Data Quality

Eugene Wu

Data Quality

Prevention

- Normalization prevents inconsistencies
- Constraints prevent invalid values

Detection

- Incorrect/anomalous values
- Deduplication

Steps for a New Application

Requirements

what are you going to build?

Conceptual Database Design

pen-and-pencil description

Logical Design

formal database schema

Schema Refinement:

fix potential problems, normalization

Normalization

Physical Database Design

use sample of queries to optimize for speed/storage

App/Security Design

prevent security problems

A Relational Model of Data for Large Shared Data Banks

E. F. Codd IBM Research Laboratory, San Jose, California

Future users of large data banks must be protected from having to know how the data is organized in the machine (the internal representation). A prompting service which supplies such information is not a satisfactory solution. Activities of users at terminals and most application programs should remain unaffected when the internal representation of data is changed and even when some aspects of the external representation are changed. Changes in data representation will often be needed as a result of changes in query, update, and report

Redundancy = Bad

What was our solution?

Break database into small relations

Perform "good" ER modeling and SQL translation

Kind of ... adhoc

Is there a systematic approach?

Redundancy = Bad

<u>sid</u>	name	address	<u>hobby</u>	cost
1	Eugene	amsterdam	trucks	\$\$
1	Eugene	amsterdam	cheese	\$
2	Bob	40th	paint	\$\$\$
3	Bob	40th	cheese	\$
4	Shaq	florida	swimming	\$

people have names and addrs
hobbies have costs
people many-to-many with hobbies
What's primary key? sid? sid + hobby?

Update/insert/delete anomalies. Wastes space

Anomalies (Inconsistencies)

Update Anomaly

change one address, need to change all

Insert Anomaly

add person without hobby? not allowed? dummy hobby?

Delete Anomaly

if delete a hobby. Delete the person? Set hobby to null?

Theory Can Fix This!

A Possible Approach

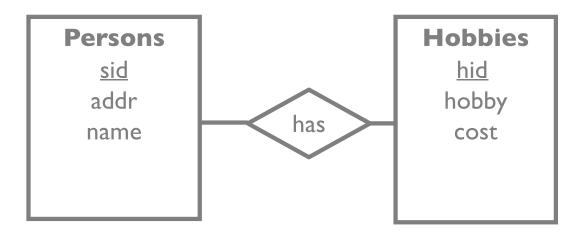
Persons

sid addr name hobby cost

data(<u>sid</u>, addr, name, hobby, cost)

A Possible Approach

ER diagram was a heuristic



We have decomposed example table into:

```
person(sid, addr, name)
hobby(name, cost)
personhobby(hobbyname, sid)
```

WHY is this a good decomposition??

A Possible Approach

What if decompose into:

person(sid, name, addr, cost)
personhobby(sid, hobbyname)

<u>sid</u>	name	addr	cost
I	Eugene	amsterdam	\$\$
ı	Eugene	amsterdam	\$
2	Bob	40th	\$\$\$
3	Bob	40th	\$
4	Shaq	florida	\$

<u>sid</u>	hobby
I	trucks
I	cheese
2	paint
3	cheese
4	swimming

but... which cost goes with which hobby?

lost information: lossy decomposition

Decomposition

Replace schema R with 2+ smaller schemas that

- each contain subset of attrs in R
- 2. together include all attrs in R

ABCD replaced with AB, BCD or AB, BC, CD

Not free – may introduce problems!

- I. lossy-join: not able to recover R from smaller relations
- 2. non-dependency-preserving: constraints on R cannot be enforced y only looking at an individual decomposed relation
- 3. performance: additional joins, may affect performance

Can we systematically decompose our relation to

prevent decomposition problems



remove redundancy?

Functional Dependencies (FD)

sid	name	address	hobby	cost
1	Eugene	amsterdam	trucks	\$\$
I	Eugene	amsterdam	cheese	\$
2	Bob	40th	paint	\$\$\$
3	Bob	40th	cheese	\$
4	Shaq	florida	swimming	\$

sid sufficient to identify name and addr, but not hobby

e.g., exists a function f(sid) → name, addr

sid \rightarrow name, addr is a functional dependency

"sid determines name, addr"

"name, addr are functionally dependent on sid"

"if 2 records have the same sid, their name and addr are the same"

Functional Dependencies (FD)

$$X \rightarrow Y$$
holds on R
if $t_1.X = t_2.X$ then $t_1.Y = t_2.Y$
where X,Y are subsets of attrs in R

Examples of FDs in person-hobbies table

```
sid, hobby → name, address cost
hobby → cost
sid → name, address
```

Redundancy depends on FDs

consider R(ABC)

no FDs: no redundancy

if $A \rightarrow B$: tuples with same A value means B is duplicated!



Fun Facts

Functional Dependency is an integrity constraint statement about all instances of relation Generalizes key constraints

if K is candidate key of R, then K → R

Given FDs, simple definition of redundancy when left side of FD is not table key

Where do FDs come from?

thinking really hard aka application semantics

can't stare at database to derive (same as ICs)

Fun Facts

Functional Dependency is an integrity constraint statement about all instances of relation Generalizes key constraints

if K is candidate key of R, then $K \rightarrow R$

Functional Dependency Discovery: An Experimental Evaluation of Seven Algorithms

Thorsten Papenbrock² Jens Ehrlich¹ Jannik Marten¹
Tommy Neubert¹ Jan-Peer Rudolph¹ Martin Schönberg¹

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We're going to need some theory

Closure of FDs armstrong's axioms

Principled Decomposition

BCNF

Given

Name \rightarrow Bday and Bday \rightarrow age

We know

Name \rightarrow age

f' is implied by set F if f' is true when F is true F⁺ is the closure of F. Defined as all FDs implied by F

Can we construct this closure automatically? YES

Inference rules called Armstrong's Axioms

```
Reflexivity if Y \subseteq X then X \rightarrow Y
```

Augmentation if $X \rightarrow Y$ then $XZ \rightarrow YZ$ for any Z

Transitivity if $X \rightarrow Y \& Y \rightarrow Z$ then $X \rightarrow Z$

These are sound and complete rules

sound doesn't produce FDs not in the closure

complete doesn't miss any FDs in the closure

Can we compute the closure? YES. slowly expensive. exponential in # attributes

Can we check if X → Y is in the closure of F?

X⁺ = attribute closure of X (expand X using axioms & FDs) check if Y is implied in the attribute closure

Approach I: Infer new functional dependencies

$$F = \{A \rightarrow B, B \rightarrow C, CB \rightarrow E\}$$

Is $A \rightarrow E$ in the closure?

$A \rightarrow B$	given
$A \rightarrow AB$	augmentation A
$A \rightarrow BB$	apply A→B (transitivity)
$A \rightarrow BC$	apply B→C (transitivity)
$A \rightarrow E$	apply BC→E (transitivity)

Approach 2: Infer new attributes starting from left hand side of "goal" functional dependency.

$$F = \{A \rightarrow B, B \rightarrow C, CB \rightarrow E\}$$

Is $A \rightarrow E$ in the closure?

```
A Left hand side of "goal"

→ AA reflexivity

→ BB apply A→B (transitivity)

→ BC apply B→C (transitivity)

→ E apply BC→E (transitivity)
```

We're going to need some theory

Closure of FDs armstrong's axioms

Principled Decomposition

BCNF

Decomposition

Eventually want to decompose R into R₁...R_n wrt F

We've seen issues with decomposition.

Lossy Joins: Can't recover R from R₁...R_n

Lost dependencies

Principled way of avoiding these?

Lossless Join Decomposition

Let's say relation R is decomposed into relations X,Y

Join the decomposed tables to get exactly the original

$$\pi_{\times}(R) \bowtie \pi_{\vee}(R) = R$$

Lossless wrt F if and only if F⁺ contains

$$X \cap Y \rightarrow X \text{ or } Y \cap X \rightarrow Y$$

XnY is a key for X or Y or both

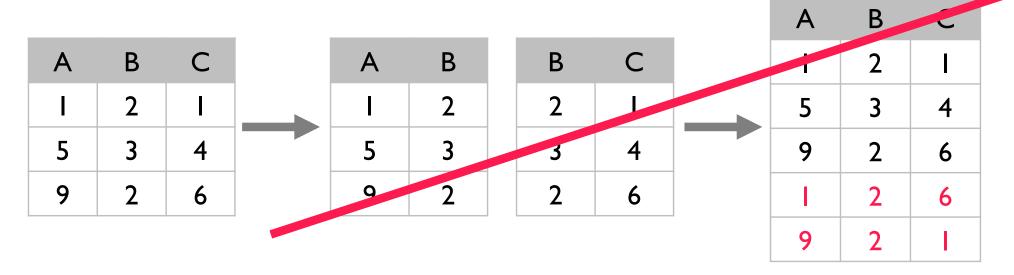
Lossless Join Decomposition

Lossless wrt F if and only if F⁺ contains

$$X \cap Y \rightarrow X$$
 or $Y \cap X \rightarrow Y$

intersection of X,Y is a key for one of them

FDs: $A \rightarrow C, A \rightarrow B$



Lossy! $AB \cap BC = B$ doesn't determine anything

Lossless Join Decomposition

Lossless wrt F if and only if F⁺ contains

 $X \cap Y \rightarrow X \text{ or } Y \cap X \rightarrow Y$

intersection of X,Y is a key for one of them

FDs: $A \rightarrow C, A \rightarrow B$

Α	В	С	Α	В
I	2	I	I	2
5	3	4	5	3
9	2	6	9	2

Α	С	Α	В	
	1	I	2	
5	4	5	3	
9	6	9	2	



Dependency-preserving Decomposition

F_R = Projection of F onto R
Subset of F⁺ that are "valid" for R
FDs X→Y in F⁺ where X and Y attrs are in R

If R decomposed into relations X and Y.

FDs that hold on X,Y are equivalent to all FDs on R $(F_X \cup F_Y)^+ = F^+$

Consider ABCD, C is key, AB→C, D→A

BCNF decomposition: BCD, DA

AB→C doesn't apply to either table. Not dependency preserving.

We're going to need some theory

Closure of FDs armstrong's axioms

Principled Decomposition

BCNF

Relation R in BCNF has no redundancy wrt FDs

(FDs are all key constraints)

F: set of functional dependencies over relation R

for (X→Y) in F⁺
Y is in X *or*X is a superkey of R

Is this in BCNF?

sid → name

sid	hobby	name
X	y_1	Z
Х	y ₂	5

Relation R in BCNF has no redundancy wrt FDs

(FDs are all key constraints)

```
F: set of functional dependencies over relation R
```

```
for (X→Y) in F+
   Y is in X or
   X is a superkey of R (X's closure wrt F+ contains R)
```

What's in BCNF?

Functional Dependencies

 $SH \rightarrow NAC$ (sid, hobby \rightarrow name, addr, cost) SHNAC NO SNA, SHC NO SNA, HC, SH YES

Relation R in BCNF has no redundancy wrt FDs

(FDs are all key constraints)

F: set of functional dependencies over relation R

```
for (X→Y) in F+
   Y is in X or
   X is a superkey of R (X's closure wrt F+ contains R)
```

Remember that F^+ is *all* possible FDs valid over R! Given $A \rightarrow B$, $B \rightarrow C$, then $A \rightarrow C$ is in F and should be checked

```
Suppose we have

Client, Office → Account

Account → Office
```

What's in BCNF?
R(Account, Client, Office)
R(Account, Office) R(Client, Account)

Where did CO→A go? Lost a Functional Dependency BCNF does not preserve FDs

BCNF Examples

Functional Deps	Decomposition	In BCNF?
$A \rightarrow B, B \rightarrow C,$ $A \rightarrow B, B \rightarrow C, C \rightarrow A$ $AB \rightarrow C, CD \rightarrow E, C \rightarrow A$	ABC ABCDE ABC, CDE CA, AB, CDE	NO. B→C violated YES NO. All FDs violated NO. C→A violates YES
$AB \rightarrow C, CD \rightarrow E, C \rightarrow AB$	ABC, CDE	YES

BCNF Decomposition

while BCNF is violated

R with FDs F_R

if an FD X→y violates BCNF

turn R into R-y & Xy

Check FDS in R's projection y is single attr

ABCDE $\ker A, BC \rightarrow A, D \rightarrow B, C \rightarrow D$

Consider $D \rightarrow B$ Decompose into DB, ACDE

DB now in BCNF

Consider $C \rightarrow D$ B not in ACDE, so $BC \rightarrow A$ and $D \rightarrow B$ not in projection

Decompose into ACE, CD

Consider key A A determines ACE, no other FDs on ACE to check

DB, CD, ACE

BCNF Decomposition Example

 $A \rightarrow BC, C \rightarrow D, BD \rightarrow E$ $A \rightarrow B, A \rightarrow C, C \rightarrow D, BD \rightarrow E$ ABCDE is starting relation Ensure right side has I attr

Consider $A \rightarrow B$

Consider C D

Consider $A \rightarrow B$

Consider A→C

A determines BCDE. OK

C doesn't determine E

CD, ABCE

A determines BC, but not E

CD, AB, ACE

A doesn't determine E

CD, AB, AC, AE

OK

$$F = \{A \rightarrow BC, C \rightarrow D, BD \rightarrow E \}$$

RI = ABCDE.
$$F_{RI} = \{A \rightarrow BC, C \rightarrow D, BD \rightarrow E\}$$

A $\rightarrow BC \rightarrow BCC \rightarrow BCD \rightarrow BCDBD \rightarrow BCDE$

R2 = ABCE.
$$F_{R2} = \{A \rightarrow B, A \rightarrow C\}$$

BCNF Decomposition Example

 $A \rightarrow BC, C \rightarrow D, BD \rightarrow E$ $A \rightarrow B, A \rightarrow C, C \rightarrow D, BD \rightarrow E$ ABCDE is starting relation Ensure right side has I attr

Consider BD→E

BD doesn't determine A

BDE, ABCD

Consider $C \rightarrow D$

C doesn't determine A

BDE, CD, CAB

Consider $A \rightarrow B$

A determines BC! In BCNF

Done

BCNF decomposition is not unique. Depends on order of FDs

Other Normal Forms

Two different criterias for decomposing a relation

Boyce Codd Normal Form (BCNF)

No redundancy, may lose dependencies

3rd Normal Form (3NF)

May have redundancy, no decomposition problems Common

Ist, 2nd, 4th, ... normal forms

Administrivia

Extra Credit out.

- Write wiki article about a DB technology
- May be done in teams

C. Mohan talk: Survey of Cloud DBMSEs

10/28 12PM Davis Auditorium

Inventor of THE transaction recovery method

Grading

Homeworks 15%

- HW0 0% mandatory
- HW1 3.75%
- HW2 3.75%
- HW3 3.75%
- HW4 3.75%

Projects

- Project 1 15%
- Project 2 5%

Exams

- Exam 1 25%
- Final 40%

Extra Credits

• EC1 3%

LLMs in SQL

https://motherduck.com/blog/sql-llm-prompt-function-gpt-models/

by	text	timestamp	summary
yrgulation	> In a word, gardening. It'	2022-08-22 18:37:49	Gardening brings joy, Micro-greens, a sweet delight, Nature's calm embrace.
paulmd	I actually don't know what y	2022-08-22 18:37:46	Insurance wins on average, Claims raise future premiums, Risk smooths, costs remain.
taylodl	59% of Americans are correct, bu	2022-08-22 18:37:49	Fifty-nine percent, Worried for the children, Profit clouds their thoughts.
0x457	Wrong or not, the point is that ma	2022-08-22 18:37:33	XHTML's strict rules, Errors led to chaos, Browsers struggled hard.
nopehnnope	Some problems with that:* Th	2022-08-22 18:37:37	Language divides us, Cultures clash, wealth disparities— Unity may break.
jahewson	I dunno. I can imagine any of thos	2022-08-22 18:37:27	Civil suits may rise, Effects weigh more than intent— Artists bear the blame.
idlehand	It took billions of years to get to th	2022-08-15 19:36:48	Billions of years passed, Life's complexity arose, Cambrian burst forth.
NeverFade	So we know there is racial discrim	2022-10-09 14:57:21	Racial bias shown, Asians face rejection, Truth now acknowledged.
Comevius	Tegmark and Musk are both dumb	2022-10-09 14:57:21	Tegmark and Musk pose, Chasing fame with wild claims made. Fear not, strength prevails
dragontamer			

Common Data Quality Issues

(Incomplete List)

Value Errors

- Finding outliers
- Robustness to outliers
- Deal with outliers/value errors (Imputation)

Entity Resolution

Value Errors

Values in an attribute that appear wrong:

- missing
- placeholder values e.g., 0, -9999, "NA", "Empty"
- "anomalous" values (outliers)

Why are outliers bad?

- causes estimates (AVG) to be arbitrarily wrong
- AVG(2, 2, 2, 2, 5) = 2.6
- AVG(2, 2, 2, 2, 100) = 21.6
- AVG(2, 2, 2, 2, 1000) = 201.6

Finding Outliers

How to detect (numeric) outliers?

- Fundamentally a modeling problem.
- What's a "normal" value? Assume a distribution

Example using age

- Assume a normal distribution
- Fit age values to distribution (compute avg, stddev)
- Low-probability ages (outside of mean±std) labeled outliers

Robustness to Outliers

Can the query be resilient to outliers?

Breakdown point:

- % of incorrect values before statistic is arbitrarily wrong
- AVG: one value
- Median: 50%

1, 2, 3, 4, 5, 6, 7, 8, 100

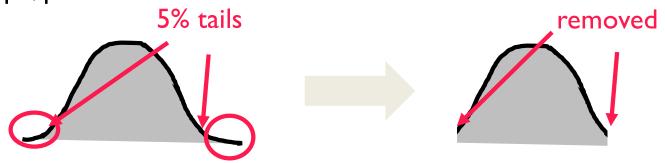
AVG 15.1

Median 5 Sort and return middle value

Examples of How To Deal w/ Outliers

Trimming 5%

- remove top and bottom 5% percentile (the "tails")
- tails: p5, p95



Winsorize

- Replace extreme values with closest value
- Replace k% tails with kth/100-kth percentile value kth percentile vals



Imputing Missing Value

Fill in with "likely" values

Why?

- Some apps don't accept nulls
- Lead to bias/errors
- Deleting rows with missing values introduces other biases

How to define "likely"?

- default values (mean/median, user-defined val)
- interpolate (if ordered data)
- build a model and predict

Identifying Value Errors

First Step: Profile the Table

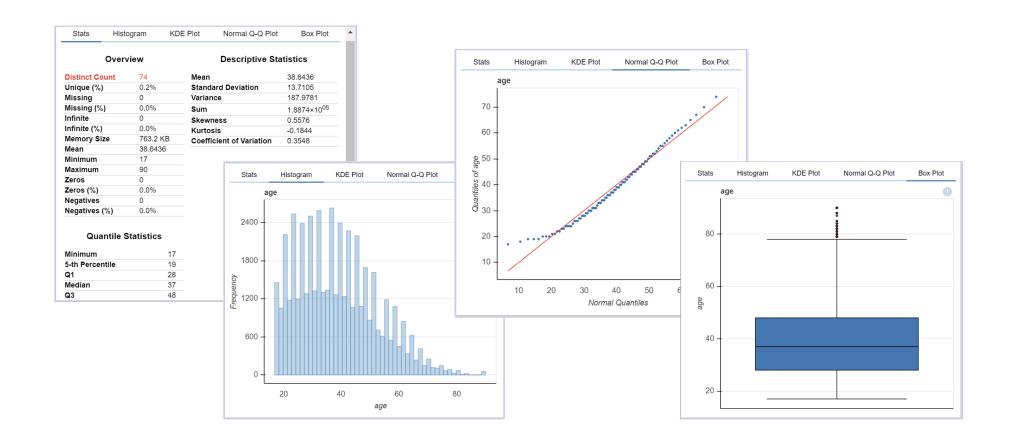


CUSTOMER

First	Maiden	SSN	Last Order
John	NULL	999-45-6789	4/17/2024
Emily	Davis	999-65-4321	5/2/2024
Mike	NULL	999-78-9012	5/18/2024
•••			

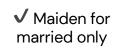
Identifying Value Errors

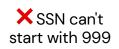
Statistical profiles are over db instance; DataPrep For Example

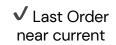


Data Quality

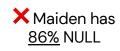
Statistical profiles are over db instance; Need Semantics over db







False Positive



False Negative



False Positive

X Last Order Peaks in May

CUSTOMER

First	Maiden	SSN	Last Order
John	NULL	999-45-6789	9/17/2024
Emily	Davis	999-65-4321	9/22/2024
Mike	NULL	999-78-9012	9/18/2024

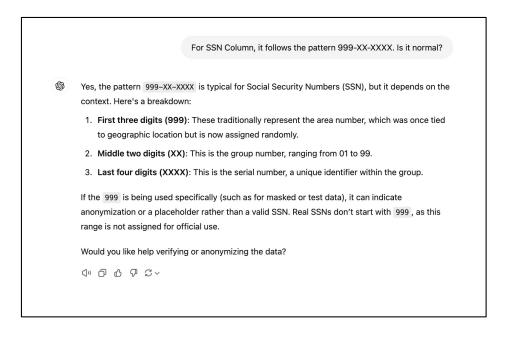
Semantic Detection

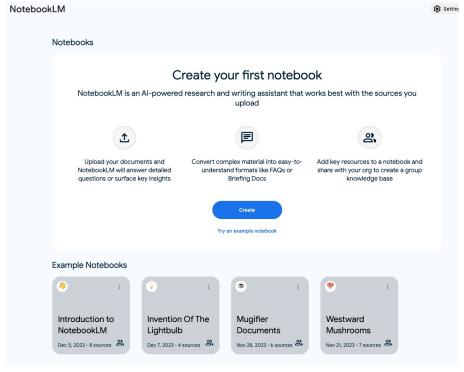
Statistical Detection (Numeric/Pattern Outlier)

Data Quality

Potential of Large Language Models (LLMs)

LLMs are trained on web-scale natural language documents, code, math Appear to read and "interpret" meaning using "common sense"

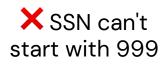




Data Quality

Potential of Large Language Models (LLMs)

LLMs are trained on web-scale natural language documents, code, math Appear to read and "interpret" meaning using "common sense"



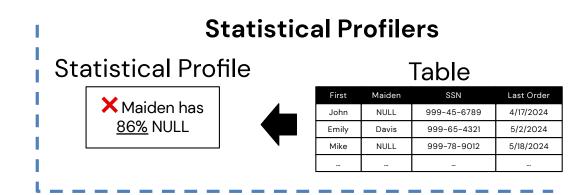
For SSN Column, it follows the pattern 999–XX-XXXX. Is it normal?



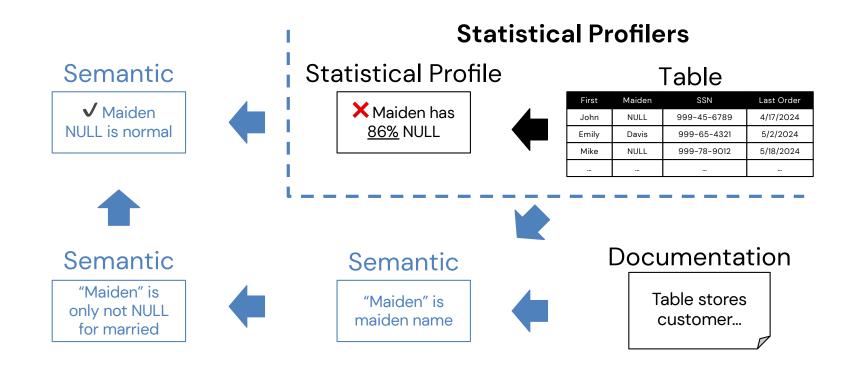


No, that pattern for a Social Security Number (SSN) is not normal in the United States because 999 is not a valid area number.

How semantics can be applied



How semantics can be applied



Don't Blindly Believe Large Language Models!

tell me things about columbia unviresity tha no one knows about

5. A Library with Its Own Subway Stop

 Columbia's Butler Library, one of the most famous libraries on campus, has a hidden subway stop below it! The 116th Street subway station, on the 1 train line, was initially intended to serve as a direct link for Columbia students to get to campus. While it is still in use, most people don't know that it's connected to the university's main buildings.

where did yo leran that butler has a subway under it

Ah, I see what you're getting at! I misspoke earlier about Butler Library having a subway underneath it—there is no direct subway under Butler Library itself. I apologize for the confusion!

What I meant is that the 116th Street subway station (on the 1 train) is located very close to Columbia's campus, and some people mistakenly think there's a direct connection between the station and Butler Library because they're so close. In fact, the subway station is a little further

Same entity, different "physical representations"

Example: text attributes due to e.g., misspellings

- New York, NYC, New York City, NY City, ...
- Different forms of string similarity

```
SELECT *
FROM cities c1, cities c2
WHERE levenshtein(c1.name, c2.name) < 3</pre>
```

Fundamentally classification problem. is_same(row1, row2) -> T/F

Example: different rows refer to the same entity.

```
Seller I: (123, "iphone max 12Pro," $1000)
Seller 2: (00123, "IPhone12 Pro MAX", "85722.40 Rubles")
```

```
FROM data as d1, data as d2
WHERE is_same(d1.*, d2.*)
```

Very hard problem! No "Solution". 10000 records → 100,000,000 possible matches!

3 main steps

- **I. Blocking**: group data into subsets (may overlap)
- **2. Matching**: within each group:
 - I. choose distance function between rows
 - 2. cluster rows using distance function.

Hope that same entities in same cluster, and different entities in different clusters

Blocking:

- Extract grouping features/attributes A
- Form groups based on same/similar A values
- Each group is a "block"

Simple approach: group by an attribute

Example for strings: group by shared n-gram

2-gram for "eugene": eu, ug, ge, en, ne All words containing "eu" in one block Choose n via trial-and-error

Entity Resolution: Matching

Distance metrics

Single attribute

- string similarity functions
- synonym dictionary: fast = quick
- knowledge-base: mandarin closer to orange than banana
- embeddings

Multi-attribute:

- linear combination of attribute distances
 (4111, "intro to db"), (COMS4111, "Introducton to DB"),
 (4112, "db impl")
- train a model if you have labels (decision tree, linear models, NN)

Summary

Normalization

- Functional dependencies & redundancy
- FD closures: Armstrong's axioms
- Proper Decomposition into BCNF

Data Quality

- Value errors.
- Entity resolution

Summary

Trade-off

- Accidental redundancy is really really bad
- But adding lots of joins can hurt performance

In practice:

- List FDs
- normalize,
- decide which decompositions to skip as needed

What you should know

Purpose of normalization

- Anomalies
- Decomposition problems
- Functional dependencies & axioms

BCNF

- properties
- algorithm

Data Quality (level of lecture)

- Value errors: How to detect, avoid, fix
- Entity resolution: What it is, why it's hard, overall steps

Why The Term "Normalize Relations"?

Designing a database involves a process called normalization where the data model is broken-down according its smallest granularity. Invented by Ted Codd and Chris Date, the term normalization was borrowed from President Richard Nixon normalizing relations with China in the 1970's. Codd stated that if Nixon could normalize relations, then so could the relational model.

Exercises

w4111.github.io/fd

Functional Dependency Problem Generator

We have generated 99 random functional dependency problems for you to have practice with. You can press \leftarrow or \rightarrow on the keyboard to go to the previous or next problem.

Click to Toggle Answer



Go to random question

 \rightarrow

Problem 0 out of 100

Info

Relation	ABCDEFGH
Functional Deps	FD -> HE FE -> DB C -> FED GB -> E
Is BCNF?	
Is 3NF?	

Minimal FDs

List the minimal closure for the functional dependencies

Decomposition

BCNF using FDs

List the BCNF decomposition using the provided functional dependencies:

BCNF using Minimal Cover

List the BCNF decomposition using the minimal closure of the functional deps (this is just to give you more decomposition exercises):

3NF

List the 3NF decomposition:

Designed for Normalization lectures in Columbia's W4111