Data Cleaning

February 12, 2019

Data Science CSCI 1951A

Brown University

Instructor: Ellie Pavlick

HTAs: Wennie Zhang, Maulik Dang, Gurnaaz Kaur

A message from your Health and Wellness Advocates!

- Health and Wellness Resource Fair Feb 20 (Wed),
 4-6PM, CIT 1st Floor Atrium
- Open Hours! wellness.advocates@lists.brown.edu

Announcements

- A message from your Health and Wellness Advocates!
- iClicker syncing—the saga continues
- Collab policy.......
- Final project teaming—fill out the form please!

Announcements

COLLABORATION POLICY!

Announcements

COLLABORATION POLICY!

SERIOUSLY, PEOPLE....

Today

- Problems with dirty data
- Cleaning and string matching heuristics
- Bash commands—greatest hits (for data scientists)

ID	Name	Street	City	State	Zip	Hours
1	J Meltzer	123 University Ave	Providence	RI	98106	42
2	Erin Bugbee	245 3rd St	Pawtucket	RI	98052-1234	30
3	David Wang	345 Broadway	PVD	Rhode Island	98101	19
4	E Bugbe	245 Third Street	Pawtucket	NULL	98052	299
5	Dave Wang	345 Broadway St	Providnce	Rhode Island	98101	19
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7	Haomo Ni	123 University Ave	Providence	Guyana	94305	NULL

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3	David Wang	345 Broadway	PVD	Rhode Island	98101	19
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Inconsistent Representations

ID	Name	Street	City	State	Zip	Hours
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Typos

Duplicates

Problems?

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Maybe Duplicates?

Data is dirty on its own

- Data is dirty on its own
- Data sets are clean on their own but combining them introduces errors (e.g. duplicates, different naming conventions)

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- Data sets are clean on their own but combining them introduces errors (e.g. duplicates, different naming conventions)
- Data doesn't "age well" (inflation, restricting)
- Any combination of the above

• Parsing input data (e.g., separator issues)

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- Naming conventions: NYC vs New York

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- Primary key violations (from data merging)

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- Naming conventions: NYC vs New York
- Formatting issues esp. dates
- Missing values and required fields (e.g., always use 0)
- Different representations (2 vs Two)
- Fields too long (get truncated)
- Primary key violations (from data merging)
- Redundant Records (from data merging)

TAS

ID	Name	City	State	Hours
1	J Meltzer	Providence	RI	42
2	Erin Bugbee	Pawtucket	RI	30
3	David Wang	PVD	Rhode Island	19
4	E Bugbe	Pawtucket	NULL	300
5	Dave Wang	Providence	Rhode Island	19
6	Jacob Meltzer	PVD	Rhode Island	42
7	Haomo Ni	Warwick	RI	NULL

ID	Name	City	State	Hours
1	J Meltzer	Providence	Rhode Island	42
2	Erin Bugbee	Pawtucket	Rhode Island	30
3	David Wang	Providence	Rhode Island	38
7	Haomo Ni	Warwick	Rhode Island	0

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	Name	City	State	Hours
1	J Meltzer	Providence	RI	42
2	Erin Bugbee	Pawtucket	RI	30
3	David Wang	PVD	Rhode Island	19
4	E Bugbe	Pawtucket	NULL	300
5	Dave Wang	Providence	Rhode Island	19
6	Jacob Meltzer	PVD	Rhode Island	42
7	Haomo Ni	Warwick	RI 🙍	NULL

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3	David Wang	Providence	Rhode Island	38
7	Haomo Ni	Warwick	Rhode Island	0

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TD	Name	City	State	Hours
1	J Meltzer	Providence	RI	42
2	Erin Bugbee	Pawtucket	RI	30
3	David Wang	PVD	Rhode Island	19
4	E Bugbe	Pawtucket	NULL	300
5	Dave Wang	Providence	Rhode Island	19
6	Jacob Meltzer	PVD	Rhode Island	42
7	Haomo Ni	Warwick	RI 🧑	NULL

ID 🕙	Name	City	State	Hours
*1	J Meltzer	Providence	Rhode Island	42
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3	David Wang	Providence	Rhode Island	1 38 1 1 1 1 1 1 1 1 1 1
7	Haomo Ni	Warwick	Rhode Island	0

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How will the dirty data affect the results of this query?
(a) Too high
(b) Too low
(c) Unaffected

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How many TAs are there?

SELECT COUNT (*)
FROM TAS

How will the dirty data affect the results of this query?
(a) Too high
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	36	Name	City	State	Hours
1		J Meltzer	Providence	RI	42
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3	3	David Wang	PVD	Rhode Island	19
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5	5	Dave Wang	Providence	Rhode Island	19
6	3	Jacob Meltzer	PVD	Rhode Island	42
7	7	Haomo Ni	Warwick	RI 🙍	NULL

ID 🗸	Name	City	State	Hours
1	J Meltzer	Providence	Rhode Island	42
2	Erin Bugbee	Pawtucket	Rhode Island	30
3	David Wang	Providence	Rhode Island	
7	Haomo Ni	Warwick	Rhode Island	0

How many TAs are there?

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(a) Too high
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Duplicates ->
Double Counting

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How many TAs have worked zero hours?

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Hov	v will the dirty
dat	a affect the results
of t	this query?
	Too high
(b)	Too low
(c)	Unaffected

ID 💜	Name	City	State	Hours
% 1	J Meltzer	Providence	Rhode Island	42
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How many TAs have worked zero hours?

NULLS arent included in the where clause

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How will the dirty data affect the results of this query?
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How many hours do my commuter TAs work?

SELECT SUM (Hours)

FROM TAS

WHERE City != "Providence"

Clicker Lightening Round!

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1	J Meltzer	Providence	RI	42
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How will the dirty data affect the results of this query?
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Inconsistent names, typos,

How many hours do my commuter TAs work? and

SELECT SUM (Hours)

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duplicates...

- Look at your data!
- Maybe set (sensible) defaults

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- When you issue a query, don't take the answer as gospel.

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 Instead...

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 Instead...wait for it...

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- Look at your data
- When you issue a query, don't take the answer as gospel.
 Instead...wait for it...look at your data!

```
SELECT City, COUNT(*) as pop
FROM PEOPLE
GROUP BY Zip_Code
ORDER BY pop
```

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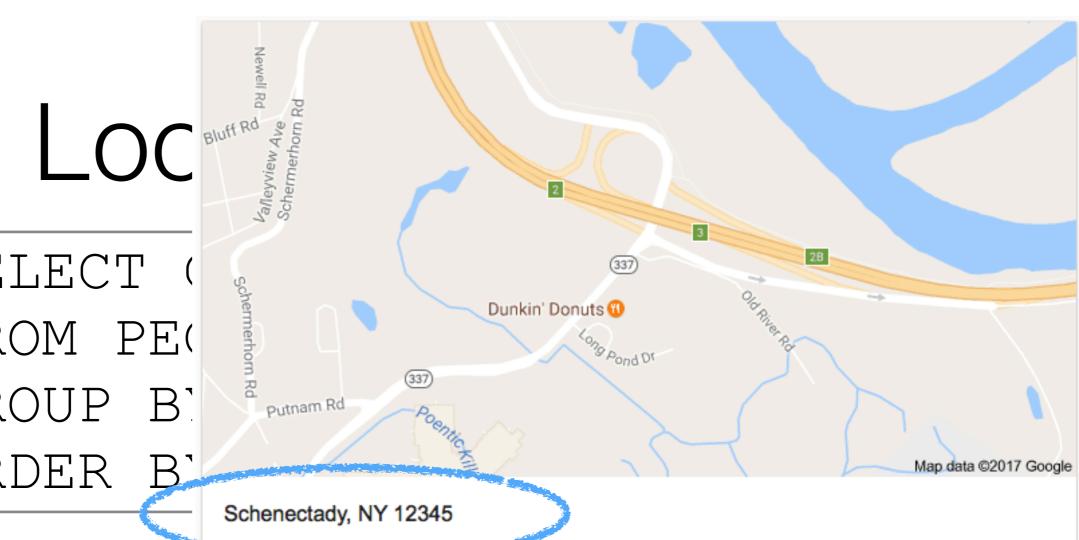
City	Count(*)
Schenectady	2,500
New York City	2,200
Los Angeles	1,900
Dallas	1,400

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SELECT City, COUNT(*) as pop
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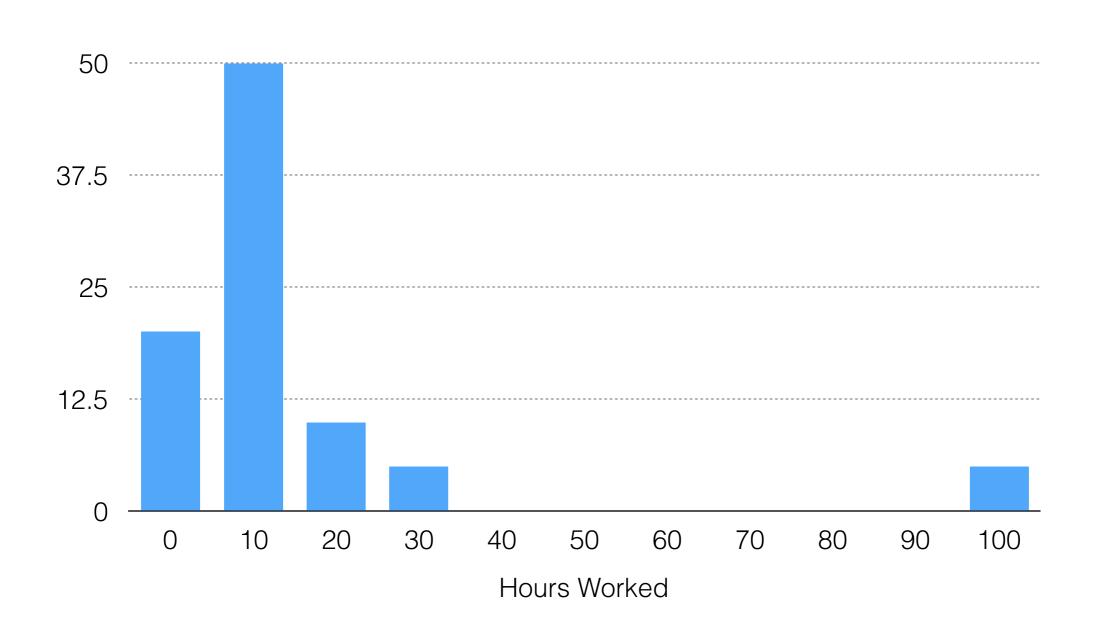
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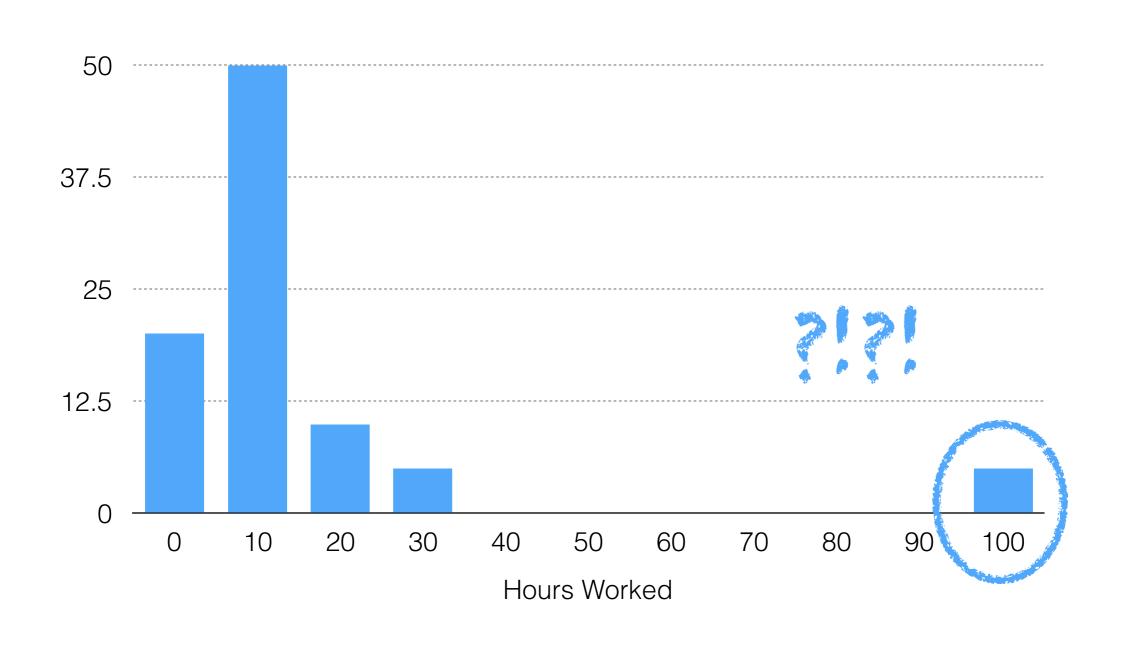


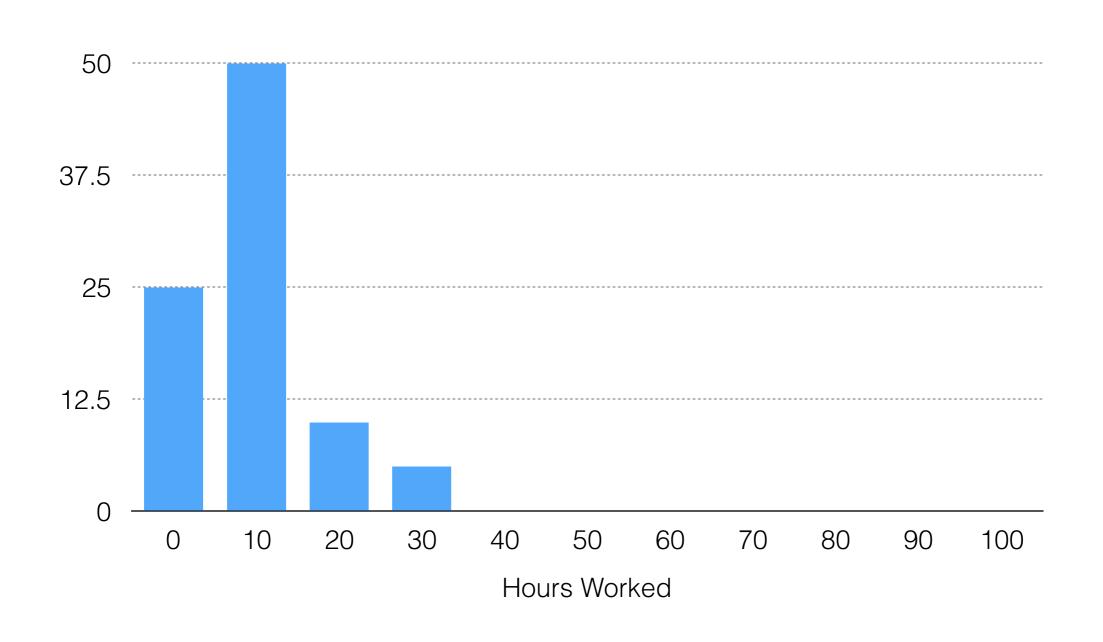
SELECT (FROM PE(GROUP B' ORDER B

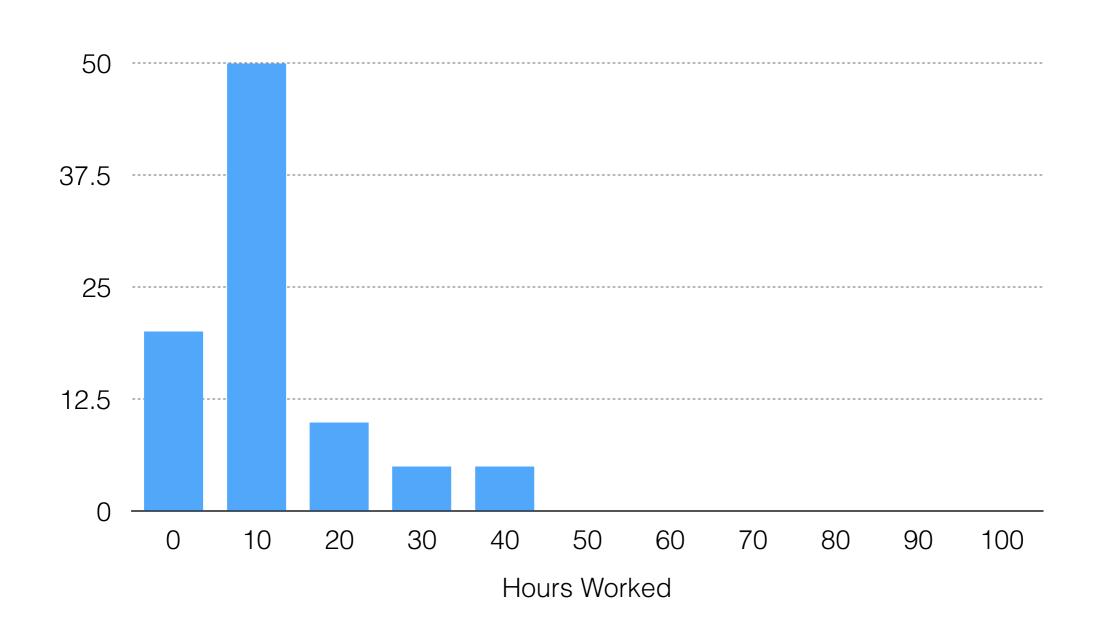


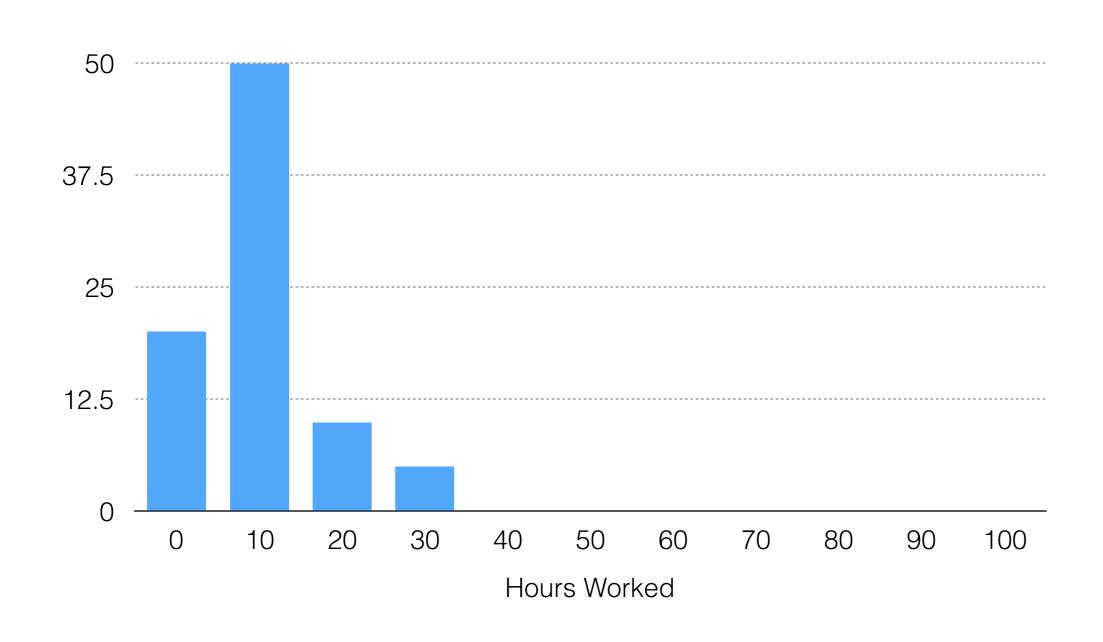
City	Count(*)
12345	2,500
10001	2,2000
90001	1,900
75001	1,400

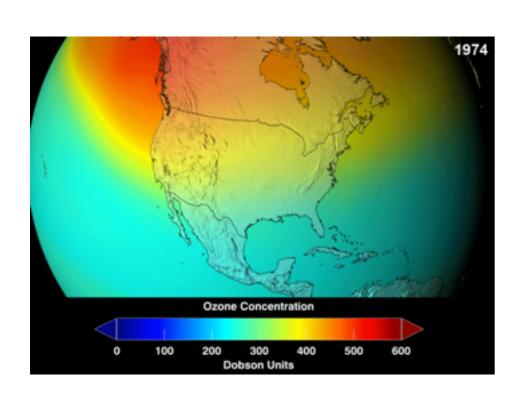






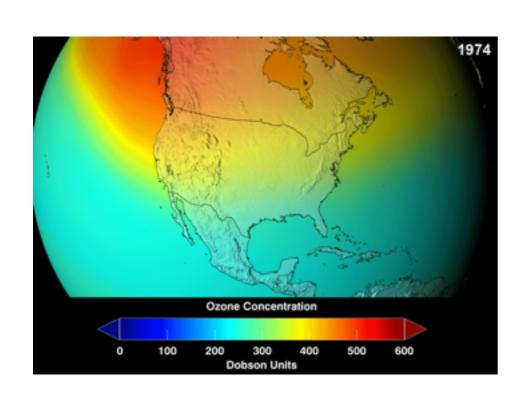






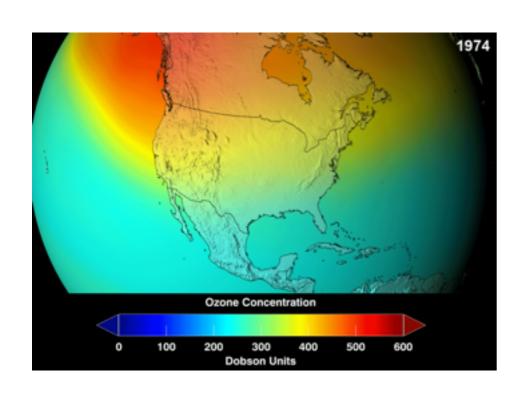
The discovery of the Antarctic "ozone hole" by British Antarctic Survey scientists Farman, Gardiner and Shanklin...came as a shock to the scientific community...[The data] were initially rejected as unreasonable by data quality control algorithms (they were filtered out as errors since the values were unexpectedly low); the ozone hole was detected only in satellite data when the raw data was reprocessed following evidence of ozone depletion in in situ observations. When the software was rerun without the flags, the ozone hole was seen as far back as 1976.

> https://en.wikipedia.org/wiki/ Ozone_depletion#Antarctic_ozone_hole



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Always always always always! Look at the data!

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> https://en.wikipedia.org/wiki/ Ozone_depletion#Antarctic_ozone_hole

String Similarity

Minimal number of edits (inserts, deletes, substitutions) needed to transform A into B.

$$egin{aligned} d_{i0} &= \sum_{k=1}^{i} w_{ ext{del}}(b_k), & ext{for } 1 \leq i \leq m \ d_{0j} &= \sum_{k=1}^{j} w_{ ext{ins}}(a_k), & ext{for } 1 \leq j \leq n \ d_{i-1,j-1} & ext{for } a_j = b_i \ d_{ij} &= egin{cases} d_{i-1,j-1} & ext{for } a_j = b_i \ d_{i,j-1} + w_{ ext{del}}(b_i) \ d_{i,j-1} + w_{ ext{sub}}(a_j) & ext{for } a_j
eq b_i \end{cases} & ext{for } 1 \leq i \leq m, 1 \leq j \leq n. \end{aligned}$$

https://en.wikipedia.org/wiki/Levenshtein_distance

115th Waterman St., Providence, RI 110th Waterman St., Providence, RI

Edit Distance = 1

Waterman Street, Providence, RI Waterman St, Providence, RI

Edit Distance = 4

Problems?

148th Ave NE, Redmond, WA 148th Ave NE, Redmond, WA

Edit Distance = 0



148th Ave NE, Redmond, WA 148th Ave NE, Redmond, WA

Edit Distance = 0



148th Ave NE, Redmond, WA 148th Ave NE, Redmond, WA

148th Ave NE, Redmond, WA NE 148th Ave, Redmond, WA

Edit Distance = 0

148th Ave NE, Redmond, WA 148th Ave NE, Redmond, WA

148th Ave NE, Redmond, WA NE 148th Ave, Redmond, WA

Edit Distance = 4

String Similarity: Jaccard Similarity

$$J(A,B) = \frac{|A \cap B|}{|A \cup B|}$$

148th Ave NE, Redmond, WA 140th Ave NE, Redmond, WA

148th Ave NE, Redmond, WA140th Ave NE, Redmond, WA

148th Ave NE, Redmond, WA140th Ave NE, Redmond, WA

Jaccard = 4 / 6 = .67

148th Ave NE, Redmond, WA NE 148th Ave, Redmond, WA

Jaccard = ???

148th Ave NE, Redmond, WA NE 148th Ave, Redmond, WA

Jaccard = 1

iPad Two 16GB WiFi White iPad 2nd generation 16GB WiFi White

```
What's the Jaccard Similarity?
(a) 3/8
(b) 4/11
(c) 4/7
```

iPad Two 16GB WiFi White iPad 2nd generation 16GB WiFi White

What's the Jaccard Similarity?
(a) 3/8
(b) 4/11

(c)) 4/7

#(iPad, 16GB, Wifi, White)

#(iPad, Two, 2nd, generation, 16GB, Wifi, White)

Michigan State University Michigan State Univ.

Michigan State University
Ohio State University

Jaccard = 0.5

Michigan State University
Michigan State Univ.



Michigan State University
Ohio State University

String Similarity: (Weighted) Jaccard Similarity

Jaccard = 0.5

Michigan 1 1

Michigan State University

Jaccard = 0.25

Michigan State University

Ohio State University

String Similarity: (Weighted) Jaccard Similarity

Jaccard = 0.5

Michigan 1 1

Michigan State University

Jaccard Michigan State University

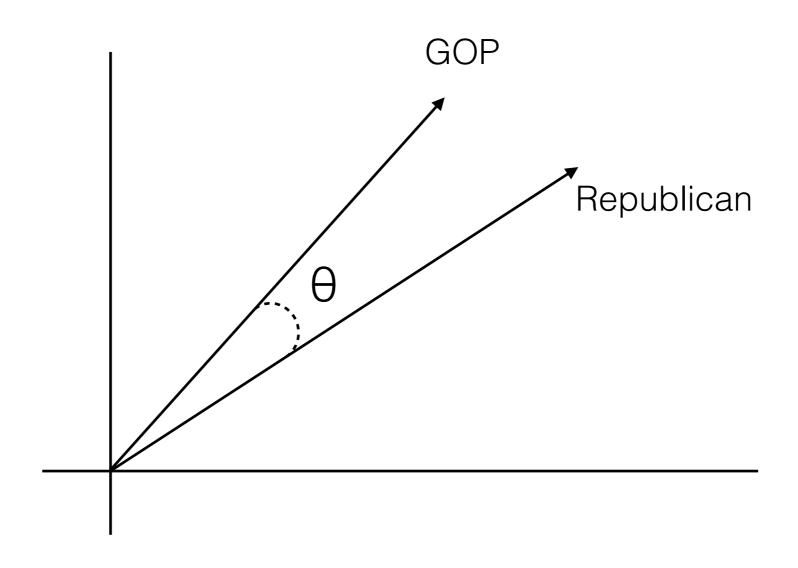
University of Michigan

String Similarity: Cosine Similarity

GOP Republican

Senator	Washington	announced	party	primary	chairman
1002	41	502	700	400	3
800	35	521	698	423	10

String Similarity: Cosine Similarity



Brown Brown Uni.

Which metric would (likely) consider the above words more similar?

- (a) Jaccard
- (b) Cosine

Brown Brown Uni.

Which metric would (likely) consider the above words more similar?



Motown Detroit

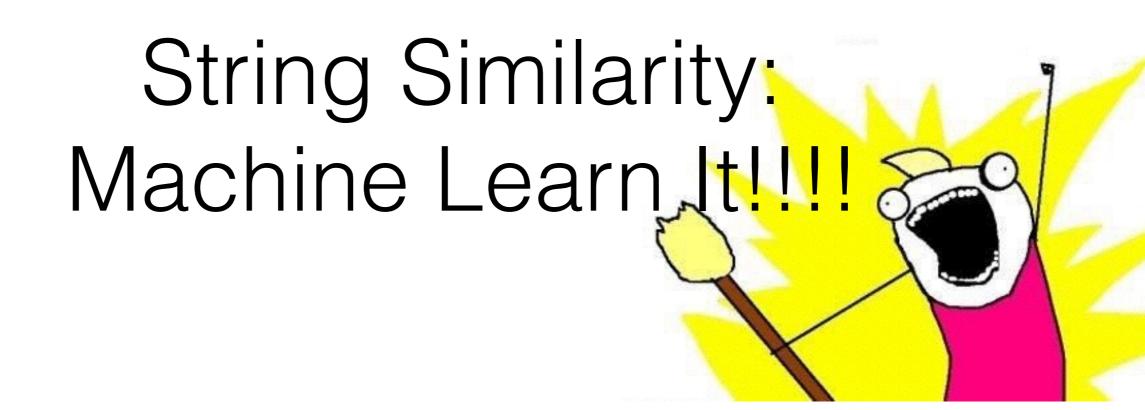
Which metric would (likely) consider the above words more similar?

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Motown Detroit

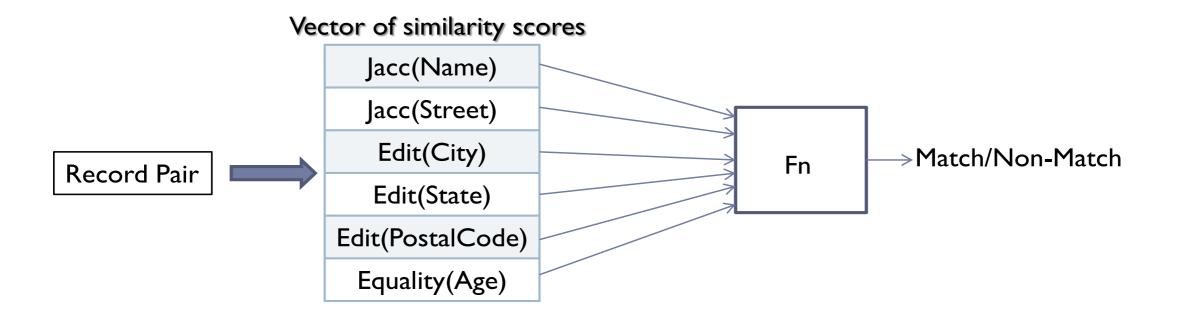
Which metric would (likely) consider the above words more similar?

(a) Jaccard (b) Cosine



Customer

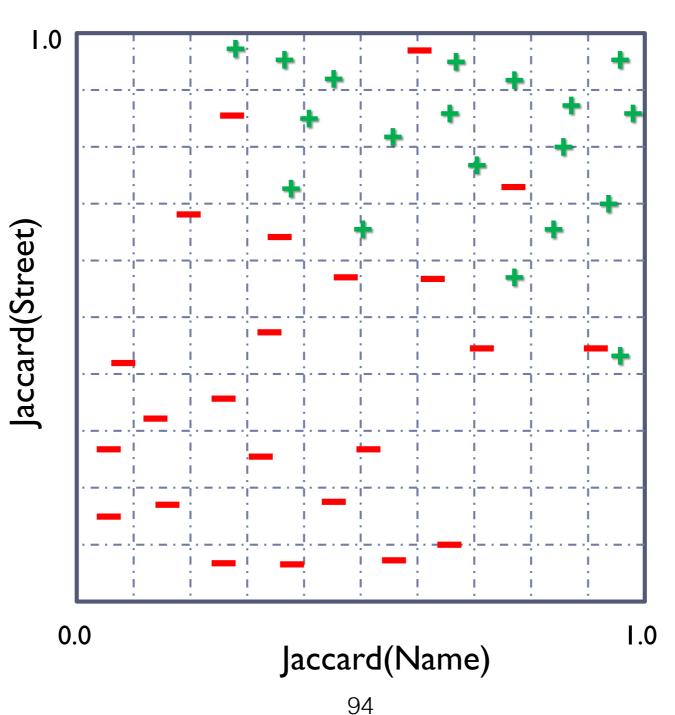
ld	Name	Street	City	State	P-Code	Age
I	J Smith	123 University Ave	Seattle	Washington	98106	42
2	Mary Jones	245 3rd St	Redmond	WA	98052-1234	30
3	Bob Wilson	345 Broadway	Seattle	Washington	98101	19
4	M Jones	245 Third Street	Redmond	NULL	98052	299
5	Robert Wilson	345 Broadway St	Seattle	WA	98101	19
6	James Smith	123 Univ Ave	Seatle	WA	NULL	41
7	J Widom	123 University Ave	Palo Alto	CA	94305	NULL
				•••	•••	

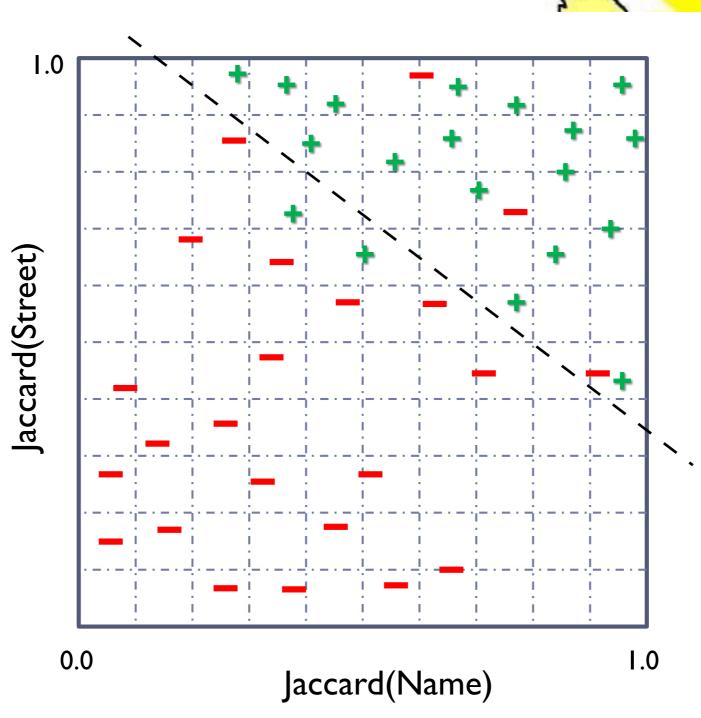


Features

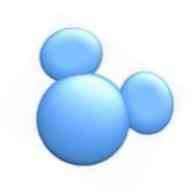
Binary Classification

Bob Wilson	345 Broadway	Seattle	Washington	98101	19	Match
Robert Wilson	345 Broadway St	Seattle	WA	98101	19	
BWilson	123 Broadway	Boise	Idaho	83712	19	Non-Match
Robert Wilson	345 Broadway St	Seattle	WA	98101	19	TAOTI-T Taccit
Mary Jones	245 3rd St	Redmond	WA	98052-1234	30	Match
M Jones	245 Third Street	Redmond	NULL	98052	299	I lacell
						_
Mary Jones	245 3rd St	Redmond	WA	98052-1234	30	Non-Match
Robert Wilson	345 Broadway St	Seattle	WA	98101	19	I NOII-I Iaccii





Code-along!



cat data.txt | cut -f 2,4 | sort | uniq -c | sort -nr | head

Check in

How comfortable would you say you are using the command line?

- (a) Not at all...I am actually not even sure what the question means.
- (b) I've used it for turnin and a few other things (e.g. copy-pasting to install things), but thats it
- (c) I can get around, but its not my "home base"
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Bash Scripting

https://cs.brown.edu/people/epavlick/articles.txt

1. ID

6. Victim Age

2. City

7. Shooter Age

3. State

8. Url

4. Date (YYYY-MM-DD)

9. Title

5. Time

10. Article Text

cat, less, head, tail

what does this data even look like?

```
# first 10 lines of file
$ head articles.txt

# first line of file
$ head -n 1 articles.txt

# random 10 lines from file
$ cat articles.txt | shuf | head
```

WC

how many articles are there

```
# how many bytes, words, and lines are
there?
$ wc articles.txt

# how many lines are there?
$ wc -l articles.txt
```

pipe (|), redirect (>)

```
$ head articles.txt | wc -l
      10
# write output to file called "tmp"
$ head articles.txt > tmp
$ wc -l tmp
      10 tmp
$ head articles.txt | wc -l > tmp
$ cat tmp
```

cut

```
$ cat articles.txt | cut -f 1 | head -n
3
Antioch
Greeley
Bridgeport
$ cat articles.txt | cut -f 3 | cut -f 1
-d '-' | head -n 3
2016
2015
2014
```

sort, uniq

```
# print the lowest 3 values (includes duplicates)
$ cat articles.txt | cut -f 4 | cut -f 1 -d '-'| sort | head -n 3
1929
1932
1932
# print lowest three values (remove duplicates but count how many
occurrances of each
$ cat articles.txt | cut -f 4 | cut -f 1 -d '-'| sort | uniq -c |
head -n 3
   1 1929
   2 1932
   3 1942
# find the most frequent years
$ cat articles.txt | cut -f 4 | cut -f 1 -d '-'| sort | uniq -c |
sort -r | head
5091 2015
1821 2016
1784 NA
```

sort, uniq

How many duplicated entries are there (using url as the uniq id)?

```
# total number of urls (lines)
$ cat articles.txt | cut -f 8 | wc -l
    9584

# number of unique urls
$ cat articles.txt | cut -f 8 | sort | uniq | wc -l
    7990

# number of duplicated urls
$ cat articles.txt | cut -f 8 | sort | uniq -d | wc -l
    981
```

regex (grep, sed, awk)

```
$ cat articles.txt | cut -f 2 | grep "NY" | head -n 5
NY
HOMINY
NYC
NY
NY
$ cat articles.txt | cut -f 2 | grep "^NY$" | head
NY
NY
NY
NY
$ cat articles.txt | cut -f 2 | grep "^NY[.]*" | head
NY
NYC
NY
NY
NY
```

regex (grep, sed, awk)

```
$ cat articles.txt | cut -f 4 | sed "s/
[0-9]/\#/g" | head -n 3
####-##-##
####-##-##
####-##-##
$ cat articles.txt | cut -f 3 | sed "s/
[A-Z][A-Z] - //g" \mid grep -v Unclear \mid
head -n 3
Minnesota
North Carolina
Michigan
```

Being all fancy...

```
# replace all non-numeric characters with blanks
$ cat articles.txt | cut -f 6 | sed "s/[^0-9]//g" |
head
# plot a histogram of all ages
cat articles.txt | cut -f 6 | sed "s/[^0-9]//g" |
grep -v "^$" | python2 -c "import sys,
matplotlib.pyplot as plt; plt.hist([int(i) for i in
sys.stdin]); plt.show()"
# plot a histogram of all ages, removing outliers
cat articles.txt | cut -f 6 | sed "s/[^0-9]//g" |
grep -v "^$" | python2 -c "import sys,
matplotlib.pyplot as plt; plt.hist([min(int(i), 100)
for i in sys.stdin]); plt.show()"
```

\$

This is funny because it is a regex joke. Please laugh and validate me. I will wait.

