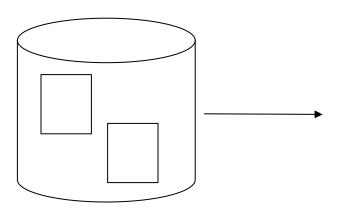
Data Cleaning, Transformation, Enrichment



AnHai Doan

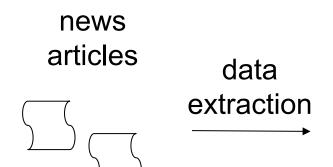
Motivation





id	name	loc
x ₁	Apple	CA
X_2	IBM	NY





id	cname	address	rev
y ₁	IBM Corp	CA	25
y ₂	Apple Inc	CA	51
y ₃	GE	NY	351

id	name	loc
x ₁	Apple	CA
x ₂	IBM	NY



id	cname	address	rev
y ₁	IBM Corp	CA	25
y ₂	Apple Inc	CA	51
y ₃	GE	NY	351

data cleaning: GE revenue: 351 → 35.1

schema matching: name = cname

loc = address

X(name, loc) schema merging:

Y(cname, address, rev) Z(name, loc, rev)

data matching:

xid	yid
x ₁	y ₂
\mathbf{x}_2	y ₁

data merging: for name, return the longer string from X.name and Y.cname

for loc, return X.loc

schema mapping: Z = select merge_name(X.name, Y.cname), X.loc, Y.rev

from X, Y, M

where X.id = M.xid and Y.id = M.yid

X

id	nam e	loc
x ₁	Apple	CA
x ₂	IBM	NY

Y

id	cname	addres s	rev
y ₁	IBM Corp	CA	25
y ₂	Apple Inc	CA	51
y ₃	GE	NY	351

Z

name	loc	rev
Apple Inc	CA	51
IBM Corp	NY	25

Three Goals for Data Cleaning

- Detect errors (aka data quality problems)
- Decide which errors to fix
- Fix errors
- Often do the above using dictionaries, rules, ML

Data

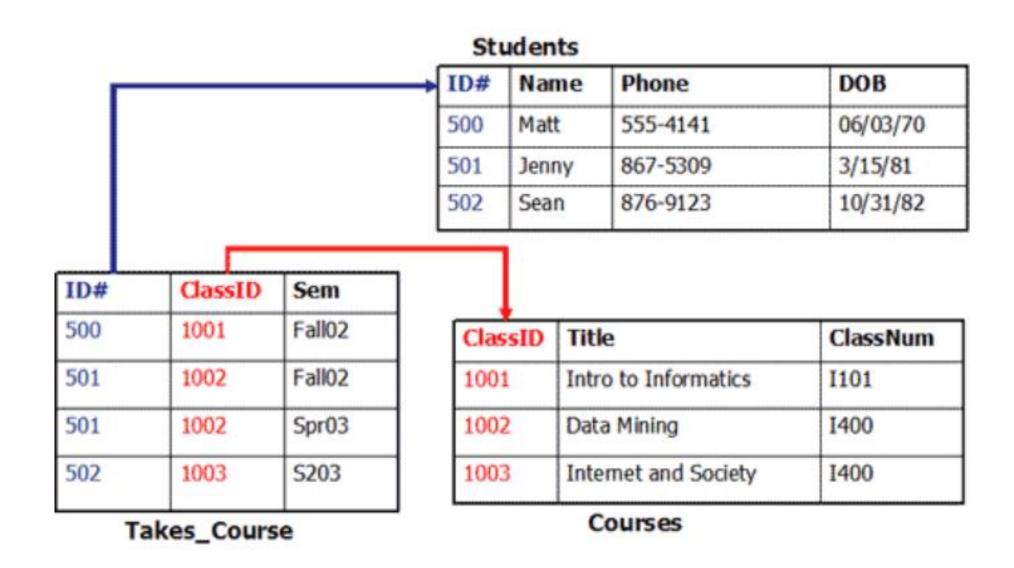
Typically taken to mean schema + data instances

		Attr	ibutes	_	Schem
	StudentID	Name	Phone	DOB	-
	111335555	Matt	555-4141	06/03/70	
Tuple →	111224444	Troy	556-9123	01/02/76	
\	999775555	Sean	876-5150	10/31/81	
1	444668888	Christy	219-7734	02/14/84	

- Ideally we should use "schema" and "data instances"
- But often we will say "schema" and "data"

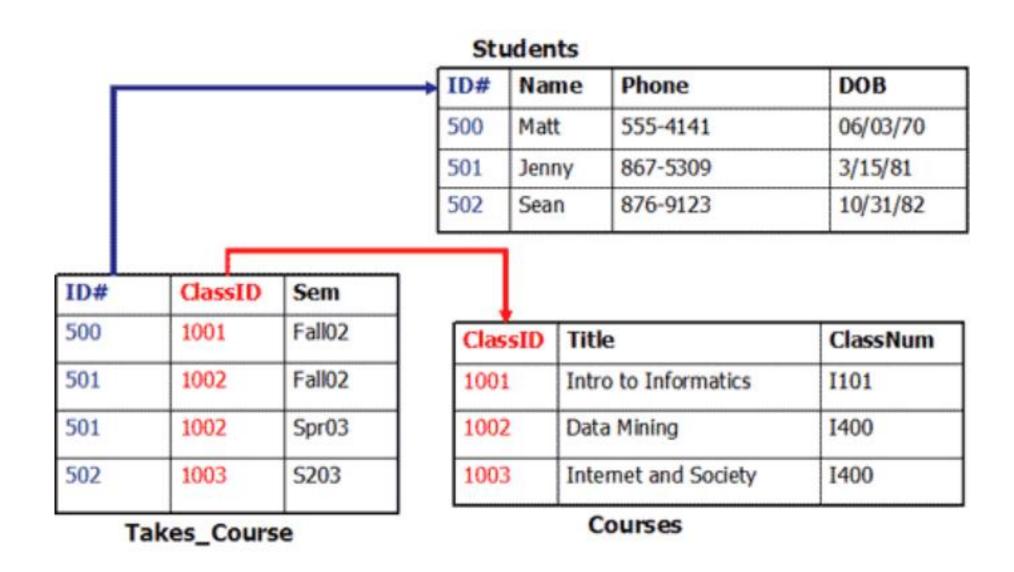
Schema Often Has Many Constraints

Key, uniqueness, functional dependencies, foreign keys



Data Often Has Many Constraints Too

value range, format, etc.



Detecting the Problems

Schema problems

- mispelt names
- violating constraints (key, uniqueness, foreign key, etc)

Data problems

- missing values
- incorrect values, illegal values, outliers
- synonyms
- mispellings
- conflicting data (eg, age and birth year)
- wrong value formats
- variations of values
- duplicate tuples

Deciding Which Problems to Fix

- Do not have to fix all
- Only those that are necessary for the business purpose
- Example
 - extract brand, color, and weight from product descriptions
 - purpose: to allow users to browse products based on brand
 - so brand values should be correct, ideally 100%
 - color and weight values do not have to be entirely correct

Fixing the Problems

- Good tools exist for certain types of attributes
 - names, addresses
- But in general no real good generic tools out there
- Much research has been done
- People mostly roll their own set of tools
 - lot of these use dictionaries (aka reference data), rules, ML
- Using dictionaries (aka reference data)
 - typically to detect standardization problems and then fix those
- Using ML
 - try to predict what a value in a column should be, then compare with the existing value
 - if different, then could be an error

Dirty Data

FirstName	Surname	CompanyName	Address1	Town
peter	jones	ones café	80 riverways	manchester
isa seffon			76 the avenue	leicester
a baker		bakery baker itd	7 main road	reading benishire
Richard	Evans1	Richard's Treats	9 charles Street	Bracknel
Alex	-	The Alex Centre	13-15 athol street	Bournmouth
Derren	Knight0	Derrens' Delights	I Transport Value	Gillingham
Janine	Sea Contract	The Janine Way	10 Fleet Place	Bracknelli
Katherine	Botton	Botton Foods	bond Street	
Emma	Wright	The Write Way Pld	280 Bath road	Birmingham
emma	W	The Write Way	280 Bath rd	Birmingham
David	Smith	Dave's Gifts	PO 80X 21	Leigh
Dave	Smith	Dave's Gift	po box	Leigh Lancs

Un-Standardised

Missing or misspelled

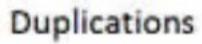
Duplications



FirstName	Surname	CompanyName.	Address1	Town
Peter	Jones	Jones Café	80 Riverways	Manchester
Lisa	Sefton		76 The Avenue	Leicester
A	Baker	Bakery Baker Ltd	7 Main Road	Reading
Richard	Evans	Richard's Treats	9 charles Street	Brackneit
	and a last to the	Water Street Street	The second second	No. of Control of the Control

Correctly Standardise

Emma	Wright	The Write Way Pld	280 Batti road	Eirmingham
emma	W	The Write Way	280 Bath rd	Birmingham
David	Smith	Dave's Gifts	PO 80X 21	Leigh
Dave:	Smith	Dave's Gift	po box	Leigh Lancs





FirstName	Surname	CompanyName.	Address1	Town
Peter	Jones	Jones Café	80 Riverways	Manchester
Lisa	Setton		76 The Avenue	Leicester
A	Baker	Bakery Baker Ltd	7 Main Road	Reading
Richard	Evans	Richard's Treats	9 charles Street	Brackneit
Alex:	Froy	The Alex Centre	13-15 athol street	Bournmouth
Derren	Knight0	Derrens' Delights	25 Camel Lane	Gillingham
Janine	Hutton	The Janine Way	10 Fleet Place	Bracknett
Catherine	Botton	Botton Foods	bond Street	London
Emmä	Wright	The Write Way Pid	280 Bath road	Birmingham
David	Smith	Dave's Gifts	PO BOX 21	Leigh

Correctly Standardise

Populated and Corre

Duplications Remove

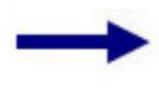
Examples in Industry (see Google Doc)

Additional Transformations

- These are not to correct something wrong in schema/data per se
- Not data cleaning
- But rather transformations of schema/data into something better suited for our purposes
- Examples
 - split a field (eg full name)
 - concat of multiple values/fields
 - schema transformation

Example of Splitting an Attribute

Name
John Smith
Henry R. White
Dr. Andy Brown
Steve D. Brook Jr.



Title	First	Middle	Last	Suffix
	John		Smith	
	Henry	R.	White	
Dr.	Andy		Brown	
	Steve	D.	Brook	Jr.

Some Other Possible Steps

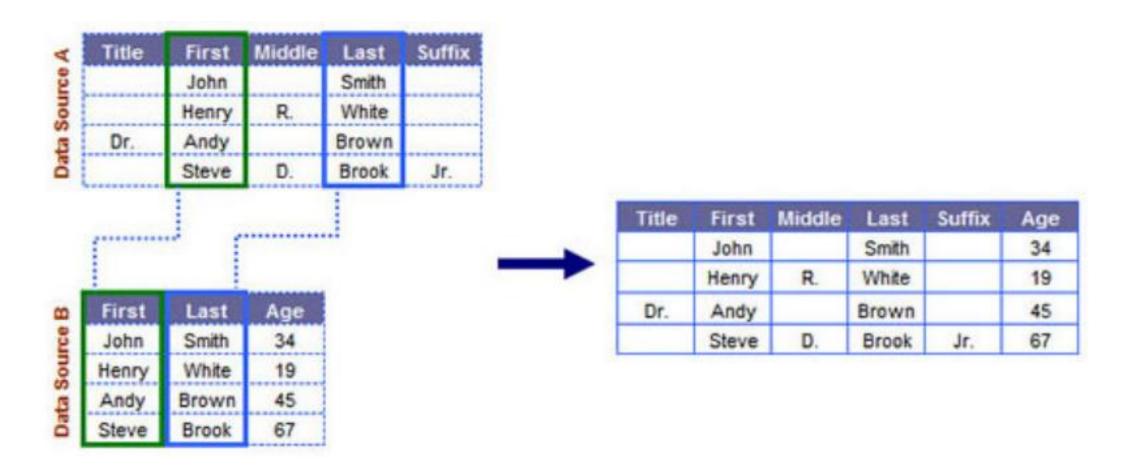
Data enrichment using additional data sources

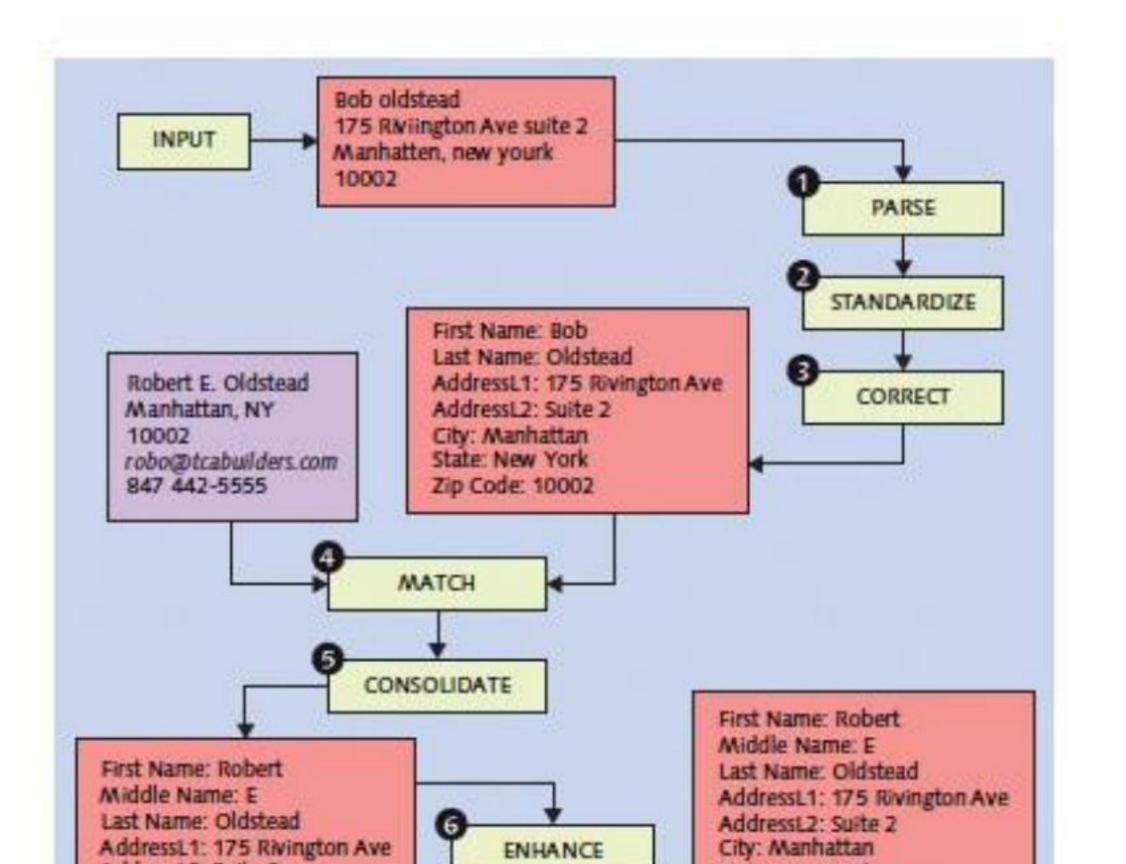
First	Last	Income
John	Smith	\$ 32,000
Henry	White	\$ 88,000
Andy	Brown	\$120,000
Steve	Brook	\$ 54,000

come L	Income U	Target
20000	39999	Α
40000	59999	В
60000	79999	С
80000	99999	D
100000	119999	E
120000	139999	F

First	Last	Income	Target
John	Smith	\$ 32,000	Α
Henry	White	\$ 88,000	D
Andy	Brown	\$120,000	F
Steve	Brook	\$ 54,000	В

Another Example





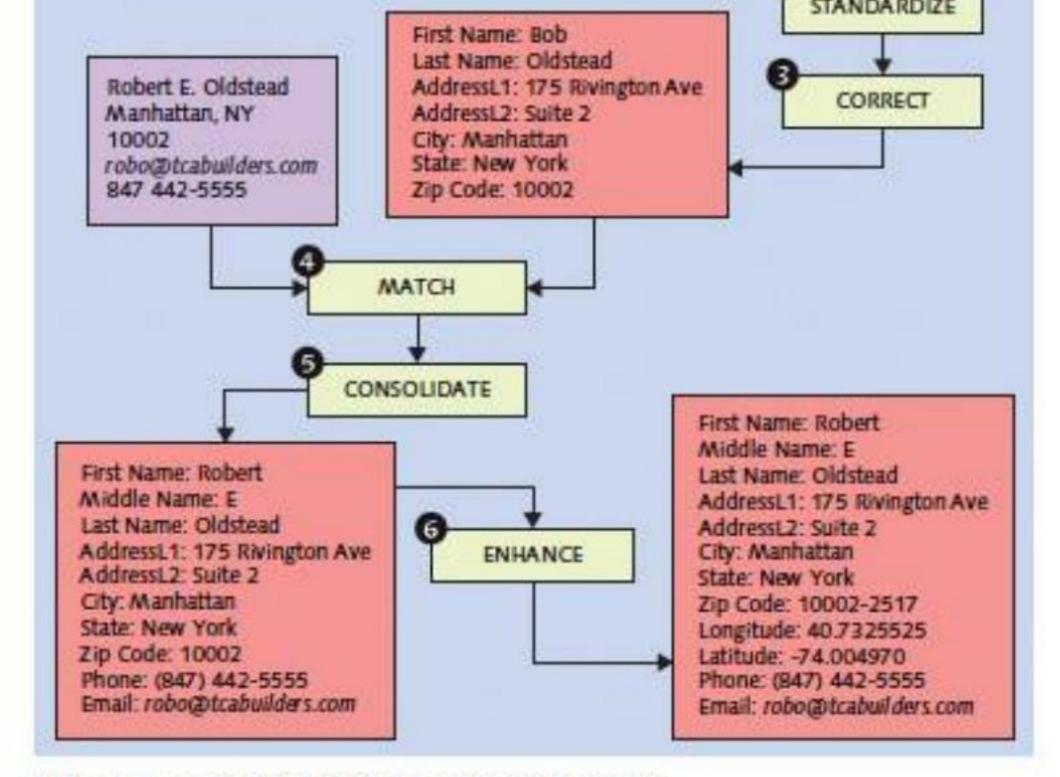


Figure 4.6 Example of the Data Quality Process

Discussion

- Exploration and cleaning are often required in many steps of the DS pipeline
 - after integrating multiple data sets, may have to explore and clean again
 - profiling and cleaning are often built into data transformation steps

Data quality is a huge topic

- the notion of data quality covers many additional aspects
 - freshness, coverage, understandability, trustworthiness, etc.
- can be addressed within a single DS pipeline, or long-term within a department or the entire company, as a part of data governance
 - e.g., assign an owner to each data set, whose job is to keep the data quality high