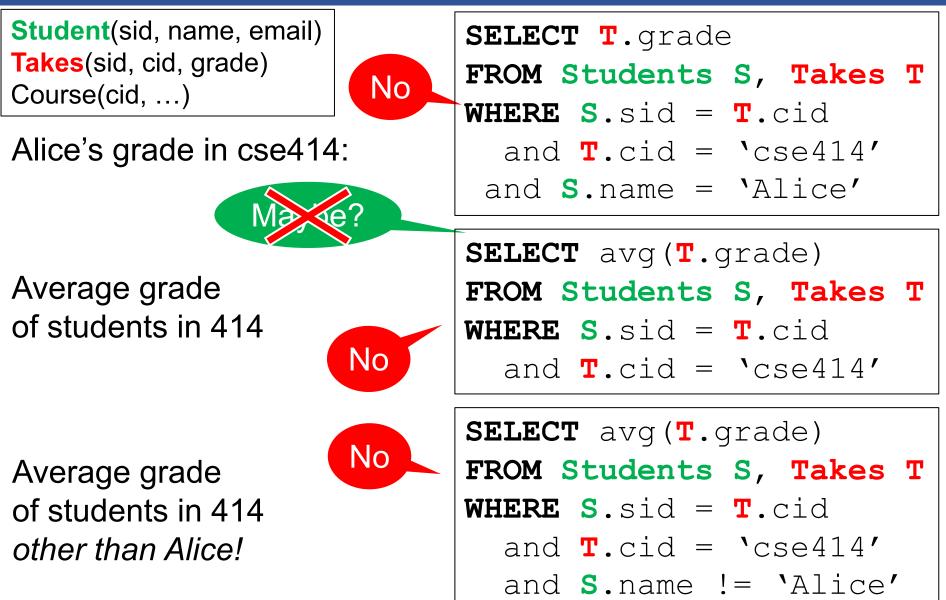
WHERE S.sid = T.cid

and T.cid = 'cse414'







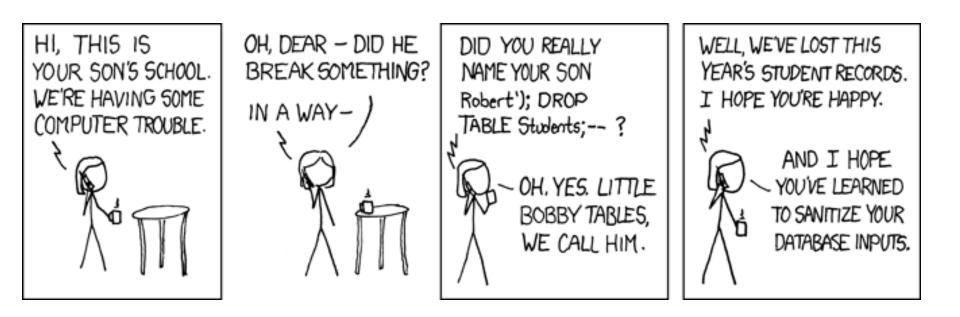


Make sure you understand how the privacy leak happened. Example:

- Sum of all grades = S
- Alice's grade = A
- 100 students in classs
- Avg grade in class:
- Avg grade w/o Alice:
- Solve for A:

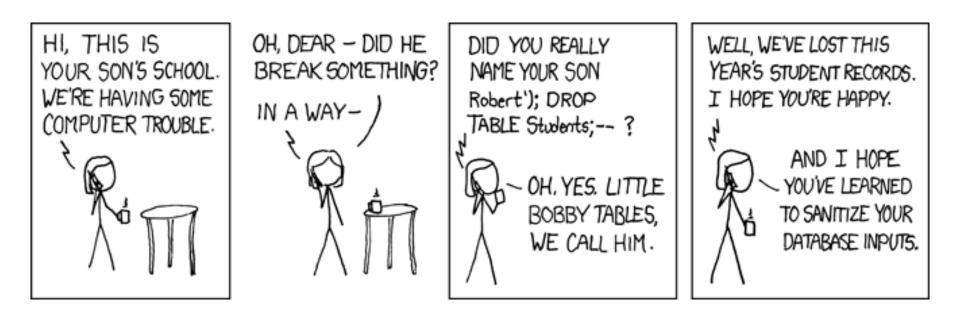
S/100 = 3.49 (S-A)/99 = 3.5

- Bucketize data and release only information on large groups
- Add noise: differential privacy



- In the application, a SQL query is a string
- Part of that string is input by the user
- A malicious user can enter a string that changes the SQL query

Demo

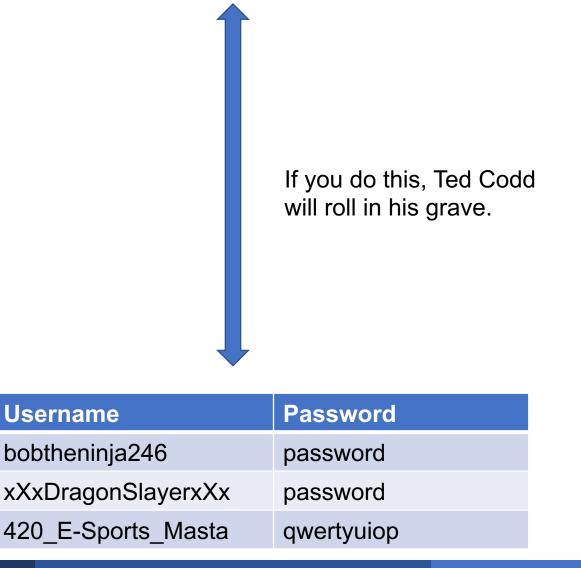


Considered a "solved" problem

- Parameterize queries using '?'
- Use 'prepared' statements

- Passwords are special
 - High potential for additional security compromises
 - Only operation that should be done is equality comparison

(bobtheninja246, password)

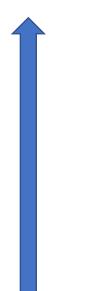


- Quick overview of hashing
 - Hash(input) \rightarrow hash value
 - Hash function takes input and generates "scrambled" output, that is always equal for the same input
 - Hashing is <u>deterministic</u>
 - Ideally hashing is <u>noninvertible</u>
 - Secure hash functions make it impossible to derive the input value from the hash value
 - Ideally hash values are uniformly spread out
 - Useful for hash tables!

Hash it!

(bobtheninja246, hash(password))

(bobtheninja246, FCgJFl9ryz)



Username	Hash
bobtheninja246	FCgJFl9ryz
xXxDragonSlayerxXx	FCgJFl9ryz
420_E-Sports_Masta	p8mel6usIF

Hash it!

(bobtheninja246, hash(password))

(bobtheninja246, FCgJFl9ryz)



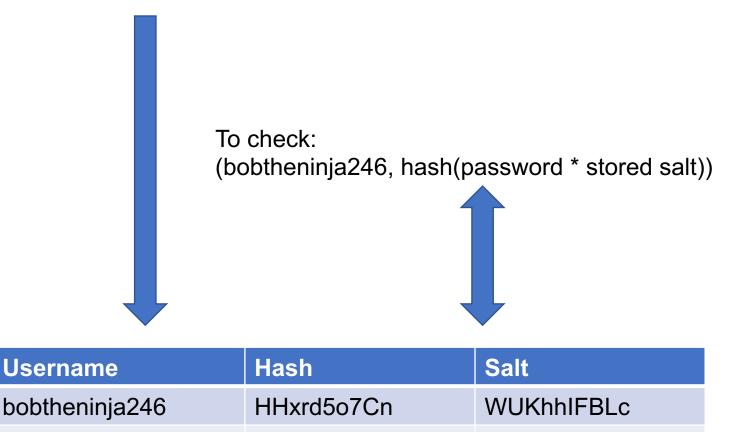
Issues/pitfalls:

- Hashing functions have precomputed "rainbow tables"
- Patterns can occur for the same passwords

Username	Hash
bobtheninja246	FCgJFl9ryz
xXxDragonSlayerxXx	FCgJFl9ryz
420_E-Sports_Masta	p8mel6usIF

Salt it and hash it!

(bobtheninja246, hash(password * random salt), random salt)



- These are just the fundamentals: companies outsource password management because it can get very complicated.
- In HW6 you are asked to do simple password management