

Tidying: The Preliminary First/Last Step(s) for Data Modeling?

Ravichandran Sarangan ABCC, FNCLR

## Announcement

Drs. Randy Johnson and Ravichandran Sarangan will be running a Data Tidying R-workshop soon. Hope you all can join us and Please look out for the announcement.

## Agenda

- What is Tidy data?
  - Definition
  - Examples

- How to tidy up the untidy data?
  - Many examples

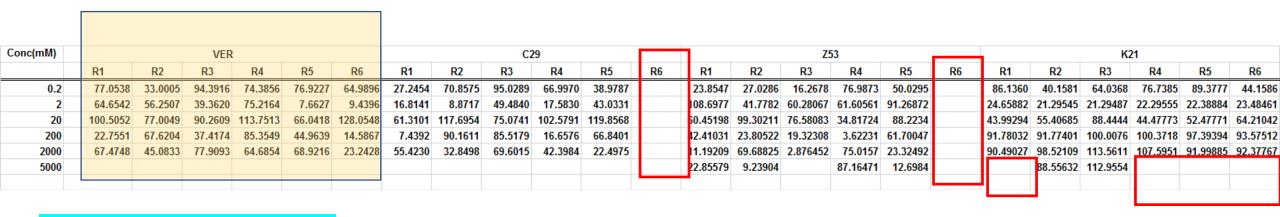
## Acknowledgement(s)

- Dr. Randy Johnson
- Statistics for Lunch team
- Ideas for the talk came from several places
  - Key resources are
    - Hadley Wickham (RStudio)
    - Karl Broaman (Univ Wisconsin)
    - Roger Peng, Brian Caffo, Jeff Leek (Johns Hopkins)
    - Rafael Irizarry (Harvard)

## Tidy Data Not For Us?

We don't model big data

#### **Excel Data as shared with us**



R1 to R6 are Replicates

**Cell Growth Data (two more drug data not shown)** 

Test the significance of difference between the drugs

#### R1 to R6 are Replicates

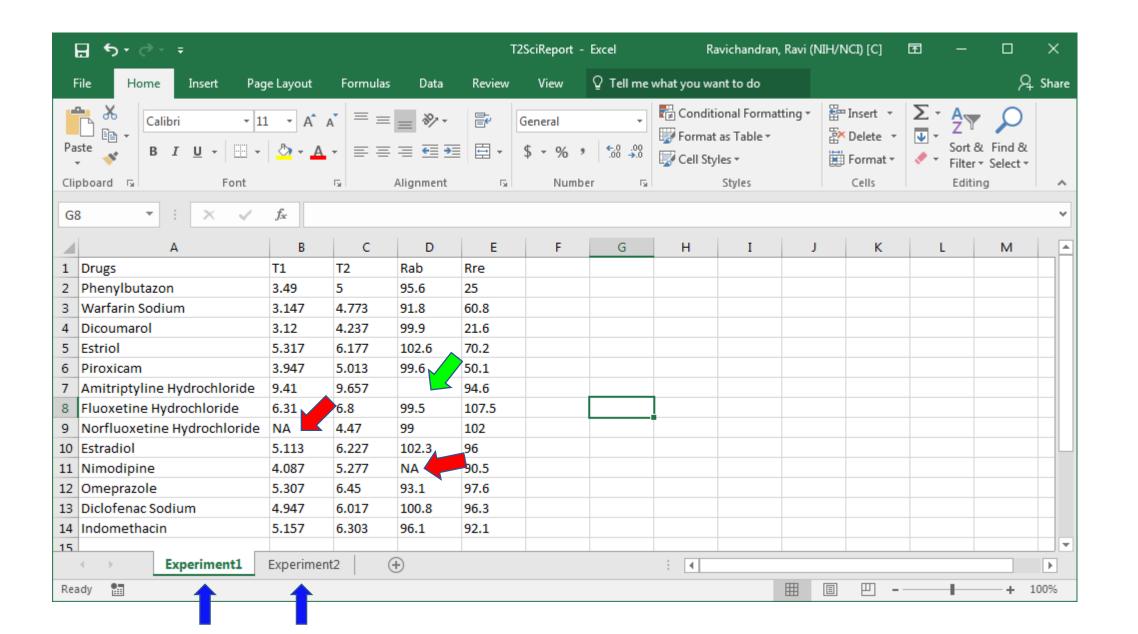
Concentration(mM)			VER			
	R1	R2	R3	R4	R5	R6
0.2	63.2915	68.6234	70.0896	38.9158	61.2060	80.7211
2	30.8031	59.2299	37.4290	79.4987	58.8359	28.4906
20	26.5052	108.7251	57.8343	51.9349	92.3215	114.4090
200	38.0781	64.5364	28.5504	54.0257	55.7642	17.1096
2000	37.1207	49.2985	30.0768	63.0939	28.1204	51.4099
5000						

		<b>Z</b> 5	3		
R1	R2	R3	R4	R5	R6
25.1243	71.3255	79.4123	20.7826	68.1928	
99.3483	28.27143	52.67069	108.1993	71.78687	
100.345	94.43867	97.50873	70.42895	88.2234	
27.76815	49.31593	16.73357	14.3842	68.24119	
14.208	68.5947	15.57453	16.63922	53.24721	
34.22645	84.47106		80.83111	21.71818	
	25.1243 99.3483 100.345 27.76815 14.208	25.1243 71.3255 99.3483 28.27143 100.345 94.43867 27.76815 49.31593	R1         R2         R3           25.1243         71.3255         79.4123           99.3483         28.27143         52.67069           100.345         94.43867         97.50873           27.76815         49.31593         16.73357           14.208         68.5947         15.57453	25.1243       71.3255       79.4123       20.7826         99.3483       28.27143       52.67069       108.1993         100.345       94.43867       97.50873       70.42895         27.76815       49.31593       16.73357       14.3842         14.208       68.5947       15.57453       16.63922	R1         R2         R3         R4         R5           25.1243         71.3255         79.4123         20.7826         68.1928           99.3483         28.27143         52.67069         108.1993         71.78687           100.345         94.43867         97.50873         70.42895         88.2234           27.76815         49.31593         16.73357         14.3842         68.24119           14.208         68.5947         15.57453         16.63922         53.24721

Concentration(mM)			C2	29		
	R1	R2	R3	R4	R5	R6
0.2	81.9251	25.9428	88.3891	46.5541	27.8573	
2	58.1008	78.8404	8.7778	5.7540	75.8824	
20	96.3247	57.1182	102.3849	101.1493	78.0180	
200	32.5744	36.5940	81.5093	14.7338	63.7707	
2000	73.5428	74.1092	74.0179	41.6010	48.7619	
5000						

Concentration(mM)			K2	21		
	R1	R2	R3	R4	R5	R6
0.2	77.2862	22.9933	26.2564	49.7822	73.4912	25.5989
2	24.65441	22.76161	22.14644	21.66432	21.87527	22.8385
20	62.09718	66.93179	88.4444	17.09395	70.17626	20.70322
200	91.78032	91.77401	100.0076	100.3718	97.39394	93.57512
2000	90.49027	98.52109	113.5611	107.5951	91.99885	92.37767
5000		88.55632	112.9554			

Conc(mM)	Drug	Rep	Cell-Growth(%)
0.2	VER	R1	63.2915
0.2	C29	R1	81.2951



Excel sheet with 10 Worksheets Raw someone sent you © No clear procedures Clean up (mostly missing data) Modify/Transform Modeling Visualize Data

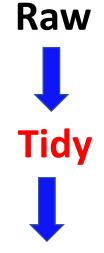
## Raw Data: Hospital Compare Data

#### Rectangular Data

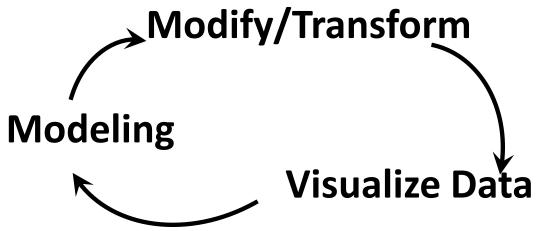
1056   MILWAUKEE   WI   53295   MILWAUKEE   Heart failure (HF) 30-Day R READM-30-HF   19.34   No different than the VHA   551   21.51   18.72   24.72   10.757   10	
1056 MILWAUKEE   WI   53295   MILWAUKEE   Heart failure (HF) 30-Day R READM-30-HF   19.34   No different than the VHA 551   21.51   18.72   24.72   10.57   10.57   10.57   10.58   MADISON   WI   53705   DANE   Death rate for chronic obst MORT-30-COPD   5.99   No different than the VHA 255   14.67   12.19   17.57   10.58   10.59   MADISON   WI   53705   DANE   Death rate for chronic obst MORT-30-COPD   5.99   No different than the VHA 170   4.65   2.81   7.15   10.50   10.50   MADISON   WI   53705   DANE   Death rate for chronic obst MORT-30-COPD   5.99   No different than the VHA 408   7.15   5.52   9.25   9.25   10.51   MADISON   WI   53705   DANE   Pneumonia (PN) 30-Day M MORT-30-HF   7.72   No different than the VHA 210   7.12   5.2   9.33   10.50   MADISON   WI   53705   DANE   Pneumonia (PN) 30-Day M MORT-30-NM   15.57   No different than the VHA 210   7.12   5.2   9.33   10.50   MADISON   WI   53705   DANE   Pneumonia (PN) 30-Day M MORT-30-COPD   16.28   No different than the VHA 185   17.9   14.27   22.65   10.50   MADISON   WI   53705   DANE   Rate of readmission for chi READM-30-COPD   16.28   No different than the VHA 185   17.9   14.27   22.65   10.50   MADISON   WI   53705   DANE   Heart failure (HF) 30-Day R READM-30-HF   19.34   No different than the VHA 460   19.78   17.27   22.99   10.50   MADISON   WI   53705   DANE   Heart failure (HF) 30-Day R READM-30-HF   19.34   No different than the VHA 460   19.78   17.27   22.99   10.50   MADISON   WI   53705   DANE   Heart failure (HF) 30-Day R READM-30-HF   19.34   No different than the VHA 460   19.78   17.27   22.99   10.50   MADISON   WI   53705   DANE   Heart failure (HF) 30-Day R READM-30-HF   19.34   No different than the VHA 460   19.78   17.27   22.99   10.50   MADISON   WI   53705   DANE   Heart failure (HF) 30-Day M MORT-30-AMI   19.34   No different than the VHA 460   19.78   17.27   19.51   10.50   MADISON   WI   53705   DANE   Heart failure (HF) 30-Day M MORT-30-AMI   19.34   No different than the VHA 460   19.78   17.27   19.51   10.50	Measure Start Date
1057 MILWAUKEE   WI   53295   MILWAUKEE   Pneumonia (PN) 30-Day R READM-30-PN   14.74   No different than the VHA   235   14.67   12.19   17.57   10   10   10   10   10   10   10   1	0/1/2012
1058 MADISON   WI   53705   DANE   Acute Myocardial Infarctio   MORT-30-AMI   9.21   No different than the VHA   142   7.8   5.56   10.28   10.28   10.28   10.29   MADISON   WI   53705   DANE   Death rate for chronic obst MORT-30-COPD   5.99   No different than the VHA   170   4.65   2.81   7.15   7.15   7.15   7.25   7.	0/1/2012
1059 MADISON   WI   53705   DANE   Death rate for chronic obst MORT-30-COPD   5.99   No different than the VHA   170   4.65   2.81   7.15   106   106   MADISON   WI   53705   DANE   Heart failure (HF) 30-Day N MORT-30-PN   8.64   No different than the VHA   210   7.12   5.2   9.33   9.33   106   106   MADISON   WI   53705   DANE   Pneumonia (PN) 30-Day M MORT-30-PN   8.64   No different than the VHA   210   7.12   5.2   9.33   9.33   9.00   106   MADISON   WI   53705   DANE   Acute Myocardial Infarctio   READM-30-COPD   16.28   No different than the VHA   156   14.76   12.14   17.34   9.30   106   MADISON   WI   53705   DANE   Rate of readmission for chi READM-30-COPD   16.28   No different than the VHA   185   17.9   14.27   22.65   9.25   106   MADISON   WI   53705   DANE   Heart failure (HF) 30-Day R READM-30-HF   19.34   No different than the VHA   460   19.78   17.27   22.99   9.00   107   106   MADISON   WI   53705   DANE   Pneumonia (PN) 30-Day R READM-30-PN   14.74   No different than the VHA   460   19.78   17.27   22.99   9.00   107	0/1/2012
1060 MADISON   WI   53705   DANE   Heart failure (HF) 30-Day N   MORT-30-HF   7.72   No different than the VHA   408   7.15   5.52   9.25   9.25   1061   MADISON   WI   53705   DANE   Pneumonia (PN) 30-Day M   MORT-30-PN   8.64   No different than the VHA   210   7.12   5.2   9.33   9.3	0/1/2012
1061 MADISON WI 53705 DANE Pneumonia (PN) 30-Day M MORT-30-PN 8.64 No different than the VHA 210 7.12 5.2 9.33 101 1062 MADISON WI 53705 DANE Acute Myocardial Infarctio READM-30-AMI 15.57 No different than the VHA 156 12.14 17.34 17.34 1063 MADISON WI 53705 DANE Rate of readmission for ch READM-30-COPD 16.28 No different than the VHA 185 17.9 14.27 22.65 1064 MADISON WI 53705 DANE Heart failure (HF) 30-Day R READM-30-HF 19.34 No different than the VHA 460 19.78 17.27 22.99 1065 MADISON WI 53705 DANE Pneumonia (PN) 30-Day R READM-30-PN 14.74 No different than the VHA 217 16.17 13.47 19.51 1066 CHEYENNE WY 82001 LARAMIE Acute Myocardial Infarctio MORT-30-AMI 9.21 Number of Cases Too Smal Not Available Not Ava	0/1/2012
MADISON   WI   53705   DANE   Acute Myocardial Infarctio   READM-30-AMI   15.57   No different than the VHA   156   14.76   12.14   17.34   17.34   1063   MADISON   WI   53705   DANE   Rate of readmission for change READM-30-COPD   16.28   No different than the VHA   185   17.9   14.27   22.65   1064   MADISON   WI   53705   DANE   Heart failure (HF) 30-Day R READM-30-HF   19.34   No different than the VHA   460   19.78   17.27   22.99   1065   MADISON   WI   53705   DANE   Pneumonia (PN) 30-Day R READM-30-PN   14.74   No different than the VHA   217   16.17   13.47   19.51   1066   CHEYENNE   WY   82001   LARAMIE   Acute Myocardial Infarctio   MORT-30-AMI   9.21   Number of Cases Too Small   Not Available   Not Available   Not Available   Not Available   Not Available   Not Available   1-The number of cases/paid   1069   CHEYENNE   WY   82001   LARAMIE   Heart failure (HF) 30-Day N MORT-30-PN   8.64   No different than the VHA   143   8.1   6.01   10.66   Not Available   N	0/1/2012
1063 MADISON   WI   53705   DANE   Rate of readmission for ch   READM-30-COPD   16.28   No different than the VHA   185   17.9   14.27   22.65   10.00   10.	0/1/2012
1064 MADISON WI 53705 DANE Heart failure (HF) 30-Day R READM-30-HF 19.34 No different than the VHA 460 19.78 17.27 22.99 100 100 100 100 100 100 100 100 100 1	0/1/2012
1065 MADISON WI 53705 DANE Pneumonia (PN) 30-Day Re READM-30-PN 14.74 No different than the VHA 217 16.17 13.47 19.51 10.00 10	0/1/2012
1066 CHEYENNE WY 82001 LARAMIE Acute Myocardial Infarctio MORT-30-AMI 9.21 Number of Cases Too Smal Not Available Not Available Not Available Not Available Not Available 1- The number of cases/pa10 1067 CHEYENNE WY 82001 LARAMIE Death rate for chronic obst MORT-30-COPD 5.99 No different than the VHA 102 6.08 3.78 10.21 1000 CHEYENNE WY 82001 LARAMIE Heart failure (HF) 30-Day N MORT-30-HF 7.72 No different than the VHA 102 6.08 3.78 10.21 10.22 10	0/1/2012
1067         CHEYENNE         WY         82001         LARAMIE         Death rate for chronic obst MORT-30-COPD         5.99         No different than the VHA         102         6.08         3.78         10.21	0/1/2012
1068         CHEYENNE         WY         82001         LARAMIE         Heart failure (HF) 30-Day N MORT-30-HF         7.72         No different than the VHA         77         7.52         5.18         10.42         <	)/1/2012
1069 CHEYENNE WY 82001 LARAMIE Pneumonia (PN) 30-Day M MORT-30-PN 8.64 No different than the VHA 143 8.1 6.01 10.66 10.66 10.000 CHEYENNE WY 82001 LARAMIE Acute Myocardial Infarction READM-30-AMI 15.57 Number of Cases Too Small Not Available Not Available Not Available Not Available 1 - The number of cases/pa 10.000 Not Available Not Av	0/1/2012
1070 CHEYENNE WY 82001 LARAMIE Acute Myocardial Infarctio READM-30-AMI 15.57 Number of Cases Too Small Not Available Not Available Not Available Not Available 1 - The number of cases/pa10	0/1/2012
	0/1/2012
1071 CHEVENINE NW 92001 LABAMIE Poto of roadmiresion for the PEADM 20 CORD 15 29 No different than the VIA 115	)/1/2012
10/1 CHETEININE WI 02/01 LANAIVIE Rate Of Teachinission for Chiracadivi-30-COPD 10.28 NO different than the VHA 115 13.4 11.00 19.39	0/1/2012
1072 CHEYENNE WY 82001 LARAMIE Heart failure (HF) 30-Day R READM-30-HF 19.34 No different than the VHA 77 18.74 15.12 22.63	0/1/2012
1073 CHEYENNE WY 82001 LARAMIE Pneumonia (PN) 30-Day Re READM-30-PN 14.74 No different than the VHA 147 15.01 12.52 18.32	0/1/2012
1074 SHERIDAN WY 82801 SHERIDAN Acute Myocardial Infarctio MORT-30-AMI 9.21 Number of Cases Too Smal Not Available Not Available Not Available Not Available 1 - The number of cases/pc 10	)/1/2012
1075 SHERIDAN WY 82801 SHERIDAN Death rate for chronic obst MORT-30-COPD 5.99 Number of Cases Too Smal Not Available Not Available Not Available Not Available 1 - The number of cases/pc 10	)/1/2012
1076 SHERIDAN WY 82801 SHERIDAN Heart failure (HF) 30-Day MORT-30-HF 7.72 Number of Cases Too Smal Not Available Not Available Not Available Not Available 1 - The number of cases/pc 10	)/1/2012
1077 SHERIDAN WY 82801 SHERIDAN Pneumonia (PN) 30-Day M MORT-30-PN 8.64 No different than the VHA 34 8.83 6.36 12.3	0/1/2012
1078 SHERIDAN WY 82801 SHERIDAN Acute Myocardial Infarctio READM-30-AMI 15.57 Number of Cases Too Small Not Available Not Available Not Available Not Available Not Available 1 - The number of cases/pa-10	)/1/2012
1079 SHERIDAN WY 82801 SHERIDAN Rate of readmission for ch READM-30-COPD 16.28 Number of Cases Too Smal Not Available Not Available Not Available Not Available 1 - The number of cases/pc 10	)/1/2012
1080 SHERIDAN WY 82801 SHERIDAN Heart failure (HF) 30-Day R READM-30-HF 19.34 Number of Cases Too Smal Not Available Not Available Not Available Not Available 1 - The number of cases/pa 10	)/1/2012
1081 SHERIDAN WY 82801 SHERIDAN Pneumonia (PN) 30-Day Re READM-30-PN 14.74 No different than the VHA 32 14.28 11.04 17.55 10	0/1/2012
1082 ORLANDO FL 32827 ORANGE Acute Myocardial Infarctio MORT-30-AMI 9.21 Not Available	)/1/2012
1083 ORLANDO FL 32827 ORANGE Death rate for chronic obst MORT-30-COPD 5.99 Not Available Not Availab	)/1/2012
1084 ORLANDO FL 32827 ORANGE Heart failure (HF) 30-Day N MORT-30-HF 7.72 Not Available	)/1/2012
r 1085 ORLANDO FL 32827 ORANGE Pneumonia (PN) 30-Day M MORT-30-PN 8.64 Not Available N	)/1/2012
1086 ORLANDO FL 32827 ORANGE Acute Myocardial Infarctio READM-30-AMI 15.57 Not Available Not Availab	)/1/2012
1087 ORLANDO FL 32827 ORANGE Rate of readmission for ch READM-30-COPD 16.28 Not Available Not Available Not Available Not Available Not Available 5 - Results are not available	)/1/2012
1088 ORLANDO FL 32827 ORANGE Heart failure (HF) 30-Day R READM-30-HF 19.34 Not Available Not Availab	)/1/2012
1089 ORLANDO FL 32827 ORANGE Pneumonia (PN) 30-Day Re READM-30-PN 14.74 Not Available	)/1/2012

NOT all Not Available data are same





Series of clearly defined steps, procedures



#### Raw Data: Starting from the basics

#### Examples of Raw Data

Census data; output from your equipment

#### Why are most data untidy?

- People collecting the data enter the data in a convenient way
- People collecting the data don't worry about variables and analysis

#### Raw Data: Starting from the basics

- What is wrong with Raw Data?
  - Almost 80% of data analysis is spent in cleaning the data (tidying the data)
  - Non-Messy, cleaned data is rare

- OK, let us clean it?
  - No proper procedure exits

#### Common Data Structure

Dataset contains Rows and Columns

Columns are labelled and Rows are labeled sometimes

# For clarity, let us look at a simple dataset

### Common Data Representations

NA: Not Available Data

Various other formats exist for missing data

	TreatmentA	TreatmentB
John Smith	NA	2
Jane Doe	16	22
Mary Johnson	3	1



	John Smith	Jane Doe	Mary Johnson
TreatmentA	NA	16	3
TreatmentB	2	11	2

## Data Semantics

#### Dataset?

 Collection of <u>values</u> (quantitative: numbers; qualitative: strings)

- Each value belongs to a <u>variable</u> and an <u>observation</u>
  - <u>Variable</u>: contains <u>all values</u> that measure the same attribute (treatmentA etc.) across units.

**Units: People or objects** 

• An <u>observation</u> contains <u>all values</u> measured on the same unit (person, John Smith etc.)

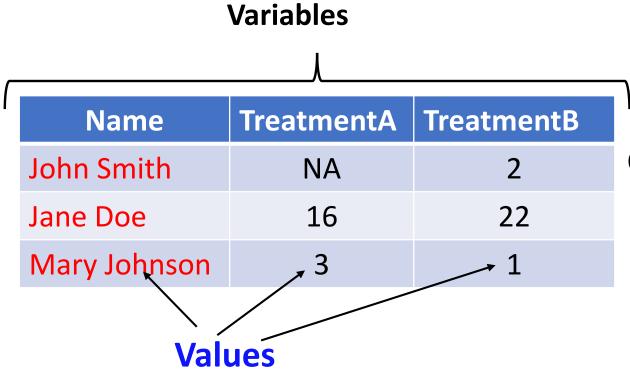
# Unstructured (or Untidy) to Tidy

## Two different presentation of a table

#### **Completely Crossed experiment**

	TreatmentA	TreatmentB
John Smith	NA	2
Jane Doe	16	22
Mary Johnson	3	1

### Two different presentation of a table



One treatment/person

TreatmentA/TreatmentB are not headers?!!
They are variables

#### **Observation 1**

#### Tidy

Name	Treatment	Result
John Doe	Α	4
Jane Doe	В	1
John Smith	Α	NA
John Smith	В	18
Mary Johnson	Α	6
Mary Johnson	В	7

## Untidy → Tidy

 Dataset contains 18 values representing 3 variables and 6 observations.

- Name: 3 possible values
  - John Smith, Jane Doe and Mary Johnson
- Treatment with 2 possible values
  - A or B

- Result with 6 possible values
  - NA, 16, 3, 2, 11 and 1

#### NA: Not Available

Name	Treatment	Result
John Doe	А	4
Jane Doe	В	1
John Smith	А	NA
John Smith	В	18
Mary Johnson	А	6
Mary Johnson	В	7

#### Issues

- What are observations/variables?
  - Sometimes difficult to identify

- Missing variables (we call NAs here)
  - Really missing or not possible

### Rules Tidy Data

1. Each column should represent a single variable (i.e., field)

2. Each row should represent a single observation (i.e., record)

3. Each cell should contain only one (single) value 

□ 20/120000

Interrelated relationship—Not possible to satisfy only two without satisfying the other one.

Treatment

A
B
A
B
A
B

Jane Doe B 1

### Tidy Rules

- Closely follows Codd's Third norm (3NF)
  - https://en.wikipedia.org/wiki/Third normal form#Definition of third normal form

## Code Book or Data documentation/dictionary

- Information about the variables
  - Units etc
- Study design information

- Additional assumptions
  - Created new groups (low-income, mid-income, high-income groups etc.)

## Code Book or Data documentation/dictionary

- What kind of study was this (prospective study)?
- Where was this carried out?
- When did the study began?
- Where was the study carried out?
- How many participants and other details.

Variable	Description	Units	Range or Count
SEX	Participants sex	1 = Male	n = 3500
		2 = Female	n = 3800
Age	Number of days since last exam		0-2300

## Untidy → Tidy

- ✓ Each cell contains single value
- ✓ Each variable forms a column
- X Each observation forms a row

	TreatmentA	TreatmentB
John Smith	NA	2
Jane Doe	16	22
Mary Johnson	3	1

Each cell contains single value Each variable forms a column Each observation forms a row

Name	Treatment	Result
Jane Doe	А	4
Jane Doe	В	1
John Smith	А	NA
John Smith	В	18
Mary Johnson	А	6
Mary Johnson	В	7

## More Examples of Tidying

### Your Turn: How will you tidy this data set?

#### **Unstructured or Untidy**

	Habitat		
Species	X	Y	Z
Α	0	3	0
В	1	0	2

#### **Structured or Tidy**

Species	Habitat	Abundance
Α	X	0
Α	Y	3
Α	Z	0
В	X	1
В	Y	0
В	Z	2

Based on Figure 1 from Baldridge et al, DOI: 10.4033/iee.2013.6b.6.f

#### Yes, all these are the same data

country	year	type	count
Afghanistan	1999	cases	745
Afghanistan	1999	population	19987071
Afghanistan	2000	cases	2666
Afghanistan	2000	population	20595360
Brazil	1999	cases	37737
Brazil	1999	population	172006362
Brazil	2000	cases	80488
Brazil	2000	population	174504898
China	1999	cases	212258
China	1999	population	1272915272
China	2000	cases	213766
China	2000	population	1280428583

Which one is tidy?

country	year	population	cases
Afghanistan	1999	19987071	745
Afghanistan	2000	20595360	2666
Brazil	1999	172006362	37737
Brazil	2000	174504898	80488
China	1999	1272915272	212258
China	2000	1280428583	213766

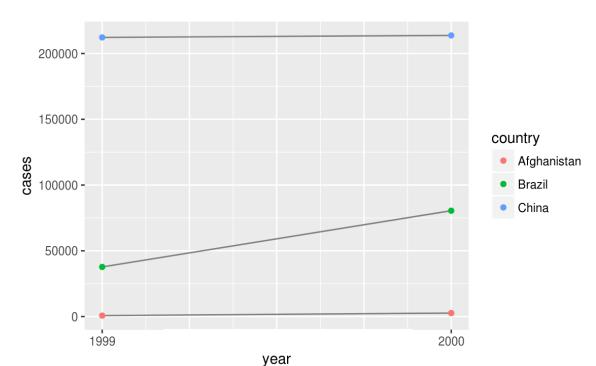
country	year	rate
Afghanistan	1999	745/19987071
Afghanistan	2000	2666/20595360
Brazil	1999	37737/172006362
Brazil	2000	80488/174504898
China	1999 2	12258/1272915272
China	2000 2	13766/1280428583

H. Wickam, Journal of Statistical Software, 4 (59), 1-23 (2014)

## Easy to Compute/Plot

#### rate = (cases/population) \* 10000

country	year	cases	population	rate
Afghanistan	1999	745	19987071	0.3727
Afghanistan	2000	2666	20595360	1.2945
Brazil	1999	37737	172006362	2.1939
Brazil	2000	80488	174504898	4.6124
China	1999	212258	1272915272	1.6675
China	2000	213766	1280428583	1.6695



country	year	population	cases
Afghanistan	1999	19987071	745
Afghanistan	2000	20595360	2666
Brazil	1999	172006362	37737
Brazil	2000	174504898	80488
China	1999	1272915272	212258
China	2000	1280428583	213766



n = # of cases

year	n
1999	250740
2000	296920

#### How can we tidy this table?

Each row should represent a single observation (i.e., record)

Drugs	T1 (min)	T2 (min)
Phenylbutazon	3.490	5.000
Warfarin Sodium	3.147	4.773
Dicoumarol	3.120	4.237
Estriol	5.317	6.177
Piroxicam	3.947	5.013
Amitriptyline Hydrochloride	9.410	9.657
Fluoxetine Hydrochloride	6.310	6.800
Norfluoxetine Hydrochloride	3.400	4.470
Estradiol	5.113	6.227
Nimodipine	4.087	5.277
Omeprazole	5.307	6.450
Diclofenac Sodium	4.947	6.017
Indomethacin	5.157	6.303

## Retention times of the Conventional HPLC column (T1) Online SPE system (T2)

How many variables are in this table?
More than one answer!!

Modified Table 2 from Sci Rep, 2015; 5, 14866, 10.1038/srep14866

Drugs	T1 (min)	T2 (min)
Phenylbutazon	3.490	5.000
Warfarin Sodium	3.147	4.773
Dicoumarol	3.120	4.237
Estriol	5.317	6.177
Piroxicam	3.947	5.013
Amitriptyline Hydrochloride	9.410	9.657
Fluoxetine Hydrochloride	6.310	6.800
Norfluoxetine Hydrochloride	3.400	4.470
Estradiol	5.113	6.227
Nimodipine	4.087	5.277
Omeprazole	5.307	6.450
Diclofenac Sodium	4.947	6.017
Indomethacin	5.157	6.303

## Retention times of the Conventional HPLC column (T1) Online SPE system (T2)

Drugs	Time(min)	values
Phenylbutazon	T1	3.49
Warfarin Sodium	T1	3.15
Dicoumarol	T1	3.12
Estriol	T1	5.32
Piroxicam	T1	3.95
Amitriptyline Hydrochlorid	e T1	9.41
Fluoxetine Hydrochloride	T1	6.31
Norfluoxetine Hydrochlorid	e T1	3.40
Estradiol	T1	5.11
Nimodipine	T1	4.09
# with 42 more rows		

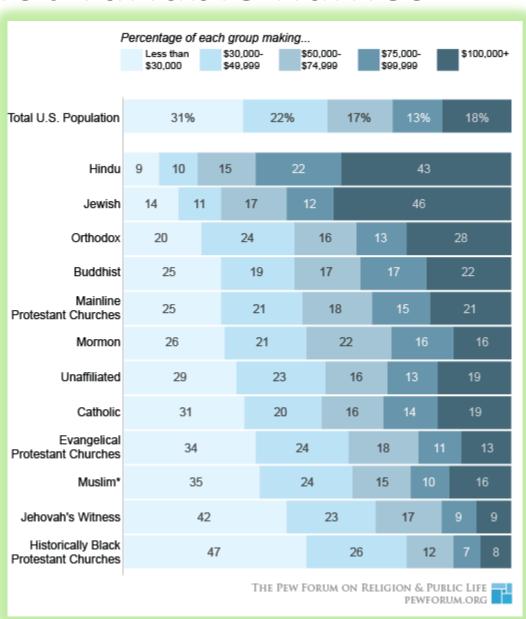
## Tidying Unstructured Datasets

Specific issues and how to tidy them

#### Column headers are values, not variable names

• Income distributions within U.S. Religious groups

Example based on H. Wickam, Journal of Statistical Software, 4 (59), 1-23 (2014)



# Column headers are values, not variable names

## Only first three lines of the dataset shown

Religion	<\$10k	<10-20k	\$20-30k	\$30-40k	\$40-50k	\$50-75k	\$75-100k	\$100-150k	>\$150k	Refused/Don't know
Agnostic	27	34	60	81	76	137	122	109	84	96
Atheist	12	27	37	52	35	70	73	59	74	76
Buddist	27	21	30	34	33	58	62	39	53	54

 Type of data representation is good for presentations but not good for analysis

How many variables are in this table?

## Only first three lines of the dataset shown

Religion	<\$10k	<10-20k	\$20-30k	\$30-40k	\$40-50k	\$50-75k	\$75-100k	\$100-150k	>\$150k	Refused/Don't know
Agnostic	27	34	60	81	76	137	122	109	84	96
Atheist	12	27	37	52	35	70	73	59	74	76
Buddist	27	21	30	34	33	58	62	39	53	54



To tidy this table, we have to stack (melt) the data. This means turning the **columns** into **rows**. Making WIDE datasets → Long or Tall

Religion	Income	freq
Agnostic	<\$10k	27
Agnostic	\$10-20k	34
Agnostic	\$20-30k	60
Agnostic	\$30-40k	81
Agnostic	\$40-50k	76
Agnostic	\$50-75k	137
Agnostic	\$75-100k	122
Agnostic	\$100-150k	109
Agnostic	>\$150k	84
Agnostic	Don't know/refused	96

## Multiple variables are stored in one column

country	year	m014	m1524	m2534	m3544	m4554	m5564	m65	mu	f014	f1524	f2534
AD	2000	0	0	1	0	0	0	0	NA	f014	NA	NA
AE	2000	2	4	4	6	5	12	10	NA	f014	16	1
AF	2000	52	228	183	149	129	94	80	NA	f014	414	565
AG	2000	0	0	0	0	0	0	1	NA	f014	1	1
AL	2000	2	19	21	14	24	19	16	NA	f014	11	10
AM	2000	2	152	130	131	63	26	21	NA	f014	24	27
AN	2000	0	0	1	2	0	0	0	NA	f014	0	1
AO	2000	186	999	1003	912	482	312	194	NA	f014	1142	1091
AR	2000	97	278	594	402	419	368	330	NA	f014	544	479
AS	2000	NA	NA	NA	NA	1	1	NA	NA	f014	NA	NA
AT	2000	1	17	30	59	42	23	41	NA	f014	11	22
AU	2000	3	16	35	25	24	19	49	NA	f014	15	19
AZ	2000	0	9	24	33	42	30	0	NA	f014	3	3
BA	2000	4	56	82	99	66	58	77	NA	f014	30	46
BB	2000	0	0	0	2	0	0	0	NA	f014	0	1
BD	2000	256	3640	5643	5750	4718	3667	2837	NA	f014	3029	3238

TB dataset (modified by Hadley); originally from WHO

m014: Male years 0-14

m1524: Male Years 15-24

Similar entries for Female.

The dataset is restricted for Year 2000

Note the zeros and NA Sometimes they mean different things in the dataset

How many variables are in this table?

## sex and age were under one column (also called column)

country	year	column	cases	country	year	sex	age	cases
$\overline{\mathrm{AD}}$	2000	m014	0	$\overline{\mathrm{AD}}$	2000	m	0–14	0
AD	2000	m1524	0	AD	2000	m	15 - 24	0
AD	2000	m2534	1	AD	2000	$\mathbf{m}$	25 - 34	1
AD	2000	m3544	0	AD	2000	$\mathbf{m}$	35 - 44	0
AD	2000	m4554	0	AD	2000	$\mathbf{m}$	45 - 54	0
AD	2000	m5564	0	AD	2000	$\mathbf{m}$	55 – 64	0
AD	2000	m65	0	AD	2000	$\mathbf{m}$	65 +	0
AE	2000	m014	2	AE	2000	$\mathbf{m}$	0 - 14	2
AE	2000	m1524	4	AE	2000	m	15 - 24	4
AE	2000	m2534	4	AE	2000	m	25 - 34	4
AE	2000	m3544	6	AE	2000	m	35 - 44	6
AE	2000	m4554	5	AE	2000	m	45 - 54	5
AE	2000	m5564	12	AE	2000	m	55 – 64	12
AE	2000	m65	10	AE	2000	$\mathbf{m}$	65 +	10
AE	2000	f014	3	AE	2000	f	0-14	3
	(a) Molt	en data			(b) T	Tidy da	ta	

country	year	m014	m1524	m2534	m3544	m4554	m5564	m65	mu	f014	f1524	f2534
AD	2000	0	0	1	0	0	0	0	NA	f014	NA	NA
AE	2000	2	4	4	6	5	12	10	NA	f014	16	1
AF	2000	52	228	183	149	129	94	80	NA	f014	414	565
AG	2000	0	0	0	0	0	0	1	NA	f014	1	1
AL	2000	2	19	21	14	24	19	16	NA	f014	11	10
AM	2000	2	152	130	131	63	26	21	NA	f014	24	27
AN	2000	0	0	1	2	0	0	0	NA	f014	0	1
AO	2000	186	999	1003	912	482	312	194	NA	f014	1142	1091
AR	2000	97	278	594	402	419	368	330	NA	f014	544	479
AS	2000	NA	NA	NA	NA	1	1	NA	NA	f014	NA	NA
AT	2000	1	17	30	59	42	23	41	NA	f014	11	22
AU	2000	3	16	35	25	24	19	49	NA	f014	15	19
AZ	2000	0	9	24	33	42	30	0	NA	f014	3	3
BA	2000	4	56	82	99	66	58	77	NA	f014	30	46
BB	2000	0	0	0	2	0	0	0	NA	f014	0	1
BD	2000	256	3640	5643	5750	4718	3667	2837	NA	f014	3029	3238

Note the final table is more meaningful

Country, Year, Sex, Age are all fixed

Tidy

Cases are unknown

$\operatorname{country}$	year	sex	age	cases
AD	2000	m	0–14	0
AD	2000	$\mathbf{m}$	15 - 24	0
AD	2000	$\mathbf{m}$	25 - 34	1
AD	2000	m	35 - 44	0
AD	2000	$\mathbf{m}$	45 - 54	0

Variables are stored in both rows and columns

 Daily temperatures in Cuernavaca, Mexico (2010)



Column Names: id, year,, month, element, d1-d31 (days 1-31)

TMIN/TMAX: Minimum/Maximum Temperature; ID: Weather station Identifier

NOT ALL NAs are the same:
Not all months have 31 days;
Can these <u>impossible</u> entries be removed?

			_	_																														
id	year mo	nth	element	d1	d2	d3	d4	d5	d6	d7	d8	d9	d10	d11	d12	d13	d14	d15	d16	d17	d18	d19	d20	d21	d22	d23	d24	d25	d26	d27	d28	d29	d30	d31
MX000017004	2010	l	TMAX	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	278	NA							
MX000017004	2010	1	TMIN	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	145	NA							
MX000017004	2010	2	TMAX	NA	273	241	NA	NA	NA	NA	NA	NA	NA	297	NA	299	NA																	
MX000017004	2010	2	TMIN	NA	144	144	NA	NA	NA	NA	NA	NA	NA	134	NA	107	NA																	
MX000017004	2010	3	TMAX	NA	NA	NA	NA	321	NA	NA	NA	NA	345	NA	NA	NA	NA	NA	311	NA														
MX000017004	2010	3	TMIN	NA	NA	NA	NA	142	NA	NA	NA	NA	168	NA	NA	NA	NA	NA	176	NA														
MX000017004	2010	1	TMAX	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	363	NA	NA	NA	NA							
MX000017004	2010	1	TMIN	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	167	NA	NA	NA	NA							
MX000017004	2010	5	TMAX	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	332	NA	NA	NA	NA							
MX000017004	2010	5	TMIN	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	182	NA	NA	NA	NA							
MX000017004	2010	5	TMAX	NA	NA	NA	NA	NA	NA	NA	NA	NA	280	NA	301	NA	NA																	
MX000017004	2010	5	TMIN	NA	NA	NA	NA	NA	NA	NA	NA	NA	175	NA	180	NA	NA																	
MX000017004	2010	7	TMAX	NA	NA	286	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	299	NA																
MX000017004	2010	7	TMIN	NA	NA	175	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	165	NA																
MX000017004	2010	3	TMAX	NA	NA	NA	NA	296	NA	NA	290	NA	NA	NA	NA	298	NA	264	NA	297	NA	NA	NA	280	NA	254								
MX000017004	2010	3	TMIN	NA	NA	NA	NA	158	NA	NA	173	NA	NA	NA	NA	165	NA	150	NA	156	NA	NA	NA	153	NA	154								
MX000017004	2010 1	0	TMAX	NA	NA	NA	NA	270	NA	281	NA	NA	NA	NA	NA	NA	295	287	NA	312	NA	NA	NA											
MX000017004	2010 1	0	TMIN	NA	NA	NA	NA	140	NA	129	NA	NA	NA	NA	NA	NA	130	105	NA	150	NA	NA	NA											
MX000017004	2010 1	1	TMAX	NA	313	NA	272	263	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	281	277	NA	NA	NA	NA
MX000017004	2010 1	1	TMIN	NA	163	NA	120	79	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	121	142	NA	NA	NA	NA
MX000017004	2010 1	2	TMAX	299	NA	NA	NA	NA	278	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
MX000017004	2010 1	2	TMIN	138	NΑ	NΑ	NΑ	NΑ	105	NΑ	NΑ	NΑ	NΑ	NΑ	NΑ	NΑ	NΑ	NΑ	NΑ	NΑ	NΑ	NΑ	NΑ	NΑ	NΑ	NΑ	NΑ	NΑ	NΑ	NΑ	NΑ	NΑ	NΑ	NΑ

How many variables are in this table?

id	yearn	nonth	element	d1	d2	d3	d4	d5	d6	d7	d8	d9	d10	d11	d12	d13	d14	d15	d16	d17	d18	d19	d20	d21	d22	d23	d24	d25	d26	d27	d28	d29	d30	d31
MX000017004	2010	1	TMAX	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	278	NA
MX000017004	2010	1	TMIN	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	145	NA
MX000017004	2010	2	TMAX	NA	273	241	NA	297	NA	299	NA																							



Tidy Data

1 MX000017004 2010 1 30 27 2 MX000017004 2010 2 2 27 3 MX000017004 2010 2 3 24 4 MX000017004 2010 2 11 29 5 MX000017004 2010 2 23 29 6 MX000017004 2010 3 5 32	73 144 11 144 17 134 19 107
3 MX000017004 2010 2 3 24 4 MX000017004 2010 2 11 29 5 MX000017004 2010 2 23 29	1 144 7 134 9 107
4 MX000017004 2010 2 11 29 5 MX000017004 2010 2 23 29	7 134 9 107
5 MX000017004 2010 2 23 29	9 107
6 MX000017004 2010 3 5 32	
	1 142
7 MX000017004 2010 3 10 34	5 168
8 MX000017004 2010 3 16 31	1 176
9 MX000017004 2010 4 27 36	3 167
10 MX000017004 2010 5 27 33	2 182
11 MX000017004 2010 6 17 28	0 175
12 MX000017004 2010 6 29 30	1 180
13 MX000017004 2010 7 3 28	6 175
14 MX000017004 2010 7 14 29	9 165
15 MX000017004 2010 8 5 29	6 158
16 MX000017004 2010 8 8 29	0 173
17 MX000017004 2010 8 13 29	8 165
18 MX000017004 2010 8 23 26	4 150
19 MX000017004 2010 8 25 29	7 156
20 MX000017004 2010 8 29 28	0 153
21 MX000017004 2010 8 31 25	4 154
22 MX000017004 2010 10 5 27	0 140
23 MX000017004 2010 10 7 28	1 129
24 MX000017004 2010 10 14 29	5 130
25 MX000017004 2010 10 15 28	7 105
26 MX000017004 2010 10 28 31	2 150
27 MX000017004 2010 11 2 31	.3 163
28 MX000017004 2010 11 4 27	2 120
29 MX000017004 2010 11 5 26	3 79
30 MX000017004 2010 11 26 28	1 121
31 MX000017004 2010 11 27 27	7 142
32 MX000017004 2010 12 1 29	9 138
33 MX000017004 2010 12 6 27	8 105

# Real-life example based on a question from a NCI investigator

#### Anonymized data

Maximal response for effect A 1000 nM	Maximal response for effect A at lower dose 100 nM	Maximal response for effect B 1000 nM	Maximal response for effect B at lower dose 100 nM
(Receptor1)	(Receptor1)	(Receptor2)	(Receptor2)
2.4459	0.9793	1.7922	1.7534
1.5752	0.0608	0.8562	-1.2930
2.6959	0.8674	2.0225	1.0184
1.2401	0.6337	3.7147	0.5151
4.5276	2.0354	2.9920	-0.8623
2.4625		0.9328	-0.7610
3.5255		1.6204	
2.3753			

Maximal responses were determined with an average +/- SEM

SEM: Standard deviation of the mean

- Two different biological responses each with two step dose response curve (Drug concentrations: 100nM and 1000nM)
- The 1000 nM drug concentration gives the maximum dose response for both effects
  - Note the absolute response is different
- Question?
  - If the 100 nM response for effect A is different from that of effect B

Tidy

### **Drug Concentration**

				Group
				A
				Α
				Α
				Α
Maximal response for	Maximal response for effect	Maximal response	Maximal response for effect	Α
effect A 1000 nM	A at lower dose 100 nM	for effect B 1000 nM	B at lower dose 100 nM	Α
(Receptor1)	(Receptor1)	(Receptor2)	(Receptor2)	A
2.4459	0.9793	1.7922	1.7534	A
1.5752	0.0608	0.8562	-1.2930	А А
2.6959	0.8674	2.0225	1.0184	Ā
1.2401	0.6337	3.7147	0.5151	A
4.5276	2.0354	2.9920	-0.8623	A
2.4625		0.9328	-0.7610	В
3.5255		1.6204		В
2.3753				В
	Untidy			В
	Circial			В
				В
				В
				В
				В
				В

	Drug concentration	
Group	(nM)	Response
Α	1000	2.4459
Α	1000	1.5752
Α	1000	2.6959
Α	1000	1.2401
Α	1000	4.5276
Α	1000	2.4625
Α	1000	3.5255
Α	1000	2.3753
Α	100	0.9793
Α	100	0.0608
Α	100	0.8674
Α	100	0.6337
Α	100	2.0354
В	1000	1.7922
В	1000	0.8562
В	1000	2.0225
В	1000	3.7147
В	1000	2.9920
В	1000	0.9328
В	1000	1.6204
В	100	1.7534
В	100	-1.2930
В	100	1.0184
В	100	0.5151
В	100	-0.8623
В	100	-0.7610

# Another fun example

## Total number of words spoken by each Race and Gender

#### **Movie 1**

The Fellowship Of The Ring

Race	Female	Male
Elf	1229	971
Hobbit	14	3644
Man	0	1995

#### **Movie 2**

The Two Towers

Race	Female	Male
Elf	331	513
Hobbit	0	2463
Man	401	3589

#### **Movie 3**

## The Return of the King

Race	Female	Male
Elf	183	510
Hobbit	2	2673
Man	268	2459

Example from Dr. Jenny Bryan, Univ of British Columbia

https://www.stat.ubc.ca/~jenny/

# Questions on Untidy dataset

Movie 1

The Fellowship Of The Ring

•	Good	Format	for	report??	?

Race	Female	Male
Elf	1229	971
Hobbit	14	3644
Man	0	1995

- Violates most of the Tidy rules, Column values are not really # of Females but the number of words spoken by Females
- If we have to compute the total number of words spoken by Male Elf (for example) from all the movies, we have to spend some time extracting data from tables. Do you think it is easy?
- Want to add more data, how easy is it?

# Let us tidy the dataset

**Analysis becomes very easy** 

All three tables were merged into one table

We are ready to answer questions like:

What race spoke most words in a movie?

How different is the dominant race in "The Two Towers" compared to the "The Return of the King"?

Example from Dr. Jenny Bryan, Univ of British Columbia

https://www.stat.ubc.ca/~jenny/

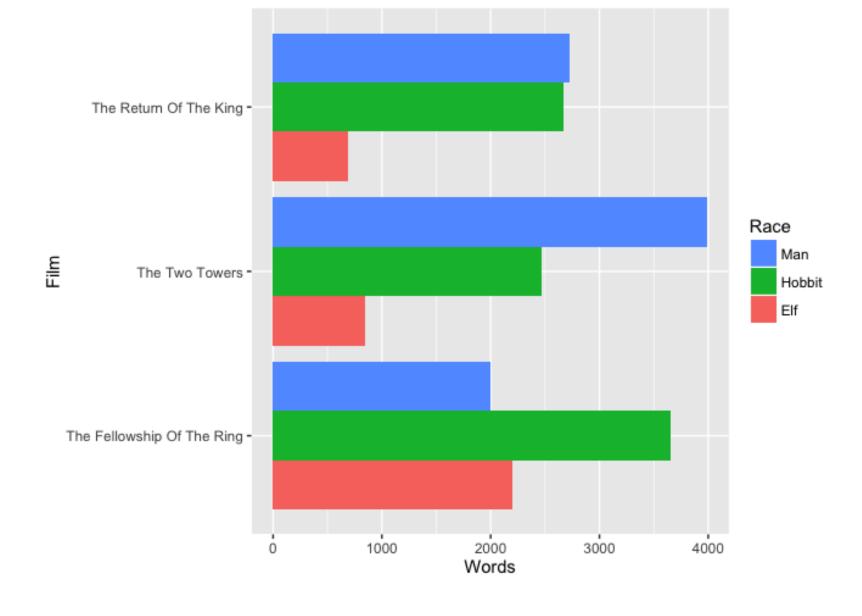




Film	Gender	Race	Words
The Fellowship Of The Ring	Female	Elf	1229
The Fellowship Of The Ring	Male	Elf	971
The Fellowship Of The Ring	Female	Hobbit	14
The Fellowship Of The Ring	Male	Hobbit	3644
The Fellowship Of The Ring	Female	Man	0
The Fellowship Of The Ring	Male	Man	1995
The Two Towers	Female	Elf	331
The Two Towers	Male	Elf	513
The Two Towers	Female	Hobbit	0
The Two Towers	Male	Hobbit	2463
The Two Towers	Female	Man	401
The Two Towers	Male	Man	3589
The Return Of The King	Female	Elf	183

Only top 13 rows of the tidy dataset shown

Final benefit is the ease of analysis and creating figures



Example from Dr. Jenny Bryan, Univ of British Columbia <a href="https://www.stat.ubc.ca/~jenny/">https://www.stat.ubc.ca/~jenny/</a>

# What about Biological (expression) data?

## High dimensional RNASeq data 64102 x 8; GSE52778

## **SAMPLES: COL; Features: ROWS**

Genes	Assay1	Assay2	Assay3	Assay4	Assay5	Assay6	Assay7	Assay8
<b>ESISGN6</b> 000000003	679	448	873	408	1138	1047	770	572
TNMD	0	0	0	0	0	0	0	0
<b>ENRS/GI</b> 000000000419	467	515	621	365	587	799	417	508
SCYL3	260	211	263	164	245	331	233	229
<b>ENS £002</b> 00000460	60	55	40	35	78	63	76	60
FGR	0	0	2	0	1	0	0	0
ENBG00000000971	3251	3679	6177	4252	6721	11027	5176	7995
FUCA2	1433	1062	1733	881	1424	1439	1359	1109
ENCSG00000001084	519	380	595	493	820	714	696	704
NFYA	394	236	464	175	658	584	360	269
<b>5NISG0</b> 0000001460	172	168	264	118	241	210	155	177
NIPAL3	2112	1867	5137	2657	2735	2751	2467	2905
<b>ENS(</b> 00000001497	524	488	638	357	676	806	493	475
ENPP4	71	51	211	156	23	38	134	172
<b>5NS/2900</b> 00001617	555	394	905	415	727	697	618	599
CFTR	10	2	9	2	10	6	5	5
<b>ENNS/CEDO</b> 000001629	1660	1251	2259	1079	2462	2514	1888	1660

High Dimensional Data (8 x 64,102)

Summarized Experiment: RNA-Seq experiment of read counts per gene for airway smooth muscles 8 different experimental and assays 64,102 gene transcripts.

Metadata

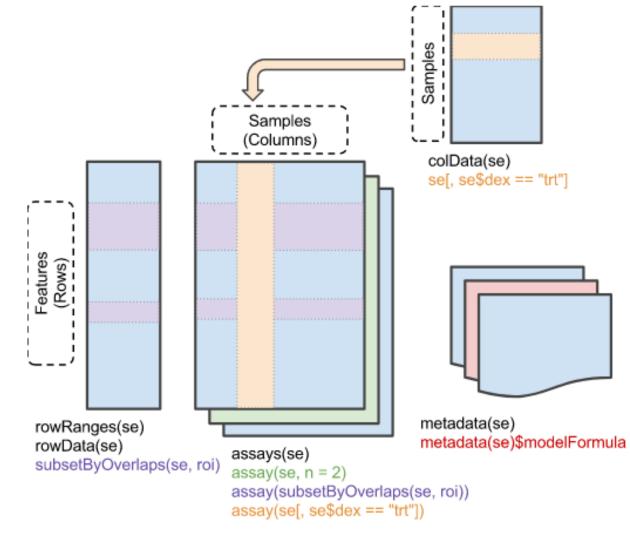
OPEN @ ACCESS Freely available online



# RNA-Seq Transcriptome Profiling Identifies *CRISPLD2* as a Glucocorticoid Responsive Gene that Modulates Cytokine Function in Airway Smooth Muscle Cells

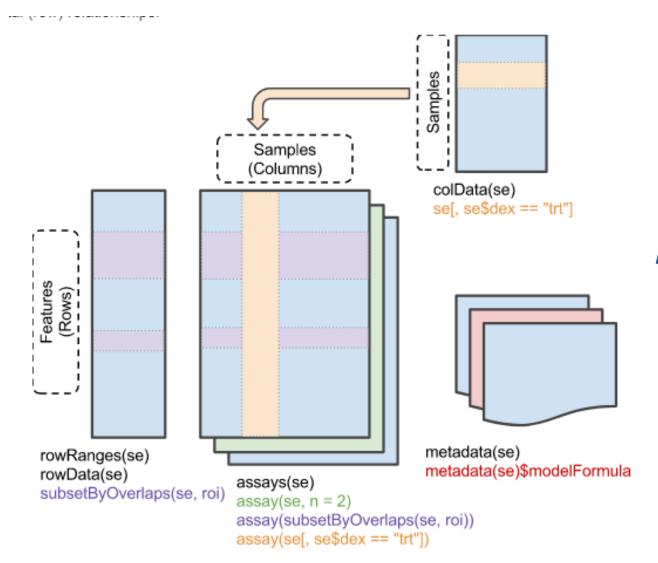
Blanca E. Himes<sup>1,2,3</sup>\*, Xiaofeng Jiang<sup>4</sup>\*, Peter Wagner<sup>4</sup>, Ruoxi Hu<sup>4</sup>, Qiyu Wang<sup>4</sup>, Barbara Klanderman<sup>2</sup>, Reid M. Whitaker<sup>1</sup>, Qingling Duan<sup>1</sup>, Jessica Lasky-Su<sup>1</sup>, Christina Nikolos<sup>5</sup>, William Jester<sup>5</sup>, Martin Johnson<sup>5</sup>, Reynold A. Panettieri Jr.<sup>5</sup>, Kelan G. Tantisira<sup>1</sup>, Scott T. Weiss<sup>1,2</sup>, Quan Lu<sup>4</sup>\*

1 Channing Division of Network Medicine, Brigham and Women's Hospital and Harvard Medical School, Boston, Massachusetts, United States of America, 2 Partners HealthCare Personalized Medicine, Boston, Massachusetts, United States of America, 3 Children's Hospital Informatics Program, Boston, Massachusetts, United States of America, 4 Program in Molecular and Integrative Physiological Sciences, Departments of Environmental Health, and Genetics and Complex Diseases, Harvard School of Public Health, Boston, Massachusetts, United States of America, 5 Pulmonary, Allergy and Critical Care Division, University of Pennsylvania, Philadelphia, Pennsylvania, United States of America

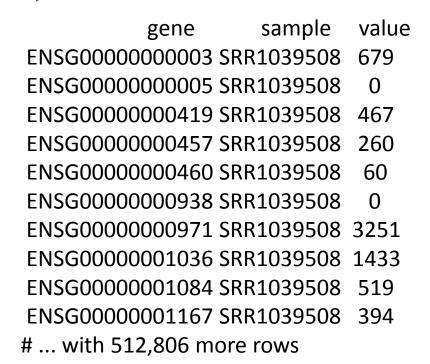


Summarized Experiment

Biology Data; Matrix form; not easily Raw cleanable **Untidy ►**Modify/Transform Modeling **Tidy Models Visualize Data Visualize Data** 



Summarized Experiment



# Summary/Final Thoughts

- Sharing Data
  - Send CSV or text files along with CODE Book
  - Excel: One sheet only
- Tidying removes redundancy
- Easy for analysis
- Reduces chances of error
- May not be suitable for some datasets
  - Biological expression
    - For now use BioConductor ©

# Thanks

ravichandrans@mail.nih.gov

## **Excel Data shared with us**

Conc(mM)			VER				C29								Z!	i3					K2	21		
	R1	R2	R3	R4	R5	R6	R1	R2	R3	R4	R5	R6	R1	R2	R3	R4	R5	R6	R1	R2	R3	R4	R5	R6
0.2	77.0538	33.0005	94.3916	74.3856	76.9227	64.9896	27.2454	70.8575	95.0289	66.9970	38.9787		23.8547	27.0286	16.2678	76.9873	50.0295		86.1360	40.1581	64.0368	76.7385	89.3777	44.1586
2	64.6542	56.2507	39.3620	75.2164	7.6627	9.4396	16.8141	8.8717	49.4840	17.5830	43.0331		108.6977	41.7782	60.28067	61.60561	91.26872		24.65882	21.29545	21.29487	22.29555	22.38884	23.48461
20	100.5052	77.0049	90.2609	113.7513	66.0418	128.0548	61.3101	117.6954	75.0741	102.5791	119.8568		60.45198	99.30211	76.58083	34.81724	88.2234		43.99294	55.40685	88.4444	44.47773	52.47771	64.21042
200	22.7551	67.6204	37.4174	85.3549	44.9639	14.5867	7.4392	90.1611	85.5179	16.6576	66.8401		42.41031	23.80522	19.32308	3.62231	61.70047		91.78032	91.77401	100.0076	100.3718	97.39394	93.57512
2000	67.4748	45.0833	77.9093	64.6854	68.9216	23.2428	55.4230	32.8498	69.6015	42.3984	22.4975		11.19209	69.68825	2.876452	75.0157	23.32492		90.49027	98.52109	113.5611	107.5951	91.99885	92.37767
5000													22.85579	9.23904		87.16471	12.6984			88.55632	112.9554			

**Cell Growth anonymized Data (two more drug data not shown)** 

Test the significance of difference between the drugs

## Example-1

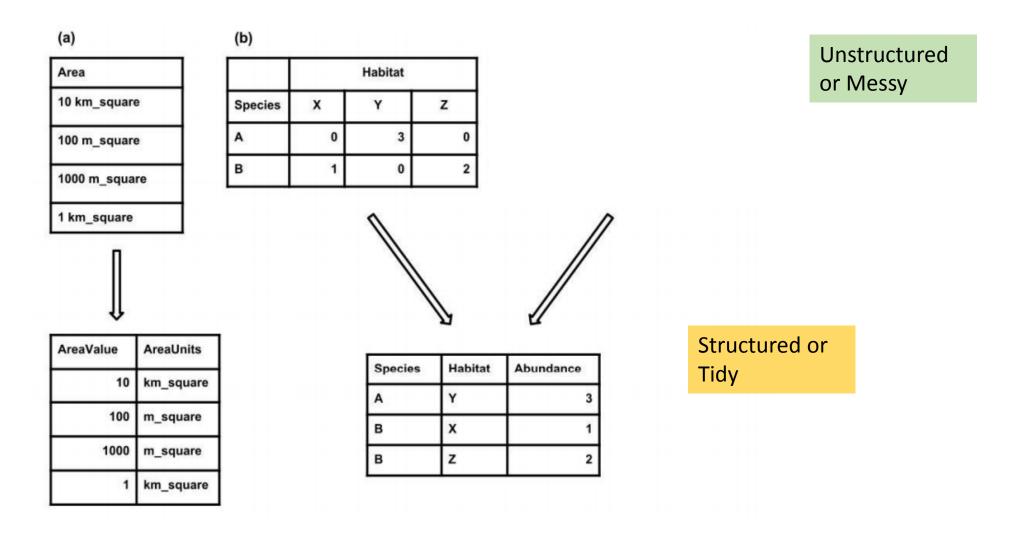


Figure 1 from Baldridge et al, DOI: 10.4033/iee.2013.6b.6.f

# Statistical significance between the effect of different drugs

## **Cell Growth Data (two more drugs not shown)**

Concentration(mM)			VER						C2	9					<b>Z</b> 5	3					K2	21		
	R1	R2	R3	R4	R5	R6	R1	R2	R3	R4	R5	R6	R1	R2	R3	R4	R5	R6	R1	R2	R3	R4	R5	R6
0.2	73.5673	110.3643	112.2544	90.2959	60.8799	98.7683	81.9251	25.9428	88.3891	46.5541	27.8573		67.4644	72.2806	44.5274	39.4477	38.8015		16.2846	88.0884	64.5890	35.9966	22.7104	18.4903
2	49.2198	18.5134	60.5415	56.6634	74.2664	66.4155	58.1008	78.8404	8.7778	5.7540	75.8824		30.85777	104.2153	95.54326	67.81979	116.3274		24.83905	22.57718	23.91723	23.94967	22.24302	22.52349
20	76.8248	66.1132	125.3349	53.8874	45.9427	107.4450	96.3247	57.1182	102.3849	101.1493	78.0180		26.08528	63.37327	40.41485	90.57895	88.2234		36.85725	43.24039	88.4444	65.62804	57.92811	28.01133
200	87.3209	97.7594	14.1516	9.9561	85.1224	57.9762	32.5744	36.5940	81.5093	14.7338	63.7707		63.44407	44.66939	36.71621	7.223918	35.65689		91.78032	91.77401	100.0076	100.3718	97.39394	93.57512
2000	31.3919	69.1952	29.3735	43.9584	43.9168	49.0450	73.5428	74.1092	74.0179	41.6010	48.7619		70.33274	7.996453	73.88588	34.52884	81.92558		90.49027	98.52109	113.5611	107.5951	91.99885	92.37767
5000													14.08076	29.51886		81.69069	64.66943			88.55632	112.9554			

Concentration(mM)			VER			
	R1	R2	R3	R4	R5	R6
0.2	63.2915	68.6234	70.0896	38.9158	61.2060	80.7211
2	30.8031	59.2299	37.4290	79.4987	58.8359	28.4906
20	26.5052	108.7251	57.8343	51.9349	92.3215	114.4090
200	38.0781	64.5364	28.5504	54.0257	55.7642	17.1096
2000	37.1207	49.2985	30.0768	63.0939	28.1204	51.4099
5000						

Concentration(mM)			Z5	3		
	R1	R2	R3	R4	R5	R6
0.2	25.1243	71.3255	79.4123	20.7826	68.1928	
2	99.3483	28.27143	52.67069	108.1993	71.78687	
20	100.345	94.43867	97.50873	70.42895	88.2234	
200	27.76815	49.31593	16.73357	14.3842	68.24119	
2000	14.208	68.5947	15.57453	16.63922	53.24721	
5000	34.22645	84.47106		80.83111	21.71818	

Concentration(mM)	C29					
	R1	R2	R3	R4	R5	R6
0.2	81.9251	25.9428	88.3891	46.5541	27.8573	
2	58.1008	78.8404	8.7778	5.7540	75.8824	
20	96.3247	57.1182	102.3849	101.1493	78.0180	
200	32.5744	36.5940	81.5093	14.7338	63.7707	
2000	73.5428	74.1092	74.0179	41.6010	48.7619	
5000						

Concentration(mM)	K21					
	R1	R2	R3	R4	R5	R6
0.2	77.2862	22.9933	26.2564	49.7822	73.4912	25.5989
2	24.65441	22.76161	22.14644	21.66432	21.87527	22.8385
20	62.09718	66.93179	88.4444	17.09395	70.17626	20.70322
200	91.78032	91.77401	100.0076	100.3718	97.39394	93.57512
2000	90.49027	98.52109	113.5611	107.5951	91.99885	92.37767
5000		88.55632	112.9554			

# Semantics and beyond

Experimental design gives us clues on the structure of observations

Here every combination of PERSON and TREATMENT was measured

- Identifying what are variables and what are observations is sometimes not trivial!!
  - Height, width
    - Could assign them to one variable, dimension

	TreatmentA	TreatmentB
John Smith	NA	2
Jane Doe	16	22
Mary Johnson	3	1

# Untidy → Tidy

 Dataset contains 18 values representing 3 variables and 6 observations.

- Name: 3 possible values
  - John Smith, Jane Doe and Mary Johnson
- Treatment with 2 possible values
  - A or B
- Result with 6 possible values
  - NA, 16, 3, 2, 11 and 1

	TreatmentA	TreatmentB
John Smith	NA	2
Jane Doe	16	22
Mary Johnson	3	1

Name	Treatment	Result
John Doe	A	4
Jane Doe	В	1
John Smith	A	NA
John Smith	В	18
Mary Johnson	A	6
Mary Johnson	В	7

# Untidy → Tidy

	TreatmentA	TreatmentB
John Smith	NA	2
Jane Doe	16	22
Mary Johnson	3	1

Name	Treatment	Result
John Doe	A	4
Jane Doe	В	1
John Smith	A	NA
John Smith	В	18
Mary Johnson	A	6
Mary Johnson	В	7

# What is tidy data?

## Rules for tidy data

- 1. Each column should represent a single variable (i.e., field)
- Each row should represent a single observation (i.e., record)
- Each type of observational unit forms a table
  - ➤ Height and weight measured on people (2 columns)
- 4. Each cell should contain only one (single) value
  - >500 mg (separate 500 and mg into two separate columns)

Different columns contain same

measurement (violation of Rule 3)

Handed- ness Gender	Right handed	Left handed	Total
Male	43	9	52
Female	44	4	48
Total	87	13	100

CODD's norm

Codd EF (1990). The Relational Model for Database Management: Version 2. Addison-Wesley Longman Publishing, Boston.

Taken from Wikipedia (contingency tables)

## Each observation forms a row

Name	Treatment	Result
Jane <del> Doe</del>	A	4
Jane <del> Doe</del>	В	1
John <del> Smith</del>	A	NA
John <del>'Smith</del>	В	18
Mar <del>y Johnson</del>	A	6
Mar <del>y Johnson</del>	В	<del>7</del>

## Each variable forms a column

Nam	e	Tre	eatment	Resu	llt
Jane	Doe	A		4	
Jane	Doe	В		1	
John	Smith	Α		NA	
John	Smith	В		18	
Mary	/ Johnson	Α		6	
Mary	Johnson	В	,	7	7

VALUES:

Name	Treatment	Result
Janoe	A	4
Janope	В	1
Johnmith	A	N
Johnith	В	1
Marohnson	A	6
Mar phnson	В	7

## Each observation forms a row

Name	Treatment	Result
Jane <del> Doe</del>	A	4
Jane <del> Doe</del>	В	1
John <del> Smith</del>	A	NA
John <del>'Smith</del>	В	18
Mar <del>y Johnson</del>	A	6
Mar <del>y Johnson</del>	В	<del>7</del> →

Each cell contains single value

VALUES:

## Each variable forms a column

Name		Treatment		Resu	llt
Jane	Doe	A	1	4	
Jane	Doe	В		1	
John	Smith	Α		NA	
John	Smith	В		18	
Mar	/ Johnson	Α		6	
Mar	Johnson	В		7	

Name	Treatment	Result
Janoe	A	4
Janepe	В	1
Johnmith	A	N
Johnith	В	1
Marohnson	A	6
Mar phnson	В	7

## Yes, all are the same data

### Entries in column are NOT the same

year	type	count
1999	cases	745
1999	population	19987071
2000	cases	2666
2000	population	20595360
1999	cases	37737
1999	population	172006362
2000	cases	80488
2000	population	174504898
1999	cases	212258
1999	population	1272915272
2000	cases	213766
2000	population	1280428583
	1999 1999 2000 2000 1999 1999 2000 2000	1999 cases 1999 population 2000 cases 2000 population 1999 cases 1999 population 2000 cases 2000 population 1999 cases 1999 population 1999 cases 1999 population 2000 cases

## Each cell NOT single valued

country	year	rate
Afghanistan	1999	745/19987071
Afghanistan	2000	2666/20595360
Brazil	1999	37737/172006362
Brazil	2000	80488/174504898
China	1999 2	212258/1272915272
China	2000 2	213766/1280428583



country	year	cases	population
Afghanistan	1999	745	19987071
Afghanistan	2000	2666	20595360
Brazil	1999	37737	172006362
Brazil	2000	80488	174504898
China	1999	212258	1272915272
China	2000	213766	1280428583

#### Entries in column are NOT the same

country	1999	2000
Afghanistan	745	2666
Brazil	37737	80488
China	212258	213766

#### Entries in column are NOT the same

country	1999	2000
Afghanistan	19987071	20595360
Brazil	172006362	174504898
China	1272915272	1280428583

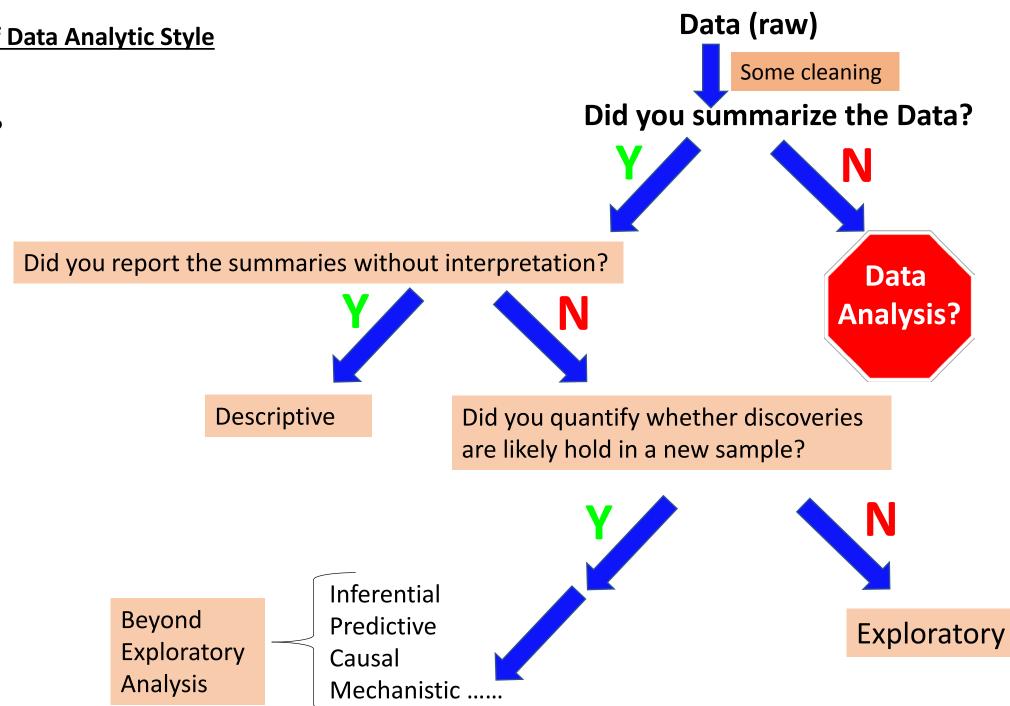
# Basics of Tidying comes from Codd's law

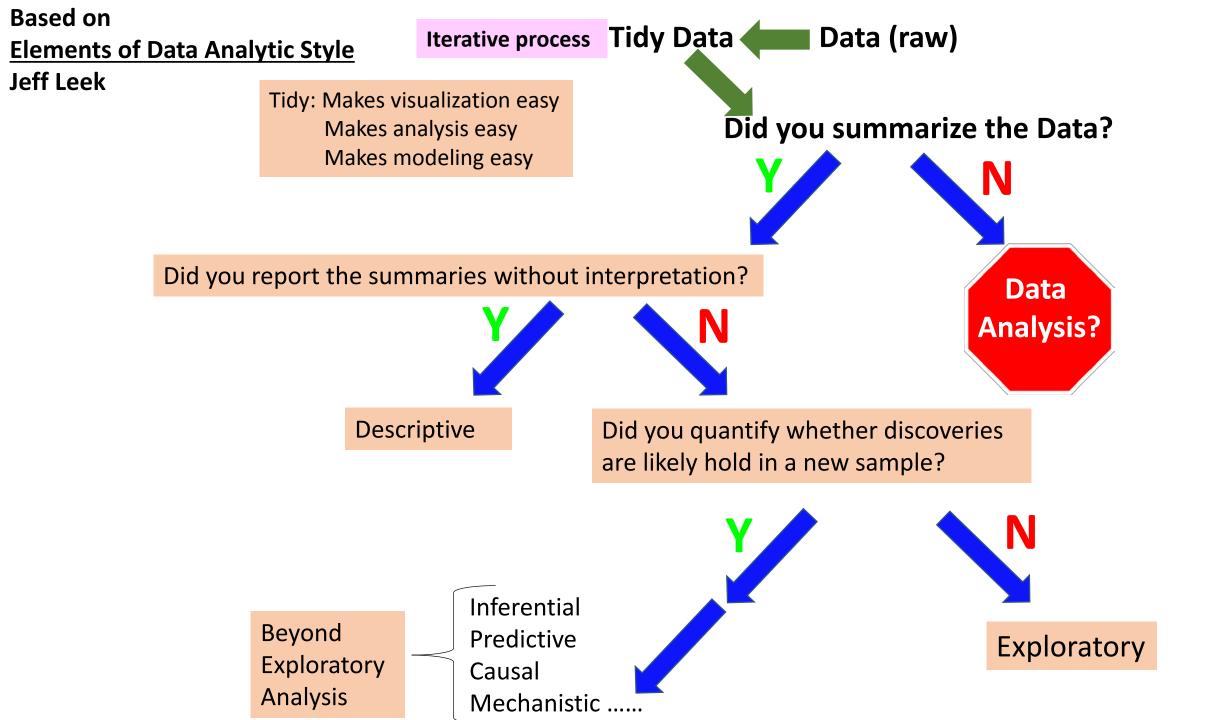
- Normalization
  - Process of organizing data in a database
    - Means creating tables and establishing relationship between tables

- Normalization leads to remove redundancy and easy maintenance
  - Ex. Person ID/name will be stored in only one place in a tidy data
  - Less effort to change if we had made mistakes

Based on **Elements of Data Analytic Style Jeff Leek** 

**CLEANUP** 





# Your Turn: How will you tidy this data set?

**Unstructured or Untidy** 

Area

10 km\_square

100 m\_square

1000 m\_square

1 km\_square

**Structured or Tidy?** 

Area	Units
10	Km_square
100	m_square
1000	m_square
1	km_square

Based on Figure 1 from Baldridge et al, DOI: 10.4033/iee.2013.6b.6.f