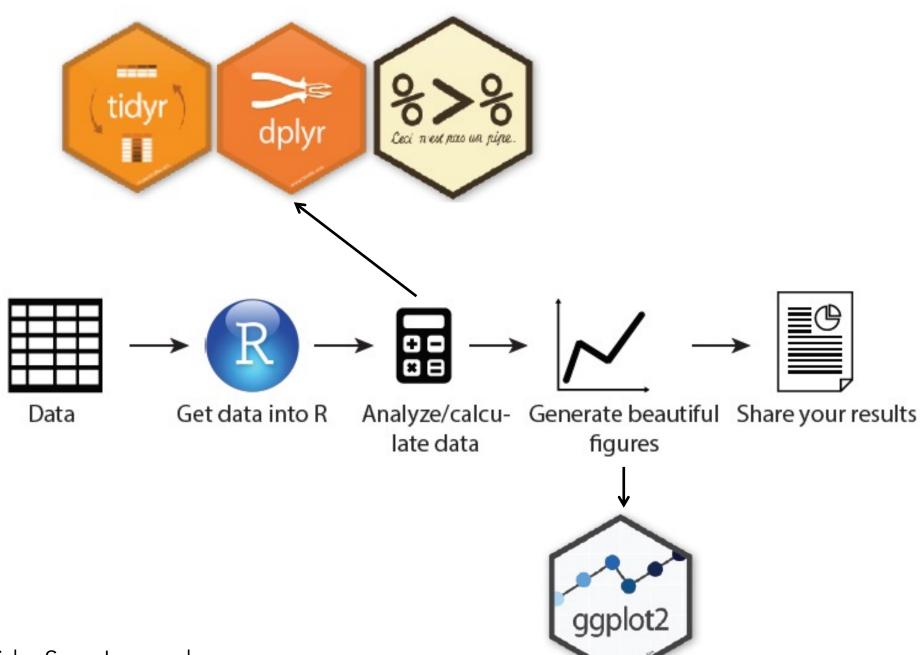
Working with data in R

E. Anne Chambers September 22, 2017



Slide: Sean Leonard

"Tidy datasets are all alike but every messy dataset is messy in its own way."

-Hadley Wickham

Tidy data: contingency table

	survived	died
drug	15	3
placebo	4	12

treatment	tment outcome	
drug	survived	15
drug	died	3
placebo	survived	4
placebo	died	12

1. Each variable forms a column

	survived	died
drug	15	3
placebo	4	12

treatment	outcome	count
drug	survived	15
drug	died	3
placebo	survived	4
placebo	died	12

2. Each observation forms a row

	survived	died
drug	15	3
placebo	4	12

treatment	outcome	count	
drug	survived	15	
drug	died	3	
placebo	survived	4	
placebo	died	12	

3. Each value has a cell

	survived	died
drug	15	3
placebo	4	12

treatment	outcome	count
drug	survived	15
drug	died	3
placebo	survived	4
placebo	died	12

Tidy data

- Rules of tidy data:
 - Each variable must have its own column
 - Each observation must have its own row
 - Each value must have its own cell

Before we go any further...

- Standard R:
 - > mean(dataframe\$column)
 - > mean(iris\$Sepal.Length)

The pipe operator!

- With pipe:
 - > iris\$Sepal.Length %>% mean()

```
%>%
"then"
```

Before we go any further...

Left and right assignment





• Left assignment:

```
> x <- 5 > x
```

Right assignment:

Combining pipe and assignment

These two lines do the same thing:

```
> iris$Sepal.Length %>% mean() -> mean.length
> mean.length <- iris$Sepal.Length %>% mean()
> mean.length
[1] 5.843333
```

- gather()
 - some columns are values of a variable

	2008	2013	2016
Reptiles	8541	9556	10228

```
> gather(variable value, key, value)
> data %>%
gather(`2008`, `2013`, `2016`,
key=Year, value=Species)
```

- gather()
 - some columns are values of a variable

	2008	2013	2016
Reptiles	8541	9556	10228

```
> gather(variable value, key, value)
> data %>%
gather(`2008`:`2016`,
key=Year, value=Species)
```

- gather()
 - some columns are values of a variable

	2008	2013	2016	
Reptiles	8541	9556	10228	

```
> gather(variable value, key, value)
> data %>%
gather(2:4,
key=Year, value=Species)
```

```
> gather(variable value, key, value)
> data %>%
gather(`2008`, `2013`, `2016`,
key=Year, value=Species) -> newdata
> newdata
```

	Year	Species
Reptiles	2008	8541
Reptiles	2013	9556
Reptiles	2016	10228

- spread()
 - observation scattered across multiple rows

Group	Туре	Year	Species
Reptiles	Lizards	2008	5079
Reptiles	Snakes	2008	3149
Reptiles	Turtles	2008	313
Reptiles	Lizards	2016	6263
Reptiles	Snakes	2016	3619
Reptiles	Turtles	2016	346

- > spread(key, value)
- > data %>%
 spread(key=Type, value=Species)

- > spread(key, value)
- > data %>%
 spread(key=Type, value=Species) -> newdata
- > newdata

Group	Year	Lizards	Snakes	Turtles
Reptiles	2008	5079	3149	313
Reptiles	2016	6263	3619	346

	Cycle [‡]	individuat	1 ‡	2 ‡	3 ÷	4 0	5 ‡	6 ‡	7 ÷	8 ‡	9 ‡	10 ‡
1	100	I(1)	1	NA								
2	100	I(10)	NA	NA	3	NA						
3	100	I(100)	NA	NA	NA	4	NA	NA	NA	NA	NA	NA
4	100	I(101)	1	NA								
5	100	I(102)	1	NA								
6	100	I(103)	NA	NA	NA	NA	NA	6	NA	NA	NA	NA
7	100	I(104)	NA	NA	NA	NA	NA	6	NA	NA	NA	NA
8	100	I(105)	NA	NA	NA	NA	NA	6	NA	NA	NA	NA
9	100	I(106)	NA	NA	NA	NA	NA	6	NA	NA	NA	NA
10	100	I(107)	NA	NA	NA	NA	NA	6	NA	NA	NA	NA

output

> tidyout <- output %>%
 gather(individual, population, -Cycle)

	Cycle [‡]	individuat	population
1	100	I(1)	1
2	200	I(1)	1
3	300	I(1)	1
4	400	I(1)	1
5	500	I(1)	1
6	600	I(1)	1
7	700	I(1)	1
8	800	I(1)	1
9	900	I(1)	1
10	1000	I(1)	1

R markdown explanation

 The remainder of our worksheets will be in R markdown

Exercise 1

Working with dplyr

Fundamental actions on data frames:

- select rows filter()
- select columns select()
- make new columns mutate()
- arrange rows arrange()
- calculate summary stats summarize()
- work on groups of data group_by()

Comparison and logical operators

```
< less than
```

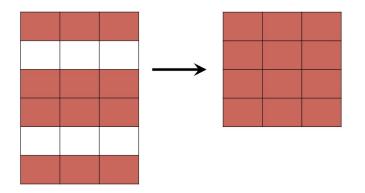
- <= less or equal to
- > greater than
- >= greater or equal to
- != does not equal
- == equal

& and

or

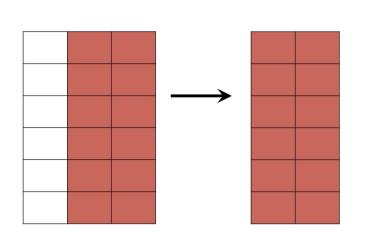
not

- filter() rows
- > tidyout %>% filter(Cycle==100 | Cycle==200)



	Cycle	individual	population
	<int></int>	<chr></chr>	<int></int>
1	100	I(1)	1
2	200	I(1)	1
3	100	I(2)	1
4	200	I(2)	2
5	100	I(3)	2
6	200	I(3)	3
7	100	I(4)	2
8	200	I(4)	3
9	100	I(5)	2
10	200	I(5)	3

- select() columns
- > tidyout %>% select(individual:population)



	individual	population	1
	<chr></chr>	<int></int>	>
1	I(1)	1	L
2	I(1)	1	
3	I(1)	1	L
4	I(1)	1	L
5	I(1)	1	L
6	I(1)	1	L
7	I(1)	1	L
8	I(1)	1	L
9	I(1)	1	
10	I(1)	1	
#	with 1.2	39 990 mor	e r

mutate() to add columns

- > dataframe %>% mutate(newcolname=fxn)
- Say I want to add a column with fraction of run that's finished...
- > tidyout %>% mutate(prop=Cycle/100000)

```
Cycle individual population prop
  <int>
             <chr>
                       <int> <dbl>
    100
              I(1)
                           1 0.001
    200
              I(1)
                           1 0.002
    300
              I(1)
                           1 0.003
    400
              I(1)
                      1 0.004
5
    500
              I(1)
                           1 0.005
    600
              I(1)
                          1 0.006
    700
              I(1)
                           1 0.007
    800
              I(1)
                           1 0.008
    900
              I(1)
                           1 0.009
   1000
              I(1)
                           1 0.010
# ... with 1,239,990 more rows
```

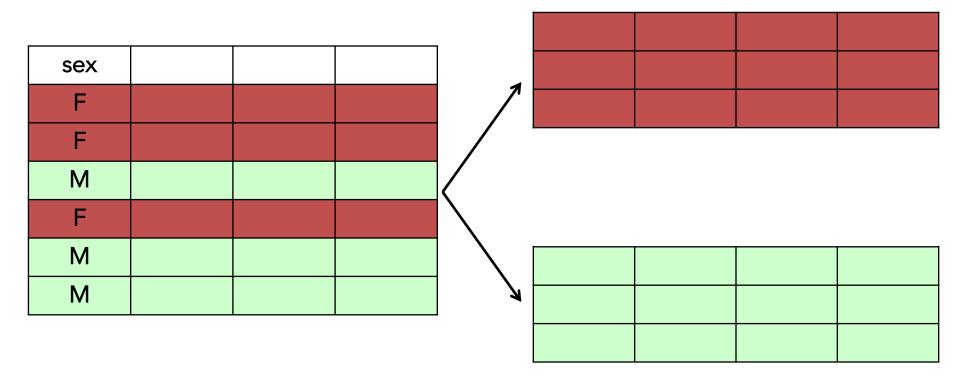
group_by() working on groups

> dataframe %>% group_by(varyouwantgrouped)

sex		
F		
F		
М		
F		
М		
М		

group_by() working on groups

> dataframe %>% group_by(varyouwantgrouped)



group_by() working on groups

- Say I want to only consider results from each individual at a time (not meaningful otherwise)
- > tidyout %>% group_by(individual) %>% ...

	Cycle ÷	individuat	population
1	100	I(1)	1
2	200	I(1)	1
3	300	I(1)	1
4	400	I(1)	1
5	500	I(1)	1
6	600	I(1)	1
7	700	I(1)	1
8	800	I(1)	1
9	900	I(1)	1
10	1000	I(1)	1

```
> tidyout %>%
  group_by(individual, population)
```

	Cycle ÷	individuat	population
1	100	I(1)	1
2	200	I(1)	1
3	300	I(1)	1
4	400	I(1)	1
5	500	I(1)	1
6	600	I(1)	1
7	700	I(1)	1
8	800	I(1)	1
9	900	I(1)	1
10	1000	I(1)	1

```
> tidyout %>%
  group_by(individual, population) %>%
  summarize(count=n())
```

	Cycle ÷	individuat	population
1	100	I(1)	1
2	200	I(1)	1
3	300	I(1)	1
4	400	I(1)	1
5	500	I(1)	1
6	600	I(1)	1
7	700	I(1)	1
8	800	I(1)	1
9	900	I(1)	1
10	1000	I(1)	1

```
> tidyout %>%
  group_by(individual, population) %>%
  summarize(count=n()) %>%
  mutate(proportion = count/100000)
  -> finalout tidyout
```

	Cycle [‡]	individuat	population
1	100	I(1)	1
2	200	I(1)	1
3	300	I(1)	1
4	400	I(1)	1
5	500	I(1)	1
6	600	I(1)	1
7	700	I(1)	1
8	800	I(1)	1
9	900	I(1)	1
10	1000	I(1)	1

```
> tidyout %>%
  group_by(individual, population) %>%
  summarize(count=n()) %>%
  mutate(proportion = count/100000)
  -> finalout
```

> finalout

	individual	population	count	proportion
	<chr></chr>	<int></int>	<int></int>	<dbl></dbl>
1	I(1)	1	10000	1.33333333
2	I(10)	1	492	0.06560000
3	I(10)	2	524	0.06986667
4	I(10)	3	4678	0.62373333
5	I(10)	4	4306	0.57413333
6	I(100)	2	1239	0.16520000
7	I(100)	3	1330	0.17733333
8	I(100)	4	4083	0.54440000
9	I(100)	5	3297	0.43960000
10	I(100)	6	51	0.00680000
#	with 582	2 more rows		

Useful summary functions

- mean() mean of values
- sum() sum values
- median() median
- sd() standard deviation
- var() variance
- cor() correlation

Exercise 2