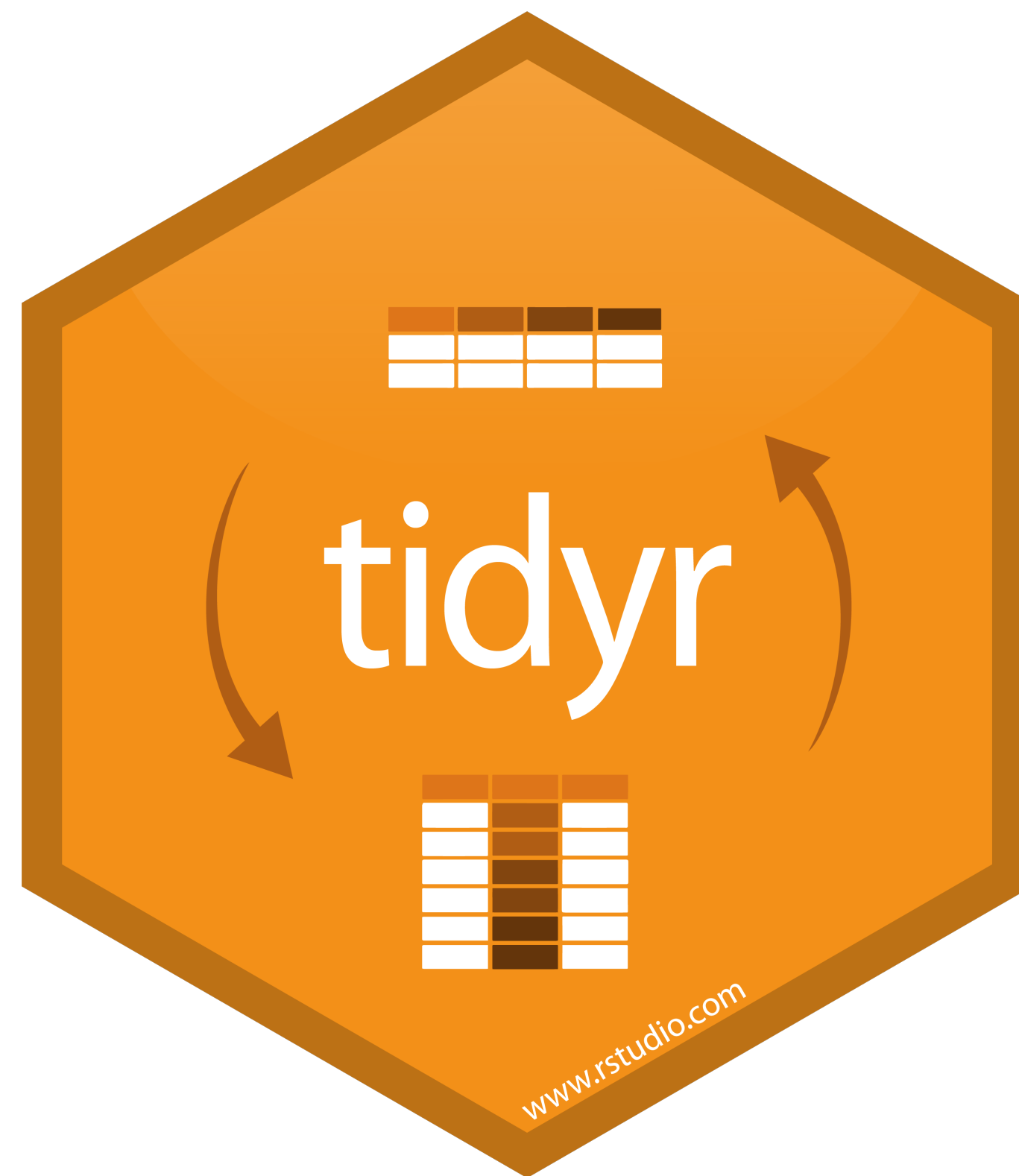


Tidy Data with

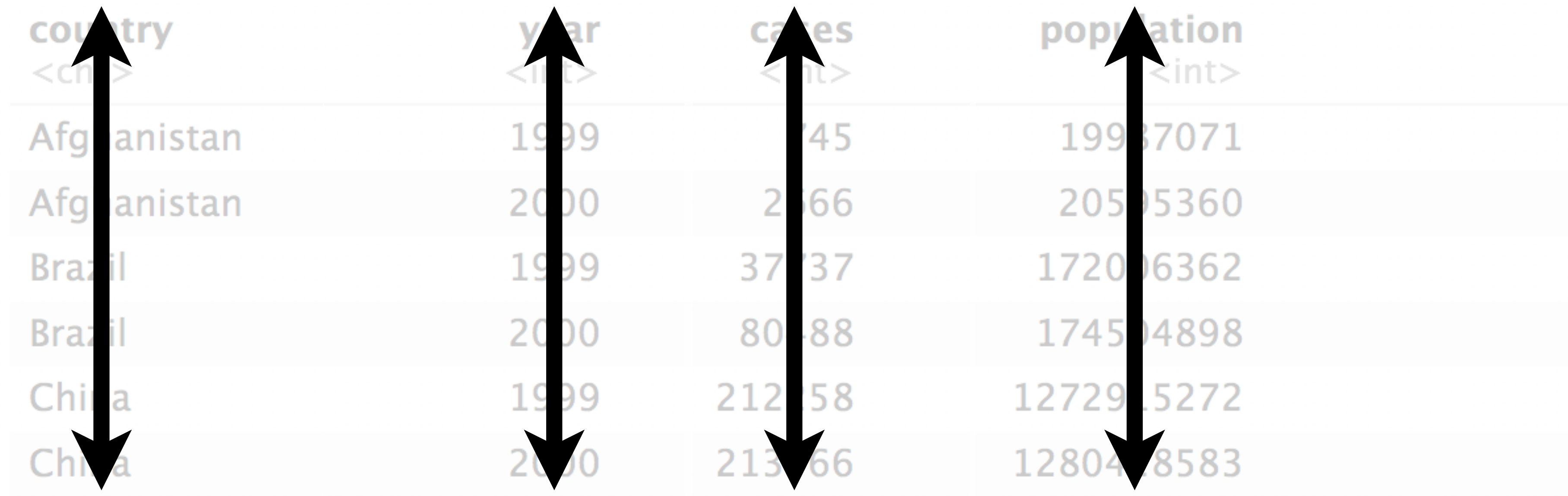


Open **03-Tidy-Data.Rmd**

Quiz

What are the variables in this data set?

table1



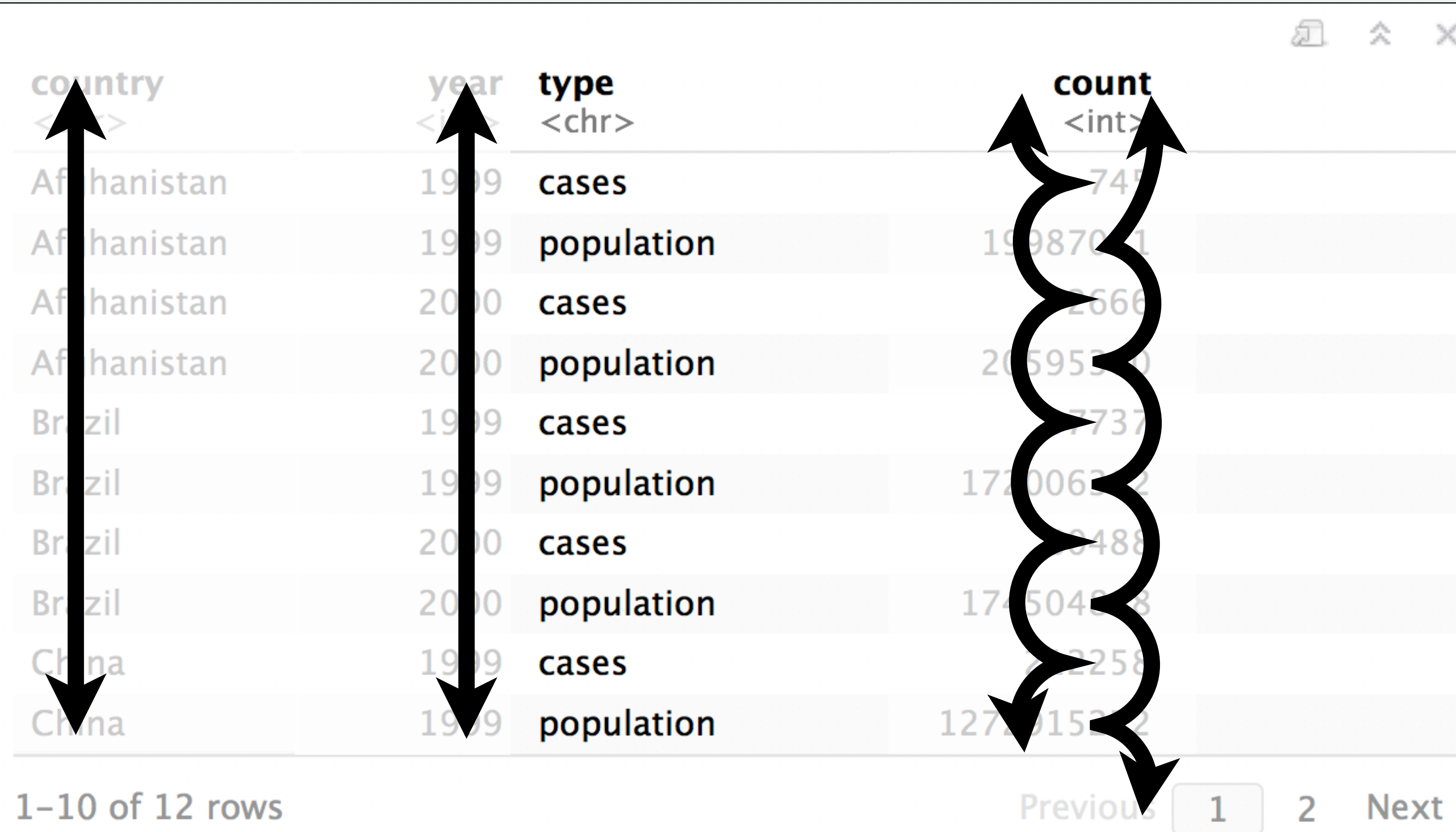
country <chr>	year <int>	cases <int>	population <int>
Afghanistan	1999	745	19987071
Afghanistan	2000	266	20595360
Brazil	1999	37737	172006362
Brazil	2000	80488	174504898
China	1999	212158	1272915272
China	2000	213766	128048583

6 rows

Quiz

What are the variables in this data set?

table2




The image shows a screenshot of a data table with four columns: country, year, type, and count. The table contains 10 rows of data. Hand-drawn arrows indicate the variables in the data set: a straight arrow for 'country', a straight arrow for 'year', and a wavy arrow for 'count'.

country <chr>	year <int>	type <chr>	count <int>
Afghanistan	1999	cases	745
Afghanistan	1999	population	1998701
Afghanistan	2000	cases	2666
Afghanistan	2000	population	2059530
Brazil	1999	cases	7737
Brazil	1999	population	17200632
Brazil	2000	cases	9488
Brazil	2000	population	17450488
China	1999	cases	2258
China	1999	population	12720152

1-10 of 12 rows

Previous 1 2 Next

table3



	country <chr>	year <int>	rate <chr>
1	Afghanistan	1999	745/19987071
2	Afghanistan	2000	2666/20595360
3	Brazil	1999	37737/172006362
4	Brazil	2000	80488/174504898
5	China	1999	212258/1272915272
6	China	2000	213766/1280428583

6 rows

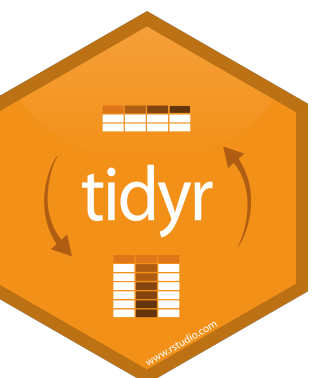


table4a

table4b

	country <chr>	1999 <int>	2000 <int>
1	Afghanistan	745	2666
2	Brazil	37737	80488
3	China	212258	213766

3 rows

	country <chr>	1999 <int>	2000 <int>
1	Afghanistan	19987071	20595360
2	Brazil	172006362	174504898
3	China	1272915272	1280428583

3 rows

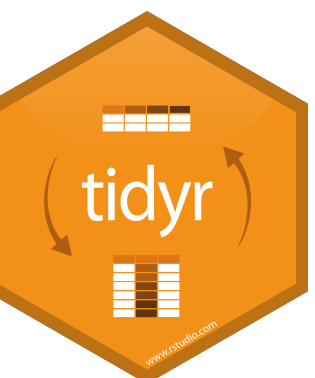
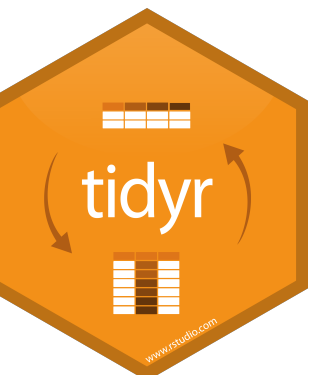


table5



	country <chr>	century <chr>	year <chr>	rate <chr>
1	Afghanistan	19	99	745/19987071
2	Afghanistan	20	00	2666/20595360
3	Brazil	19	99	37737/172006362
4	Brazil	20	00	80488/174504898
5	China	19	99	212258/1272915272
6	China	20	00	213766/1280428583

6 rows



"Data comes in many formats, but R prefers just one: tidy data. "

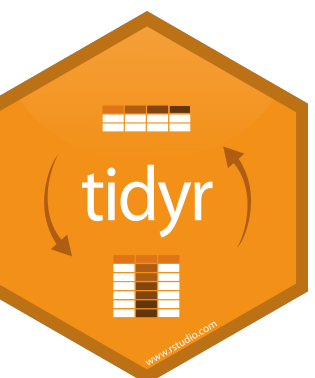
- Garrett Grolemund


Tidy data

country	year	cases	pop
Afghanistan	1999	745	10137321
Afghanistan	2000	666	20125120
Afghanistan	2001	3153	17525122
China	2000	3455	12113133
India	1998	22200	12713272
India	2000	3760	126142363

A data set is **tidy** iff:

1. Each **variable** is in its own **column**
2. Each **case** is in its own **row**
3. Each **value** is in its own **cell**

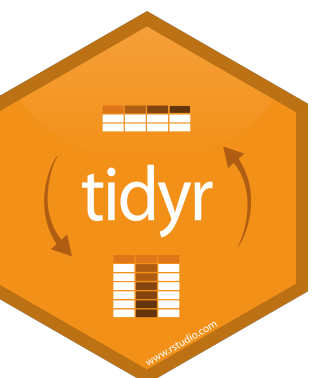




country <chr>	year <int>	cases <int>	population <int>
Afghanistan	1999	745	19987071
Afghanistan	2000	2666	20595360
Brazil	1999	37737	172006362
Brazil	2000	80488	174504898
China	1999	212258	1272915272
China	2000	213766	1280428583

6 rows

```
table1$country  
table1$year  
table1$cases  
table1$population
```



📄 ⬆ ✕

country <chr>	year <int>	type <chr>	count <int>
Afghanistan	1999	cases	745
Afghanistan	1999	pop	19987071
Afghanistan	2000	cases	2666
Afghanistan	2000	pop	195360
Brazil	1999	cases	737
Brazil	1999	pop	162
Brazil	2000	cases	88
Brazil	2000	pop	108
China	1999	cases	58
China	1999	pop	172

1–10 of 12 rows

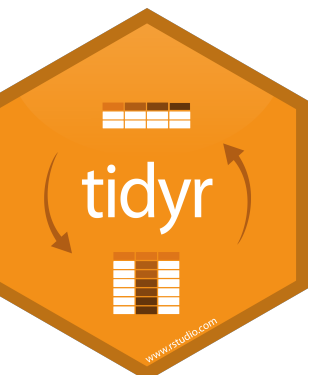
Previous 1 2 Next

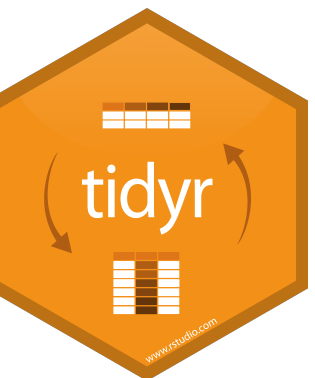
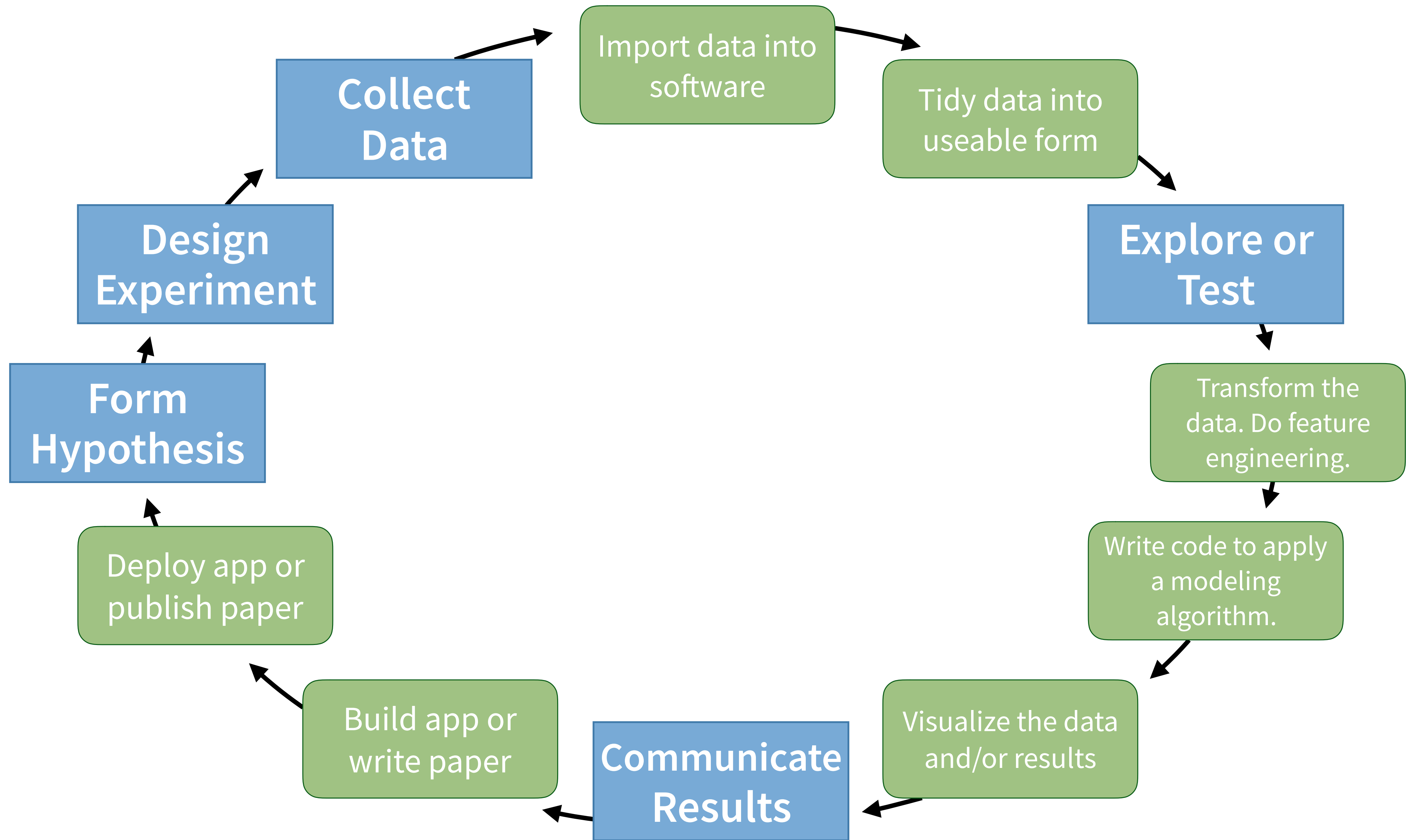
```
table2$country
table2$year
table2$count[c(1,3,5,7,9,11)]
table2$count[c(2,4,6,8,10,12)]
```

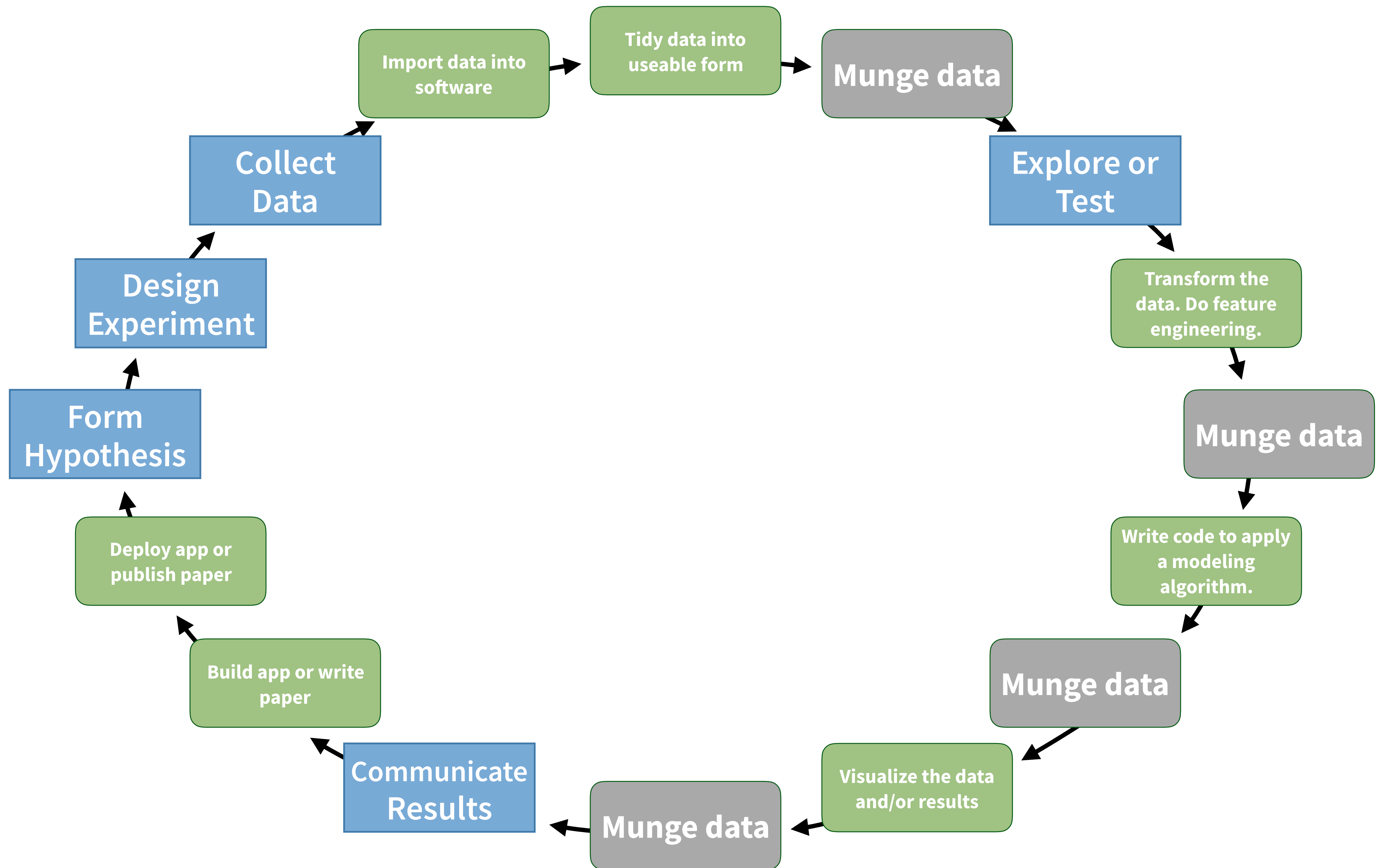

country <chr>	year <int>	cases <int>	population <int>	rate <dbl>
Afghanistan	1999	745	19987071	0.0000372741
Afghanistan	2000	2666	20595360	0.0001294466
Brazil	1999	37737	172006362	0.0002193930
Brazil	2000	80488	174504898	0.0004612363
China	1999	212258	1272915272	0.0001667495
China	2000	213766	1280428583	0.0001669488

6 rows

```
table1$cases / table1$population -> table1$rate
```







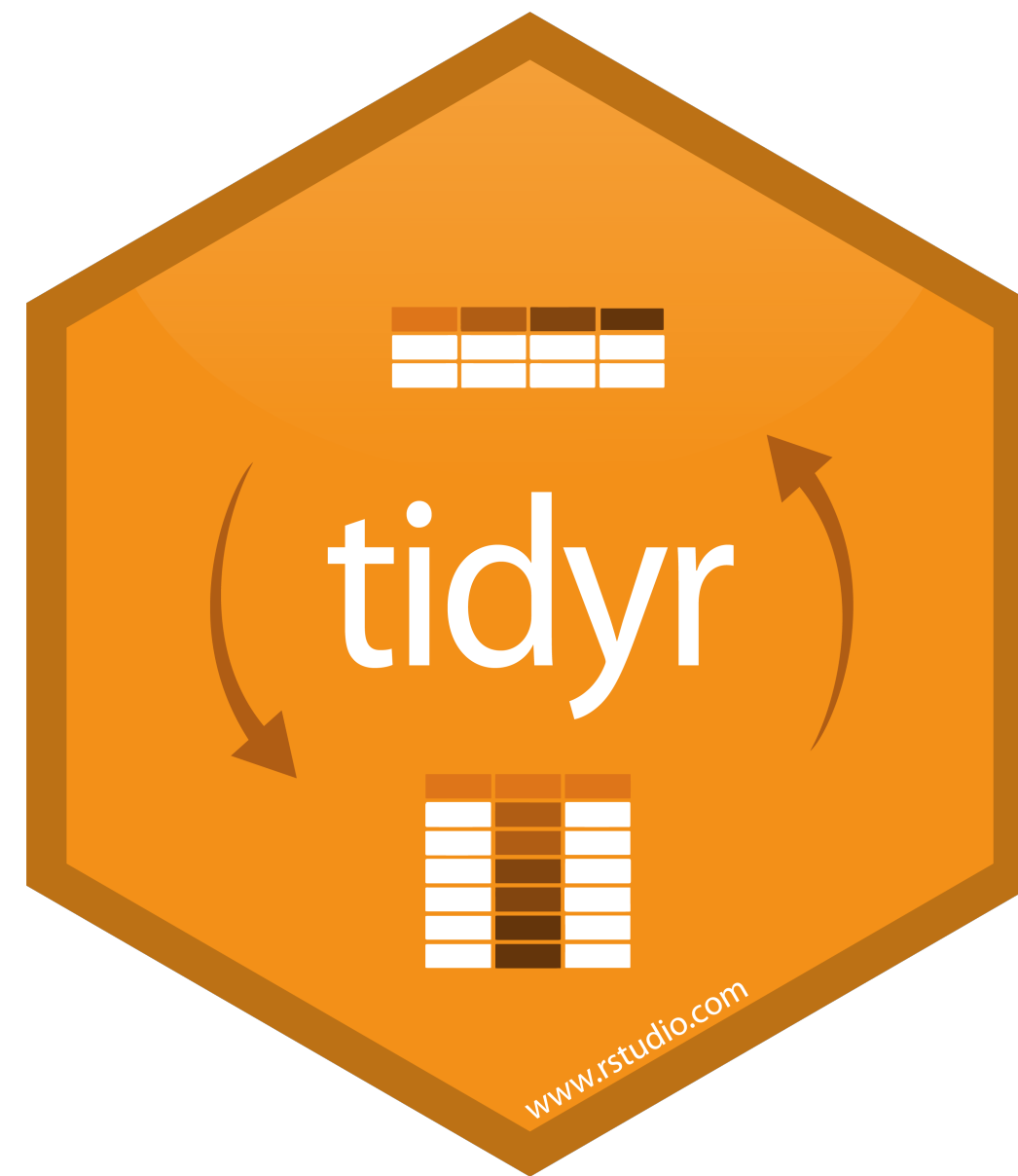
"Tidy data sets are all alike; but every messy data set is messy in its own way."

- Hadley Wickham

tidyr



tidyr



A package that reshapes the layout of tabular data.



gather()



Toy data

```
03-Tidy-Data.Rmd x
1 ---
2 title: "Tidy Data"
3 output: html_notebook
4 ---
5
6 ```{r setup}
7 library(tidyverse)
8 library(babynames)
9
10 # Toy data
11 cases <- tribble(
12   ~Country, ~"2011", ~"2012", ~"2013",
13   "FR",      7000,    6900,    7000,
14   "DE",      5800,    6000,    6200,
15   "US",     15000,   14000,    13000,
16 )
17
18 pollution <- tribble(
19   ~city, ~size, ~amount,
20   "New York", "large", 23,
21   "New York", "small", 14,
22   "London", "large", 22,
23   "London", "small", 16,
24   "Beijing", "large", 121,
25   "Beijing", "small", 121,
26 )
27
28 x <- tribble(
29   ~x1, ~x2,
30   "A", 1,
31   "B", NA,
32   "C", NA,
33   "D", 3,
34   "E", NA,
35 )
```

```
cases <- tribble(
  ~Country, ~"2011", ~"2012", ~"2013",
  "FR",      7000,    6900,    7000,
  "DE",      5800,    6000,    6200,
  "US",     15000,   14000,    13000
)
```



Quiz

What are the variables in cases?

Country	2011	2012	2013
FR	7000	6900	7000
DE	5800	6000	6200
US	15000	14000	13000

Quiz

What are the variables in cases?

Country	2011	2012	2013
FR	7000	6900	7000
DE	5800	6000	6200
US	15000	14500	13000

- Country
- Year
- Count

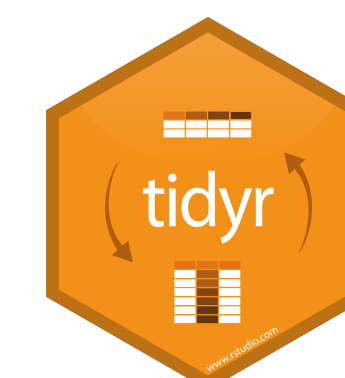
Your Turn 1

On a sheet of paper, draw how the cases data set would look if it had the same values grouped into three columns: *country*, *year*, *n*

Country	2011	2012	2013
FR	7000	6900	7000
DE	5800	6000	6200
US	15000	14000	13000

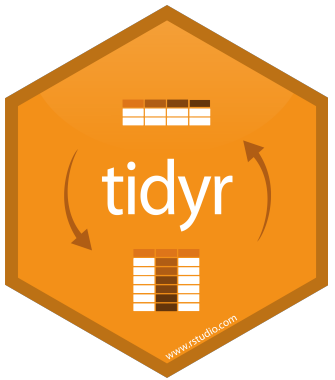
04:00

Country	2011	2012	2013
FR	7000	6900	7000
DE	5800	6000	6200
US	15000	14000	13000



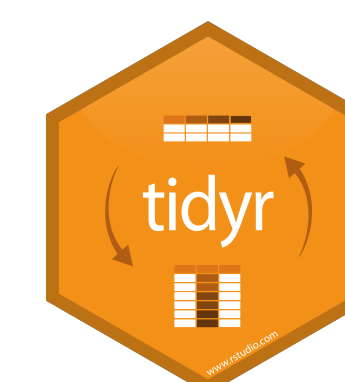
Country	2011	2012	2013
FR	7000	6900	7000
DE	5800	6000	6200
US	15000	14000	13000

Country	Year	n
---------	------	---



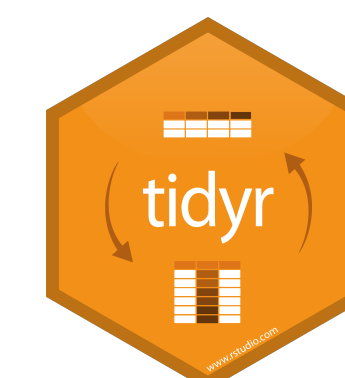
Country	2011	2012	2013
FR	7000	6900	7000
DE	5800	6000	6200
US	15000	14000	13000

Country	Year	n
FR	2011	7000



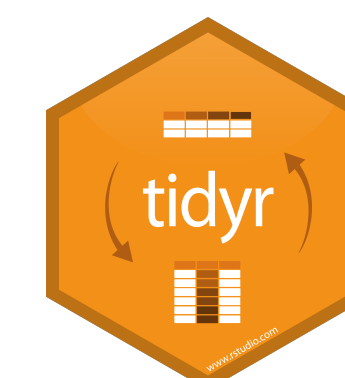
Country	2011	2012	2013
FR	7000	6900	7000
DE	5800	6000	6200
US	15000	14000	13000

Country	Year	n
FR	2011	7000
DE	2011	5800



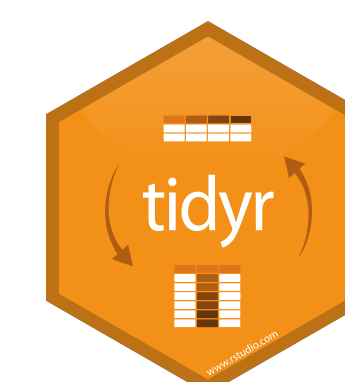
Country	2011	2012	2013
FR	7000	6900	7000
DE	5800	6000	6200
US	15000	14000	13000

Country	Year	n
FR	2011	7000
DE	2011	5800
US	2011	15000



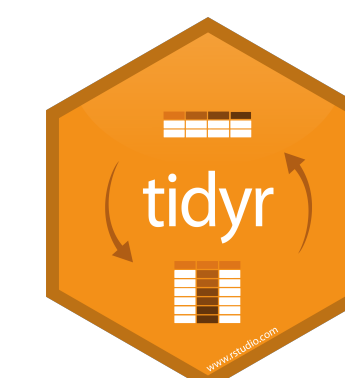
Country	2011	2012	2013
FR	7000	6900	7000
DE	5800	6000	6200
US	15000	14000	13000

Country	Year	n
FR	2011	7000
DE	2011	5800
US	2011	15000
FR	2012	6900



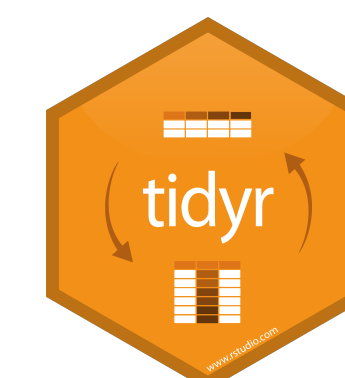
Country	2011	2012	2013
FR	7000	6900	7000
DE	5800	6000	6200
US	15000	14000	13000

Country	Year	n
FR	2011	7000
DE	2011	5800
US	2011	15000
FR	2012	6900
DE	2012	6000



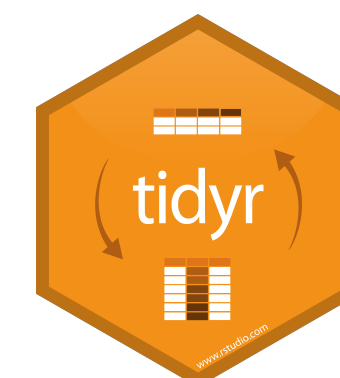
Country	2011	2012	2013
FR	7000	6900	7000
DE	5800	6000	6200
US	15000	14000	13000

Country	Year	n
FR	2011	7000
DE	2011	5800
US	2011	15000
FR	2012	6900
DE	2012	6000
US	2012	14000



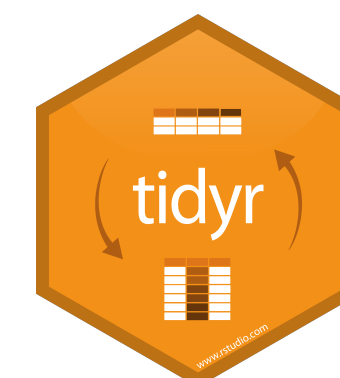
Country	2011	2012	2013
FR	7000	6900	7000
DE	5800	6000	6200
US	15000	14000	13000

Country	Year	n
FR	2011	7000
DE	2011	5800
US	2011	15000
FR	2012	6900
DE	2012	6000
US	2012	14000
FR	2013	7000



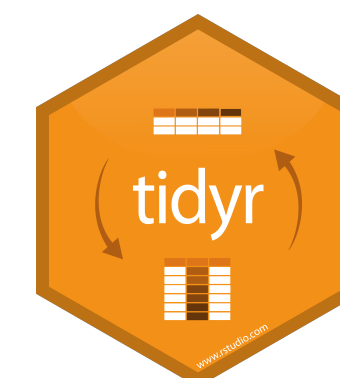
Country	2011	2012	2013
FR	7000	6900	7000
DE	5800	6000	6200
US	15000	14000	13000

Country	Year	n
FR	2011	7000
DE	2011	5800
US	2011	15000
FR	2012	6900
DE	2012	6000
US	2012	14000
FR	2013	7000
DE	2013	6200



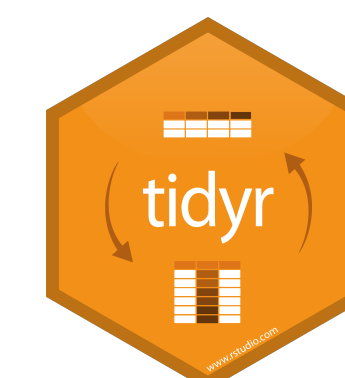
Country	2011	2012	2013
FR	7000	6900	7000
DE	5800	6000	6200
US	15000	14000	13000

Country	Year	n
FR	2011	7000
DE	2011	5800
US	2011	15000
FR	2012	6900
DE	2012	6000
US	2012	14000
FR	2013	7000
DE	2013	6200
US	2013	13000



Country	2011	2012	2013
FR	7000	6900	7000
DE	5800	6000	6200
US	15000	14000	13000

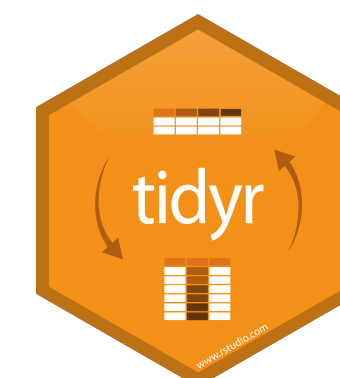
Country	Year	Revenue
FR	2011	7000
DE	2011	5800
US	2011	15000
FR	2012	6900
DE	2012	6000
US	2012	14000
FR	2013	7000
DE	2013	6200
US	2013	13000



Country	2011	2012	2013
FR	7000	6900	7000
DE	5800	6000	6200
US	15000	14000	13000

gather()

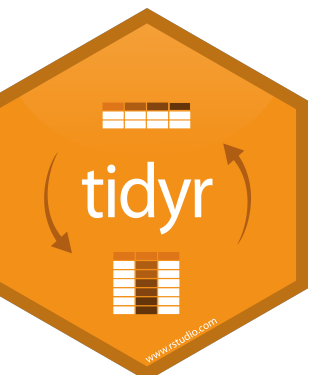
Country	Year	n
FR	2011	7000
DE	2011	5800
US	2011	15000
FR	2012	6900
DE	2012	6000
US	2012	14000
FR	2013	7000
DE	2013	6200
US	2013	13000



Country	2011	2012	2013
FR	7000	6900	7000
DE	5800	6000	6200
US	15000	14000	13000

12

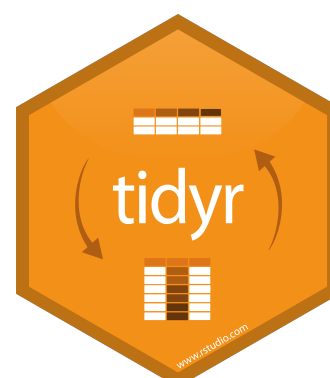
Country	Year	n
FR	2011	7000
DE	2011	5800
US	2011	15000
FR	2012	6900
DE	2012	6000
US	2012	14000
FR	2013	7000
DE	2013	6200
US	2013	13000



key (former column names)

Country	2011	2012	2013
FR	7000	6900	7000
DE	5800	6000	6200
US	15000	14000	13000

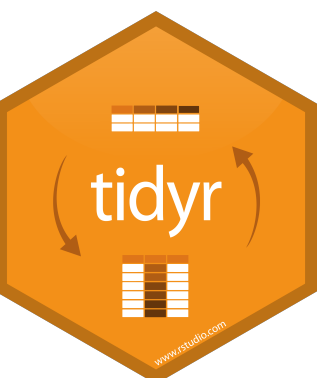
Country	Year	n
FR	2011	7000
DE	2011	5800
US	2011	15000
FR	2012	6900
DE	2012	6000
US	2012	14000
FR	2013	7000
DE	2013	6200
US	2013	13000



key **value** (former cells)

Country	2011	2012	2013
FR	7000	6900	7000
DE	5800	6000	6200
US	15000	14000	13000

Country	Year	n
FR	2011	7000
DE	2011	5800
US	2011	15000
FR	2012	6900
DE	2012	6000
US	2012	14000
FR	2013	7000
DE	2013	6200
US	2013	13000



gather()

```
cases %>% gather(key = "year", value = "n", 2:4)
```

**data frame to
reshape**

**name of the
new key
column**
(a character
string)

**name of the
new value
column**
(a character
string)

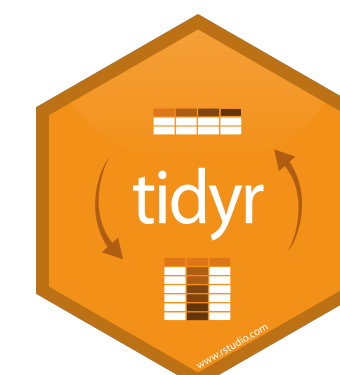
**numeric
indexes of
columns to
collapse**
(or names)

gather()

```
cases %>% gather("year", "n", 2:4)
```

numeric
indexes

Country <chr>	2 2011 <dbl>	3 2012 <dbl>	4 2013 <dbl>
FR	7000	6900	7000
DE	5800	6000	6200
US	15000	14000	13000



gather()

```
cases %>% gather("year", "n", "2011", "2012", "2013")
```

names

	2011	2012	2013
Country <chr>	2011 <dbl>	2012 <dbl>	2013 <dbl>
FR	7000	6900	7000
DE	5800	6000	6200
US	15000	14000	13000

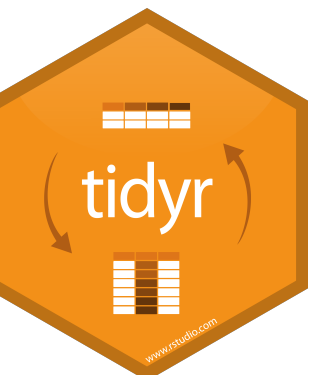
gather()

```
cases %>% gather("year", "n", -Country)
```

Everything
except...

Not Country Not Country Not Country

Country <chr>	2011 <dbl>	2012 <dbl>	2013 <dbl>
FR	7000	6900	7000
DE	5800	6000	6200
US	15000	14000	13000



Your Turn 2


Use **gather()** to reorganize **table4a** into three columns: *country*, *year*, and *cases*.

	country <chr>	1999 <int>	2000 <int>
1	Afghanistan	745	2666
2	Brazil	37737	80488
3	China	212258	213766

3 rows

03:00



```
table4a %>%  
  gather(key = "year", value = "n", 2:3)
```



country <chr>	year <chr>	n <int>
Afghanistan	1999	745
Brazil	1999	37737
China	1999	212258
Afghanistan	2000	2666
Brazil	2000	80488
China	2000	213766

6 rows

```
table4a %>%  
  gather(key = "year", value = "n", 2:3, convert = TRUE)
```



country <chr>	year <int>	n <int>
Afghanistan	1999	745
Brazil	1999	37737
China	1999	212258
Afghanistan	2000	2666
Brazil	2000	80488
China	2000	213766

6 rows



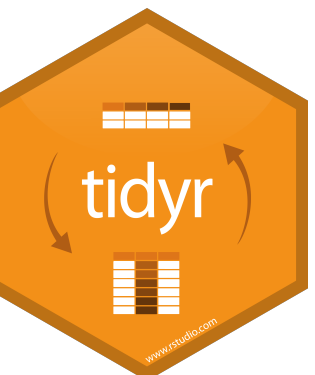
spread()



Toy data

```
03-Tidy-Data.Rmd x
1 ---
2 title: "Tidy Data"
3 output: html_notebook
4 ---
5
6 ```{r setup}
7 library(tidyverse)
8 library(babynames)
9
10 # Toy data
11 cases <- tribble(
12   ~Country, ~"2011", ~
13     "FR", 7000,
14     "DE", 5800,
15     "US", 15000,
16 )
17
18 pollution <- tribble(
19   ~city, ~size, ~
20     "New York", "large",
21     "New York", "small",
22     "London", "large",
23     "London", "small",
24     "Beijing", "large",
25     "Beijing", "small",
26 )
27
28 x <- tribble(
29   ~x1, ~x2,
30     "A", 1,
31     "B", NA,
32     "C", NA,
33     "D", 3,
34     "E", NA
35 )
```

```
pollution <- tribble(
  ~city, ~size, ~amount,
  "New York", "large", 23,
  "New York", "small", 14,
  "London", "large", 22,
  "London", "small", 16,
  "Beijing", "large", 121,
  "Beijing", "small", 56
)
```



Quiz

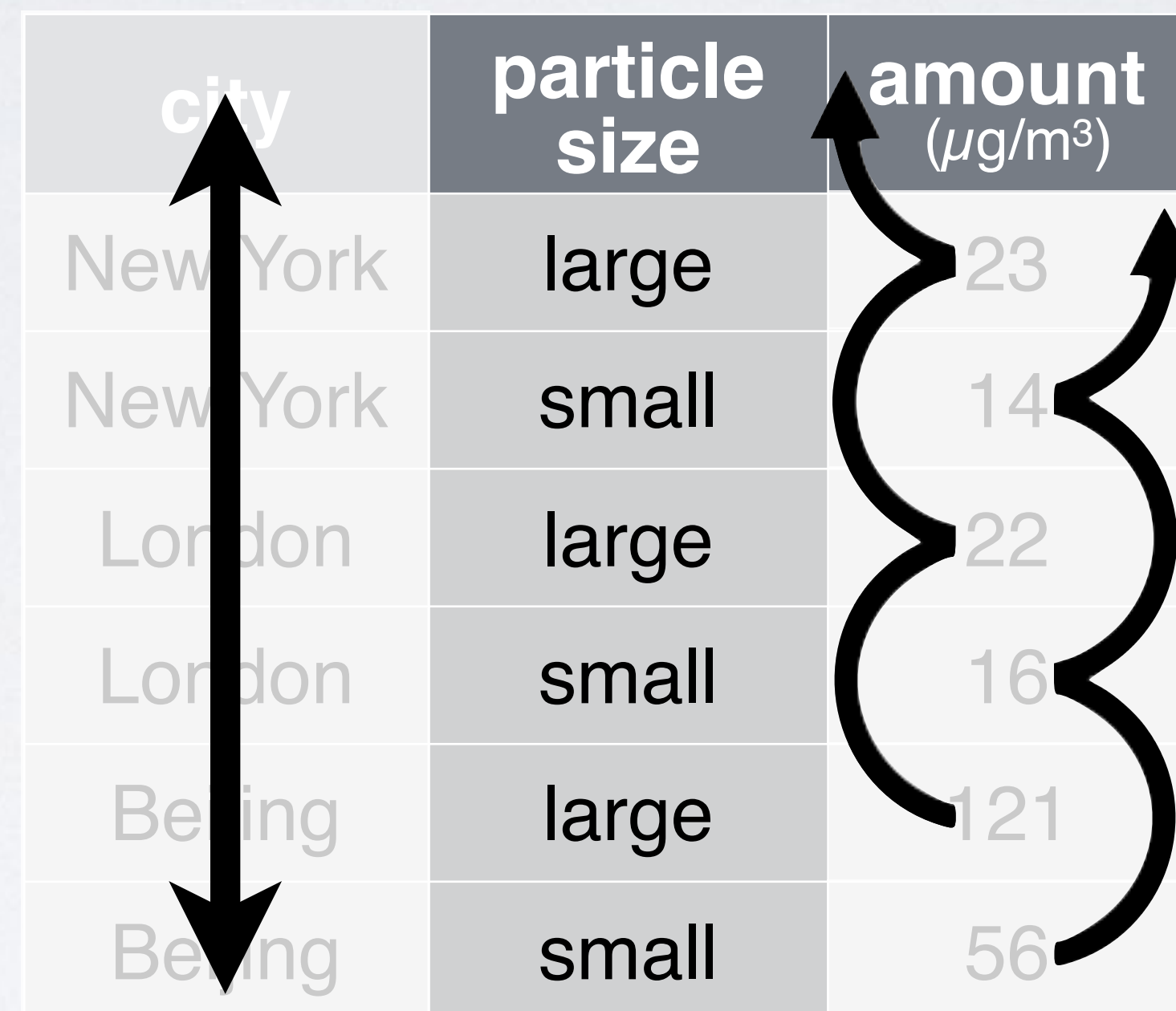
What are the variables in pollution?

city	particle size	amount ($\mu\text{g}/\text{m}^3$)
New York	large	23
New York	small	14
London	large	22
London	small	16
Beijing	large	121
Beijing	small	56

Quiz

What are the variables in pollution?

city	particle size	amount ($\mu\text{g}/\text{m}^3$)
New York	large	23
New York	small	14
London	large	22
London	small	16
Beijing	large	121
Beijing	small	56



The diagram illustrates the relationships between the variables in the table. A long vertical double-headed arrow spans the 'city' column, indicating that 'city' is a variable. A curved double-headed arrow connects the 'particle size' and 'amount' columns, indicating that 'particle size' is a variable. Another curved double-headed arrow connects the two rows for each city in the 'amount' column, indicating that 'amount' is a variable.

- City
- Amount of large particulate
- Amount of small particulate

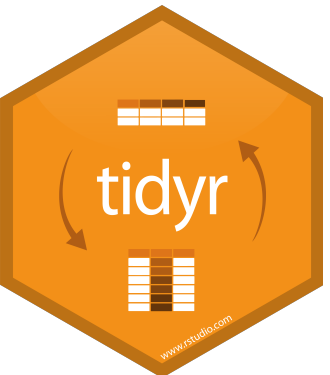
Your Turn 3

On a sheet of paper, draw how this data set would look if it had the same values grouped into three columns: *city*, *large*, *small*

city	size	amount
New York	large	23
New York	small	14
London	large	22
London	small	16
Beijing	large	121
Beijing	small	56

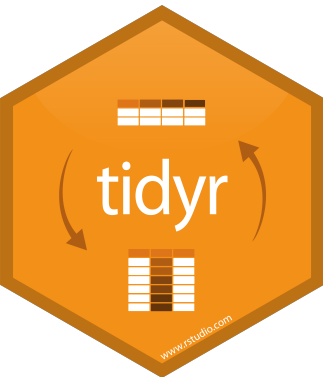
03:00

city	size	amount
New York	large	23
New York	small	14
London	large	22
London	small	16
Beijing	large	121
Beijing	small	56



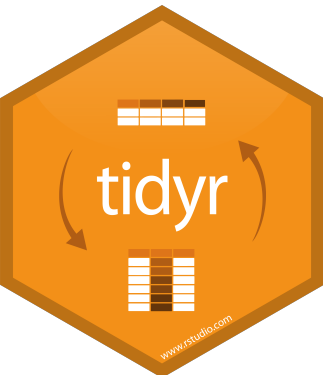
city	size	amount
New York	large	23
New York	small	14
London	large	22
London	small	16
Beijing	large	121
Beijing	small	56

city	large	small
------	-------	-------



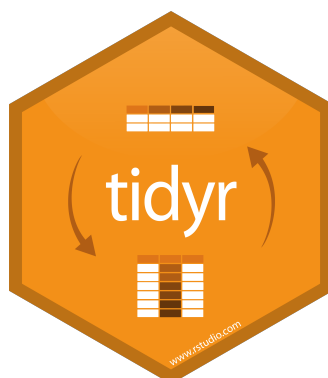
city	size	amount
New York	large	23
New York	small	14
London	large	22
London	small	16
Beijing	large	121
Beijing	small	56

city	large	small
New York	23	



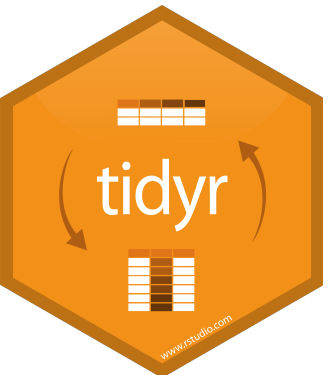
city	size	amount
New York	large	23
New York	small	14
London	large	22
London	small	16
Beijing	large	121
Beijing	small	56

city	large	small
New York	23	14



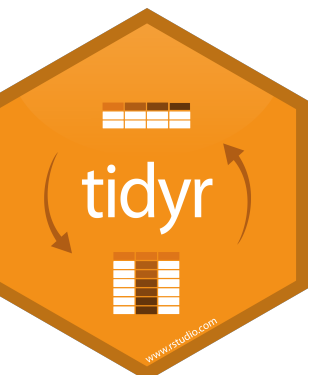
city	size	amount
New York	large	23
New York	small	14
London	large	22
London	small	16
Beijing	large	121
Beijing	small	56

city	large	small
New York	23	14
London	22	



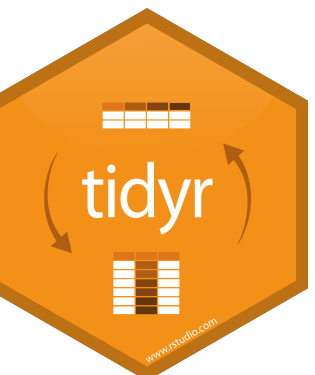
city	size	amount
New York	large	23
New York	small	14
London	large	22
London	small	16
Beijing	large	121
Beijing	small	56

city	large	small
New York	23	14
London	22	16



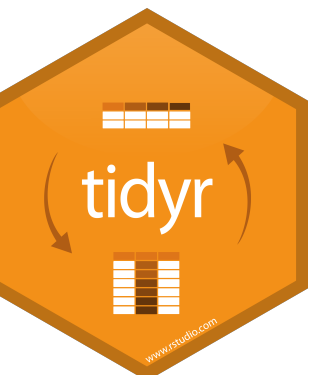
city	size	amount
New York	large	23
New York	small	14
London	large	22
London	small	16
Beijing	large	121
Beijing	small	56

city	large	small
New York	23	14
London	22	16
Beijing	121	



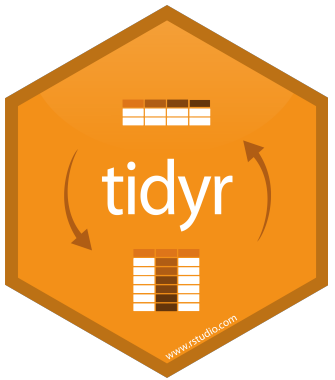
city	size	amount
New York	large	23
New York	small	14
London	large	22
London	small	16
Beijing	large	121
Beijing	small	56

city	large	small
New York	23	14
London	22	16
Beijing	121	56



city	size	amount
New York	large	23
New York	small	14
London	large	22
London	small	16
Beijing	large	121
Beijing	small	56

city	large	small
New York	23	14
London	22	16
Beijing	121	56



city	size	amount
New York	large	23
New York	small	14
London	large	22
London	small	16
Beijing	large	121
Beijing	small	56



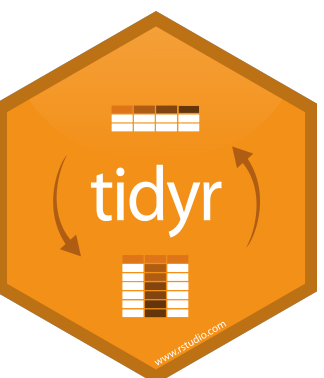
city	large	small
New York	23	14
London	22	16
Beijing	121	56

1

2

city	size	amount
New York	large	23
New York	small	14
London	large	22
London	small	16
Beijing	large	121
Beijing	small	56

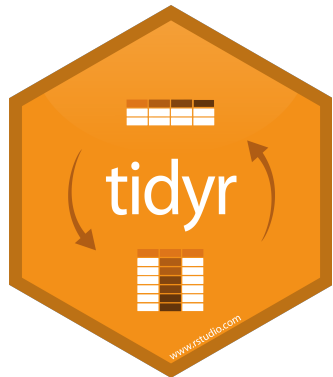
city	large	small
New York	23	14
London	22	16
Beijing	121	56



key (new column names)

city	size	amount
New York	large	23
New York	small	14
London	large	22
London	small	16
Beijing	large	121
Beijing	small	56

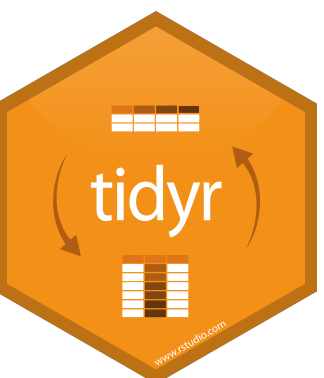
city	large	small
New York	23	14
London	22	16
Beijing	121	56



key **value** (new cells)

city	size	amount
New York	large	23
New York	small	14
London	large	22
London	small	16
Beijing	large	121
Beijing	small	56

city	large	small
New York	23	14
London	22	16
Beijing	121	56



spread()

```
pollution %>% spread(key = size, value = amount)
```

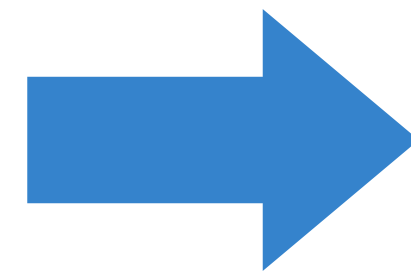
**data frame to
reshape**

column to use for keys
(becomes new
column names)

column to use for values
(becomes new
column cells)

```
pollution %>% spread(size, amount)
```

	city	size	amount
1	New York	large	23
2	New York	small	14
3	London	large	22
4	London	small	16
5	Beijing	large	121
6	Beijing	small	56



	city	large	small
1	Beijing	121	56
2	London	22	16
3	New York	23	14


Your Turn 4

Use **spread()** to reorganize **table2** into four columns: *country*, *year*, *cases*, and *population*.

country <chr>	year <int>	type <chr>	count <int>
Afghanistan	1999	cases	745
Afghanistan	1999	population	19987071
Afghanistan	2000	cases	2666
Afghanistan	2000	population	20595360
Brazil	1999	cases	37737
Brazil	1999	population	172006362

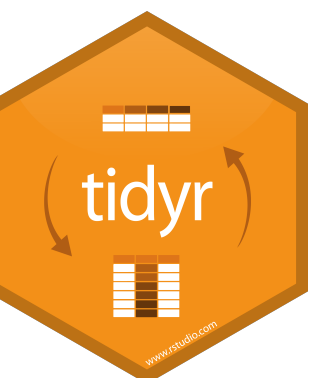
03:00

```
table2 %>%  
  spread(key = type, value = count)
```



	country <chr>	year <int>	cases <int>	population <int>
1	Afghanistan	1999	745	19987071
2	Afghanistan	2000	2666	20595360
3	Brazil	1999	37737	172006362
4	Brazil	2000	80488	174504898
5	China	1999	212258	1272915272
6	China	2000	213766	1280428583

6 rows



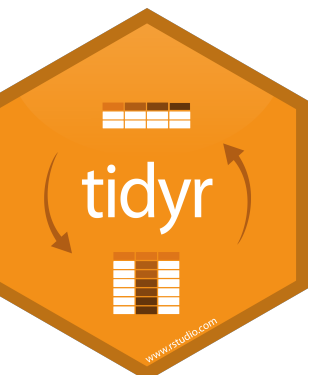
who
(Untidy Data)



who

Tuberculosis (TB) cases broken down by year, country, age, gender, and diagnosis method from the *2014 World Health Organization Global Tuberculosis Report*

```
View(who)
```



~/Dropbox (RStudio)/RStudio/training/U-Master-the-tidyverse - RStudio Source Editor															
who															
Filter															
	country	iso2	iso3	year	new_sp_m014	new_sp_m1524	new_sp_m2534	new_sp_m3544	new_sp_m4554	new_sp_m5564	new_sp_m65	new_sp_f014	new_sp_f1524	new_sp_f2534	new_sp_f3544
1	Afghanistan	AF	AFG	1980	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
2	Afghanistan	AF	AFG	1981	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
3	Afghanistan	AF	AFG	1982	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
4	Afghanistan	AF	AFG	1983	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
5	Afghanistan	AF	AFG	1984	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
6	Afghanistan	AF	AFG	1985	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
7	Afghanistan	AF	AFG	1986	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
8	Afghanistan	AF	AFG	1987								NA	NA	NA	NA
9	Afghanistan	AF	AFG	1988								NA	NA	NA	NA
10	Afghanistan	AF	AFG	1989								NA	NA	NA	NA
11	Afghanistan	AF	AFG	1990								NA	NA	NA	NA
12	Afghanistan	AF	AFG	1991								NA	NA	NA	NA
13	Afghanistan	AF	AFG	1992								NA	NA	NA	NA
14	Afghanistan	AF	AFG	1993								NA	NA	NA	NA
15	Afghanistan	AF	AFG	1994	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
16	Afghanistan	AF	AFG	1995	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
17	Afghanistan	AF	AFG	1996	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
18	Afghanistan	AF	AFG	1997	0	10	6	3	5	2	0	5	38	36	14
19	Afghanistan	AF	AFG	1998	30	129	128	90	89	64	41	45	350	419	194
20	Afghanistan	AF	AFG	1999	8	55	55	47	34	21	8	25	139	160	110
21	Afghanistan	AF	AFG	2000	52	228	183	149	129	94	80	93	414	565	339
22	Afghanistan	AF	AFG	2001	129	379	349	274	204	139	103	146	799	888	586

Showing 1 to 22 of 7,240 entries

What variables does this data set contain?

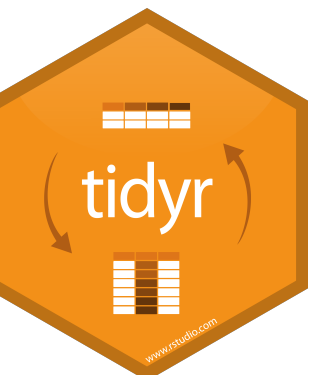
who variables

country		iso2	iso3	year	new_sp_m014
----------------	--	-------------	-------------	-------------	--------------------

country, iso2, iso3 - country identifiers

year - year

other columns names - encode **type** of TB case, **sex**, and **age**



who codes

new_sp_m014

Type of TB case

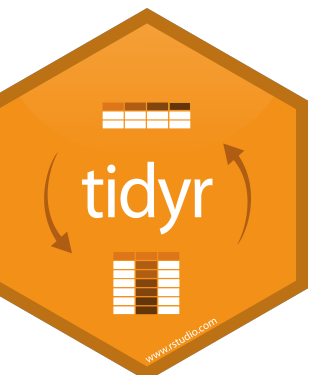
- **rel** - relapse
- **ep** - extra-pulmonary
- **sn** - pulmonary, smear negative
- **sp** - pulmonary, smear positive

Gender

- **m** - male
- **f** - female

Age group

- **014** - 0 to 14 years old
- **1524** - 15 to 24 years old
- **2534** - 25 to 34 years old
- **3544** - 35 to 44 years old
- **4554** - 45 to 54 years old
- **5564** - 55 to 64 years old
- **65** - 65 and older



Your Turn 5

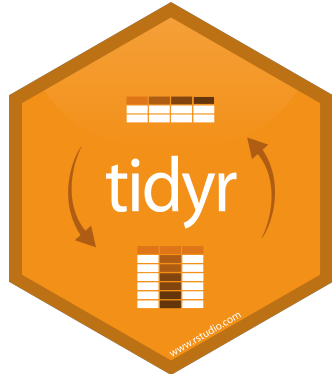
Gather the **5th through 60th** columns of `who` into a pair of key:value columns named ***codes*** and ***n***.

Then select just the **`county`**, **`year`**, **`codes`** and **`n`** variables.

05:00

```
who %>%  
  gather("codes", "n", 5:60) %>%  
  select(-iso2, -iso3)
```

country <chr>	year <int>	codes <chr>	n <int>
Afghanistan	1980	new_sp_m014	NA
Afghanistan	1981	new_sp_m014	NA
Afghanistan	1982	new_sp_m014	NA
Afghanistan	1983	new_sp_m014	NA
Afghanistan	1984	new_sp_m014	NA
Afghanistan	1985	new_sp_m014	NA
Afghanistan	1986	new_sp_m014	NA
Afghanistan	1987	new_sp_m014	NA
Afghanistan	1988	new_sp_m014	NA
Afghanistan	1989	new_sp_m014	NA



separate()



separate()

Splits a column by dividing values at a specific character.

```
who %>%
```

```
gather("codes", "n", 5:60) %>%
```

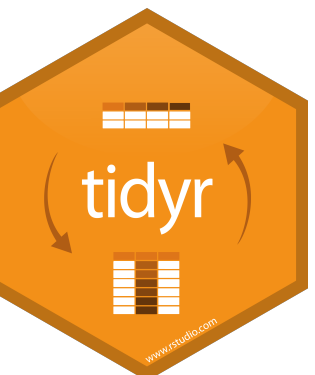
```
select(-iso2, -iso3) %>%
```

```
separate(codes, into = c("new", "type", "sexage"), sep = "_")
```

**a column to
split**

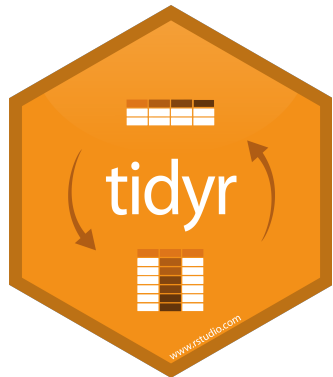
**names of new
columns to
make**

string to split on
(Defaults to any non_alpha-
numeric character)



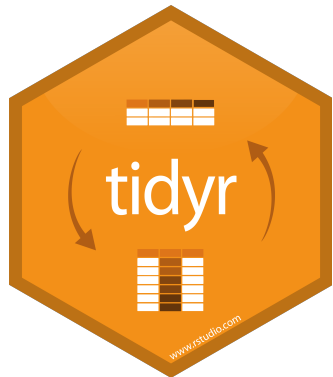
```
who %>%  
  gather("codes", "n", 5:60) %>%  
  select(-iso2, -iso3) %>%  
  separate( )
```

country <chr>	year <int>	codes <chr>	n <int>
Afghanistan	1980	new_sp_m014	NA
Afghanistan	1981	new_sp_m014	NA
Afghanistan	1982	new_sp_m014	NA
Afghanistan	1983	new_sp_m014	NA
Afghanistan	1984	new_sp_m014	NA
Afghanistan	1985	new_sp_m014	NA
Afghanistan	1986	new_sp_m014	NA
Afghanistan	1987	new_sp_m014	NA
Afghanistan	1988	new_sp_m014	NA
Afghanistan	1989	new_sp_m014	NA



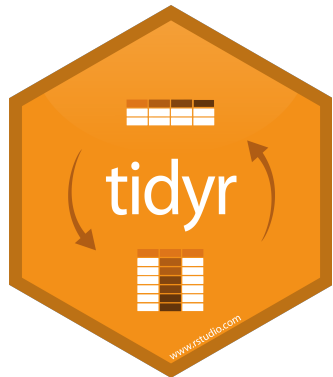

```
who %>%
  gather("codes", "n", 5:60) %>%
  select(-iso2, -iso3) %>%
  separate(codes
```

country <chr>	year <int>	codes <chr>	n <int>
Afghanistan	1980	new_sp_m014	NA
Afghanistan	1981	new_sp_m014	NA
Afghanistan	1982	new_sp_m014	NA
Afghanistan	1983	new_sp_m014	NA
Afghanistan	1984	new_sp_m014	NA
Afghanistan	1985	new_sp_m014	NA
Afghanistan	1986	new_sp_m014	NA
Afghanistan	1987	new_sp_m014	NA
Afghanistan	1988	new_sp_m014	NA
Afghanistan	1989	new_sp_m014	NA



```
who %>%
  gather("codes", "n", 5:60) %>%
  select(-iso2, -iso3) %>%
  separate(codes, into = c("new", "type", "sexage"))
```

country<chr>	year<int>	codes<chr>	new<chr>	type<chr>	sexage<chr>	n<int>
Afghanistan	1980	new_sp_m014				NA
Afghanistan	1981	new_sp_m014				NA
Afghanistan	1982	new_sp_m014				NA
Afghanistan	1983	new_sp_m014				NA
Afghanistan	1984	new_sp_m014				NA
Afghanistan	1985	new_sp_m014				NA
Afghanistan	1986	new_sp_m014				NA
Afghanistan	1987	new_sp_m014				NA
Afghanistan	1988	new_sp_m014				NA
Afghanistan	1989	new_sp_m014				NA



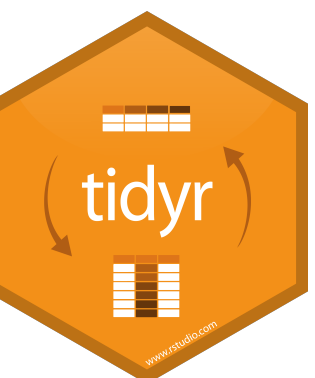

```
who %>%
```

```
gather("codes", "n", 5:60) %>%
```

```
select(-iso2, -iso3) %>%
```

```
separate(codes, into = c("new", "type", "sexage"), sep = "_")
```

country <chr>	year <int>	codes <chr>	new <chr>	type <chr>	sexage <chr>	n <int>
Afghanistan	1980	new_sp_m014	new	sp	m014	NA
Afghanistan	1981	new_sp_m014	new	sp	m014	NA
Afghanistan	1982	new_sp_m014	new	sp	m014	NA
Afghanistan	1983	new_sp_m014	new	sp	m014	NA
Afghanistan	1984	new_sp_m014	new	sp	m014	NA
Afghanistan	1985	new_sp_m014	new	sp	m014	NA
Afghanistan	1986	new_sp_m014	new	sp	m014	NA
Afghanistan	1987	new_sp_m014	new	sp	m014	NA
Afghanistan	1988	new_sp_m014	new	sp	m014	NA
Afghanistan	1989	new_sp_m014	new	sp	m014	NA



```
who %>%
```

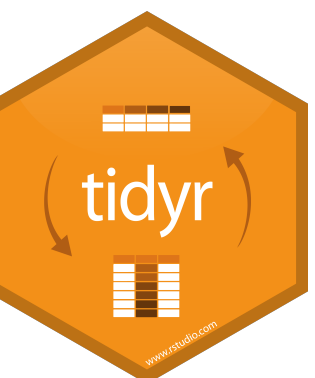
```
gather("codes", "n", 5:60) %>%
```

```
select(-iso2, -iso3) %>%
```

```
separate(codes, c("new", "type", "sexage"), sep = "_") %>%
```

```
select(-new)
```

country <chr>	year <int>	type <chr>	sexage <chr>	n <int>
Afghanistan	1980	sp	m014	NA
Afghanistan	1981	sp	m014	NA
Afghanistan	1982	sp	m014	NA
Afghanistan	1983	sp	m014	NA
Afghanistan	1984	sp	m014	NA
Afghanistan	1985	sp	m014	NA
Afghanistan	1986	sp	m014	NA
Afghanistan	1987	sp	m014	NA



separate()

Splits a column by dividing values at a specific character.

```
who %>%
```

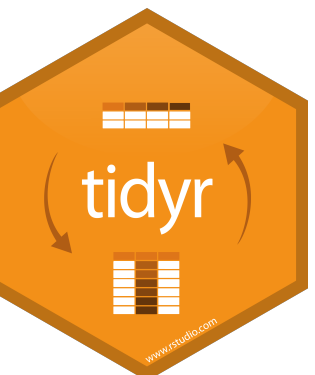
```
  gather("codes", "n", 5:60) %>%
```

```
  select(-iso2, -iso3) %>%
```

```
  separate(codes, c("new", "type", "sexage"), sep = c(4, 7))
```

locations to split at

(Split after 4th and 7th
characters)



Your Turn 6

Separate the sexage column into sex and age columns.

02:00

who %>%


gather("codes", "n", 5:60) %>%

select(-iso2, -iso3) %>%

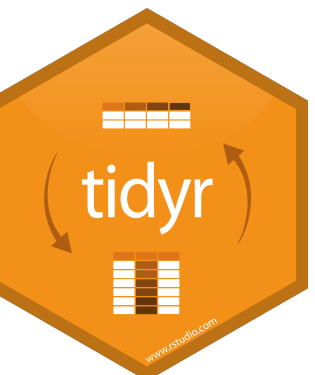
separate(codes, c("new", "type", "sexage"), sep = "_") %>%

select(-new) %>%

separate(sexage, into = c("sex", "age"), sep = 1)



	country <chr>	year <int>	type <chr>	sex <chr>	age <chr>	n <int>
1	Afghanistan	1980	sp	m0	14	NA
2	Afghanistan	1981	sp	m0	14	NA
3	Afghanistan	1982	sp	m0	14	NA
4	Afghanistan	1983	sp	m0	14	NA
5	Afghanistan	1984	sp	m0	14	NA
6	Afghanistan	1985	sp	m0	14	NA
7	Afghanistan	1986	sp	m0	14	NA



unite()



unite()

Unites columns into single column by combining cells.

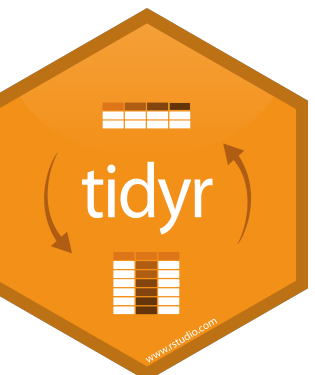
```
unite(data, col, ..., sep = "")
```

**data frame
to reshape**

**name of new
column to
make
(in quotes)**

**two or more
columns to
combine**

**separator to place
between elements in
new column**
(Defaults to an underscore)



Missing Values



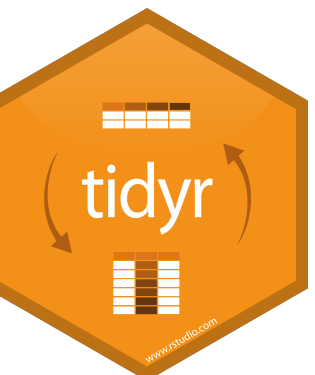
drop_na()

Drops rows that contain NA's in the specified columns.

```
drop_na(x, x2)
```

**data frame to
transform**

**column(s) to
screen for NA's**



drop_na()

Drops rows that contain NA's in the specified columns.

```
drop_na(x, x2)
```

^x
X

x1	x2
A	1
B	NA
C	NA
D	3
E	NA

→

x1	x2
A	1
D	3

who %>%

```
gather("codes", "n", 5:60) %>%
```

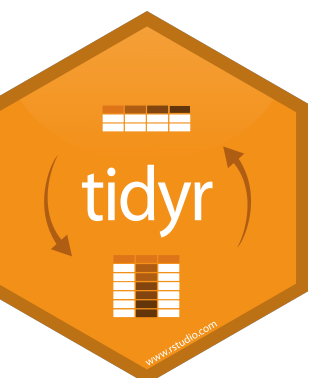
```
separate(codes, c("new", "type", "sexage"), sep = "_") %>%
```

```
select(-new, -iso2, -iso3) %>%
```

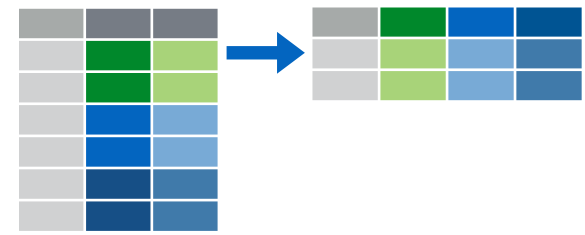
```
separate(sexage, c("sex", "age"), sep = 2) %>%
```

```
drop_na(n)
```

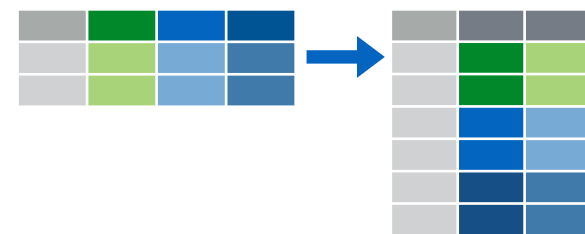
	country	year	type	sex	age	n
1	Afghanistan	1997	sp	m0	14	0
2	Afghanistan	1998	sp	m0	14	30
3	Afghanistan	1999	sp	m0	14	8
4	Afghanistan	2000	sp	m0	14	52
5	Afghanistan	2001	sp	m0	14	129
6	Afghanistan	2002	sp	m0	14	90
7	Afghanistan	2003	sp	m0	14	127
8	Afghanistan	2004	sp	m0	14	139
9	Afghanistan	2005	sp	m0	14	151



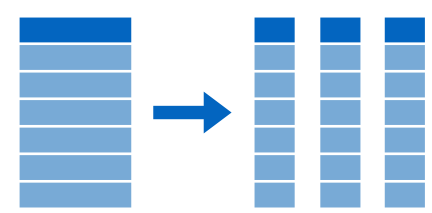
Recap



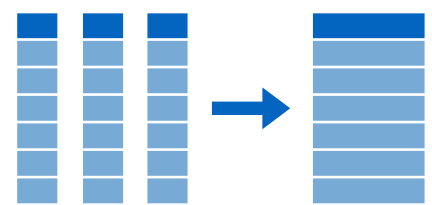
Move values into column names with **spread()**



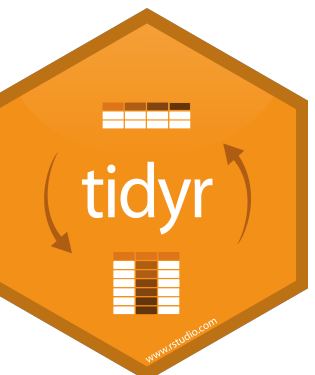
Move column names into values with **gather()**



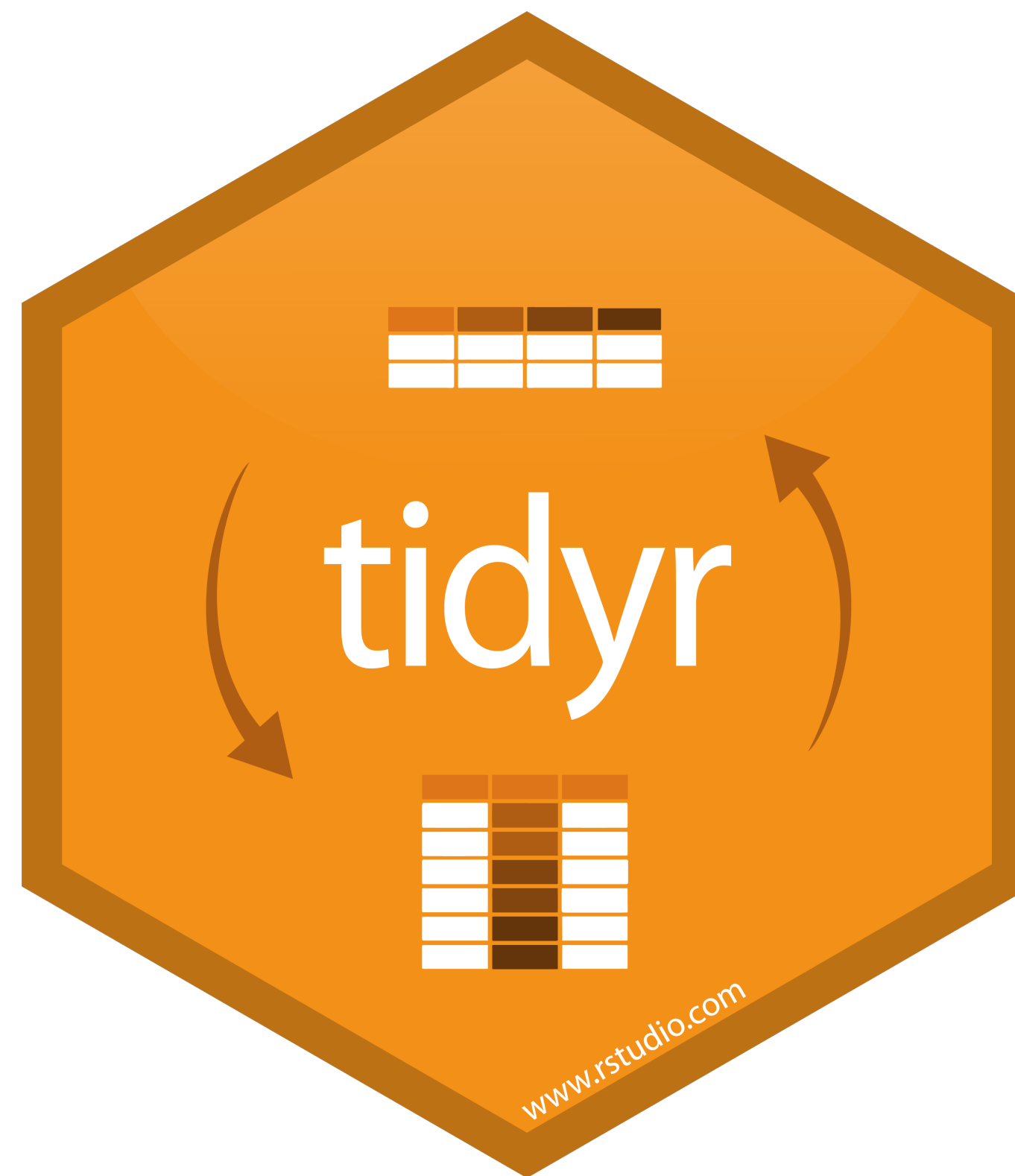
Split a column with **separate()** or **separate_rows()**



Unite columns with **unite()**



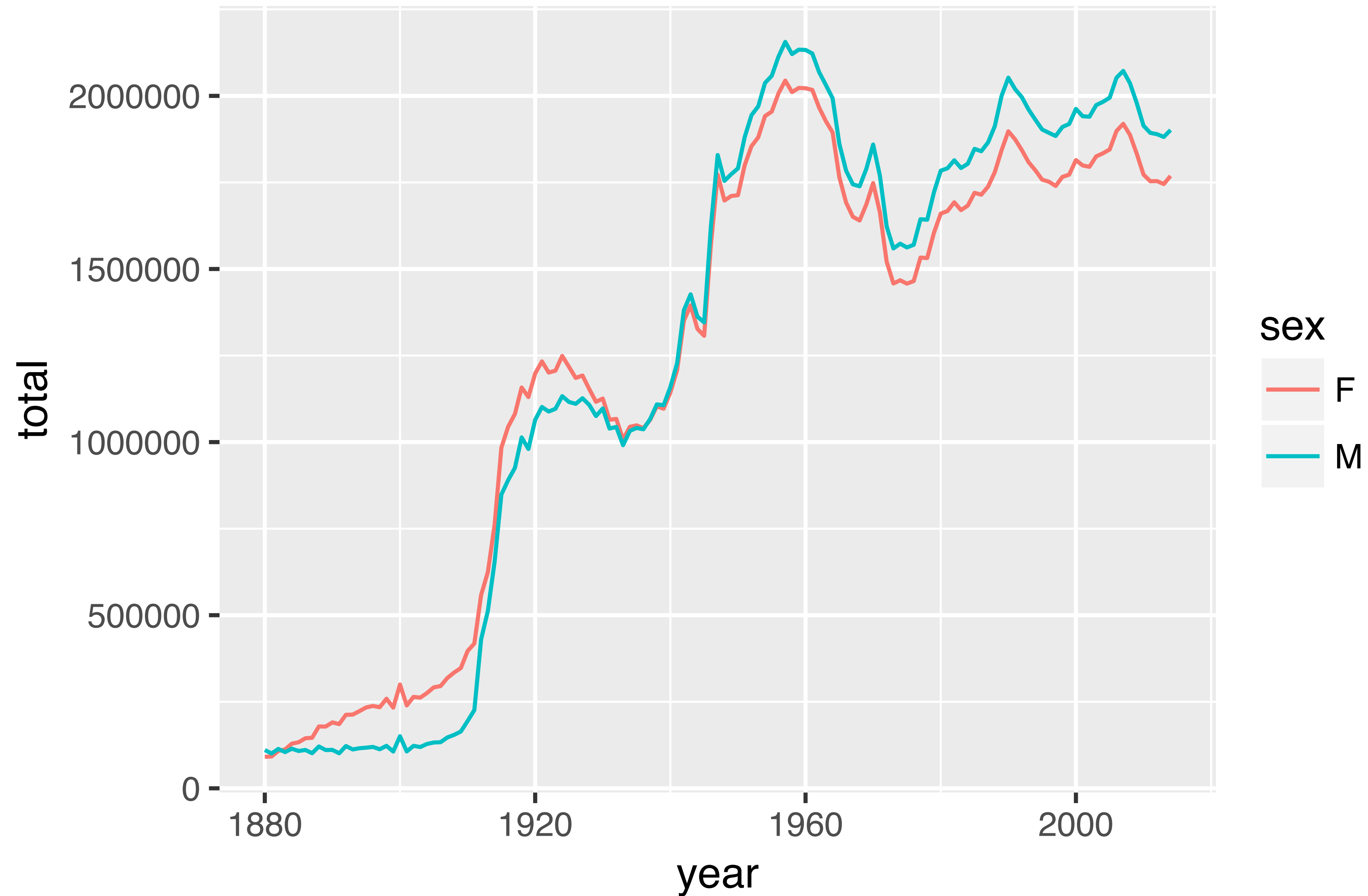
Tidy Data with



Reshaping Final Exam



Number of children by year and gender



Can we calculate the ratio of boys to girls?

```
babynames %>%  
  group_by(year, sex) %>%  
  summarise(n = sum(n))
```

	year	sex	n
	<dbl>	<chr>	<int>
1	1880	F	90993
2	1880	M	110491
3	1881	F	91954
4	1881	M	100745
5	1882	F	107850
6	1882	M	113688

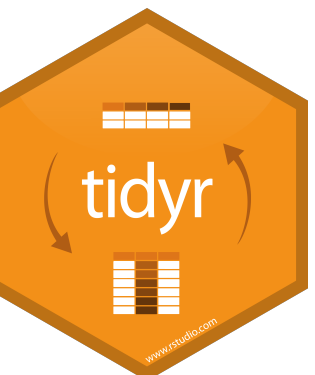


Can we calculate the ratio of boys to girls?

```
babynames %>%  
  group_by(year, sex) %>%  
  summarise(n = sum(n))
```

	year	sex	n
	<dbl>	<chr>	<int>
1	1880	F	90993
2	1880	M	110491
3	1881	F	91954
4	1881	M	100745
5	1882	F	107850
6	1882	M	113688

**Now
what?**

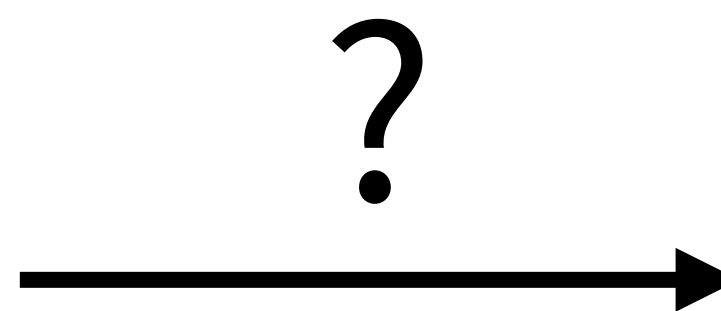


Can we calculate the ratio of boys to girls?

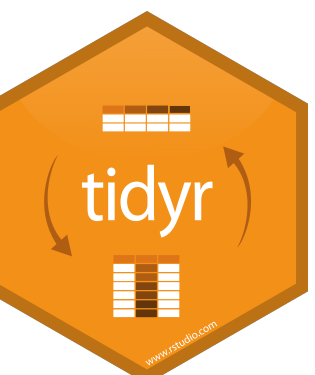
```
better_layout %>%
```

```
  mutate(percent_male = M / (M + F) * 100)
```

	year	sex	n
	<dbl>	<chr>	<int>
1	1880	F	90993
2	1880	M	110491
3	1881	F	91954
4	1881	M	100745
5	1882	F	107850
6	1882	M	113688



	year	F	M
	<dbl>	<int>	<int>
1	1880	90993	110491
2	1881	91954	100745
3	1882	107850	113688
4	1883	112321	104629
5	1884	129022	114445
6	1885	133055	107800



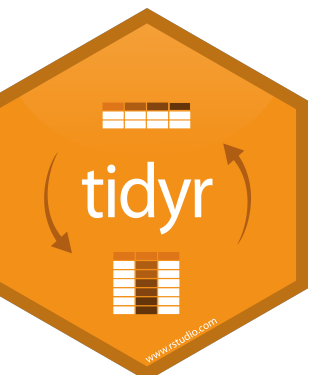
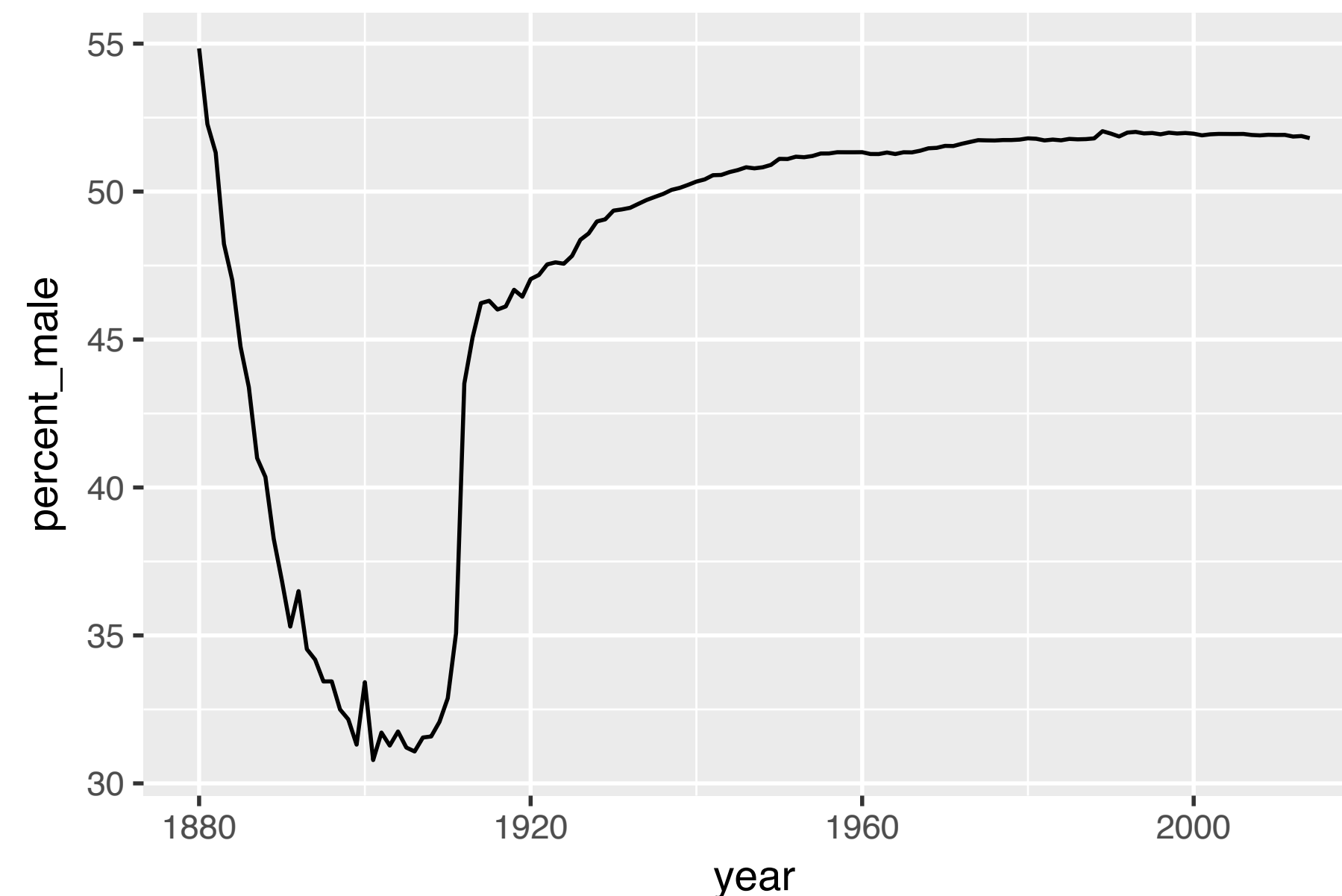
Your Turn 7

Reshape the layout of this data. Calculate the percent of male (or female) children by year. Then plot the percent over time.

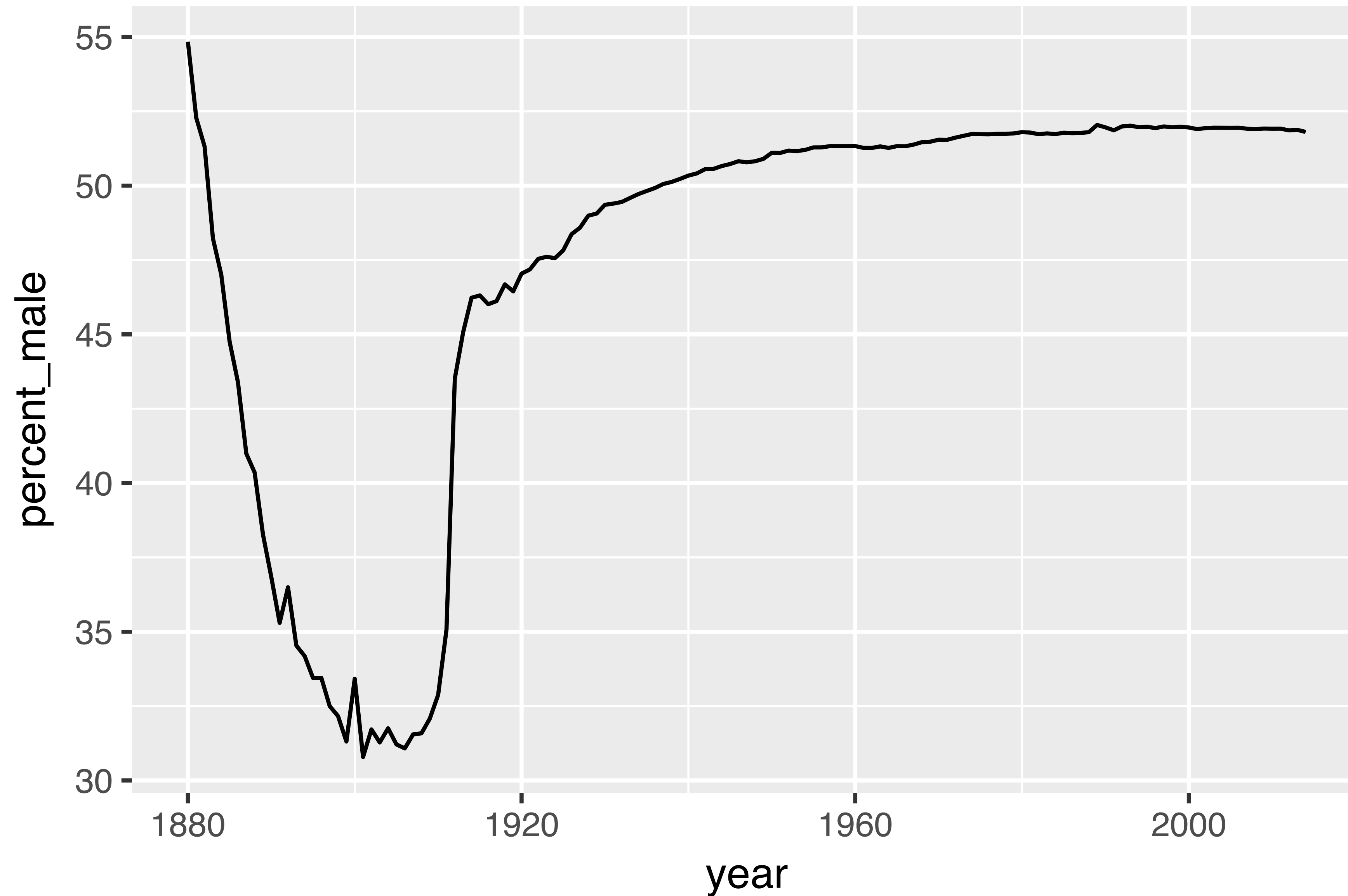
```
babynames %>%  
  group_by(year, sex) %>%  
  summarise(n = sum(n))
```

05:00


```
babynames %>%  
  group_by(year, sex) %>%  
  summarise(n = sum(n)) %>%  
  spread(sex, n) %>%  
  mutate(percent_male = M / (M + F) * 100) %>%  
  ggplot(aes(year, percent_male)) + geom_line()
```



Percent of children that are male by year



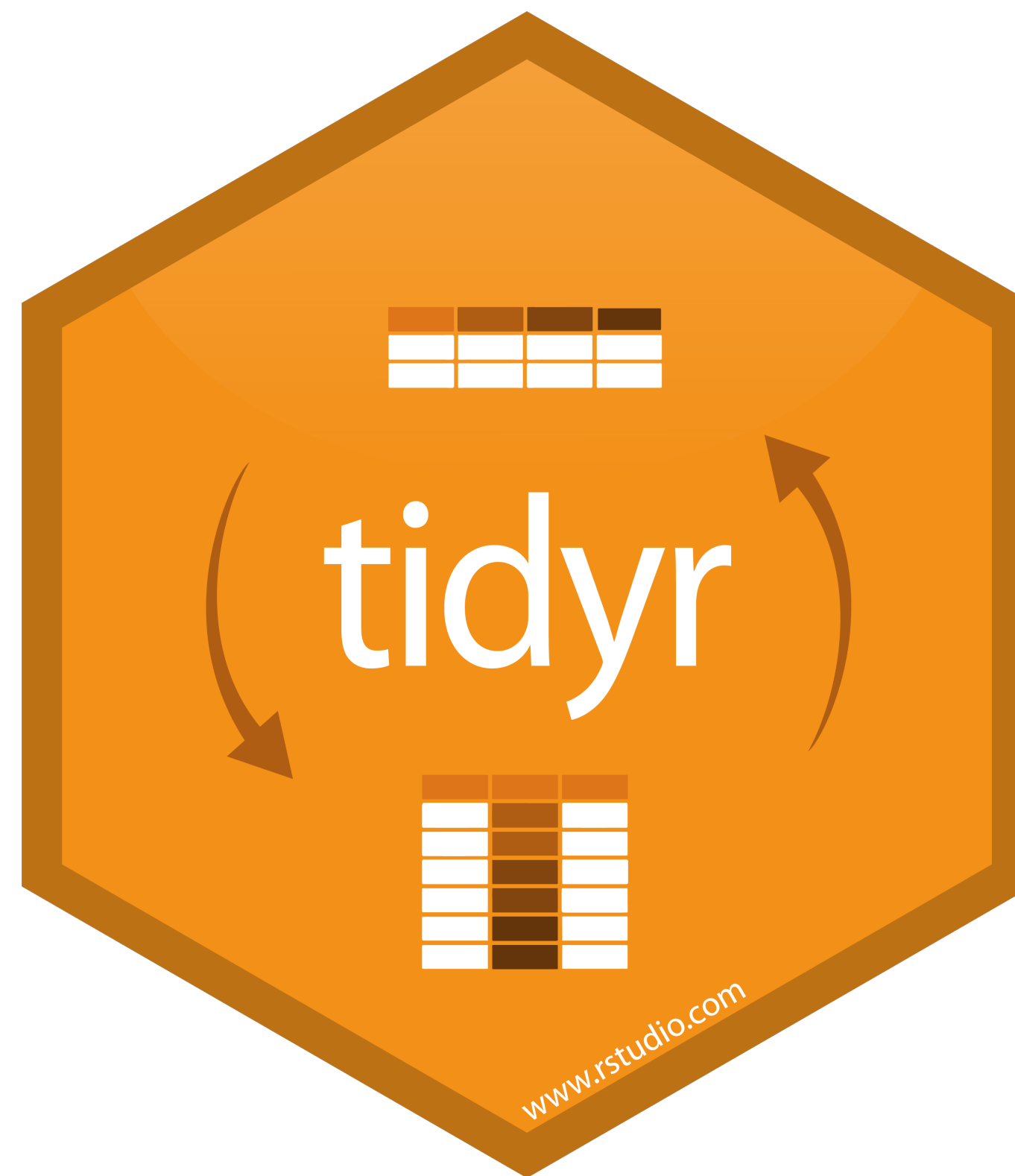
General advice

Describe what you want to do in an **equation**. Each variable in the equation should correspond to a variable in your data:

- "color by sex"
`color = sex`
- "calculate the proportion of males"
`prop male = number of males / number of females + number of males`



Tidy Data with



www.rstudio.com