

Week5CH12

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June 8, 2019

Tidy data

Required package

```
library(tidyverse)
```

```
## Warning: package 'tidyverse' was built under R version 3.5.3
## -- Attaching packages ----- tidyverse 1.2.1 --
## v ggplot2 3.2.0      v purrr   0.2.5
## v tibble  2.1.3      v dplyr   0.8.0.1
## v tidyr   0.8.1      v stringr 1.3.1
## v readr   1.1.1      v forcats 0.3.0
## Warning: package 'ggplot2' was built under R version 3.5.3
## Warning: package 'tibble' was built under R version 3.5.3
## Warning: package 'dplyr' was built under R version 3.5.3
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()     masks stats::lag()
```

```
library(tinytex)
```

```
## Warning: package 'tinytex' was built under R version 3.5.3
```

See table

```
table1
```

```
## # A tibble: 6 x 4
##   country      year  cases population
##   <chr>      <int> <int>      <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
```

See table2

```
table2
```

```
## # A tibble: 12 x 4
##   country      year type      count
##   <chr>      <int> <chr>      <int>
## 1 Afghanistan 1999 cases         745
## 2 Afghanistan 1999 population 19987071
## 3 Afghanistan 2000 cases         2666
## 4 Afghanistan 2000 population 20595360
```

```
## 5 Brazil      1999 cases      37737
## 6 Brazil      1999 population 172006362
## 7 Brazil      2000 cases      80488
## 8 Brazil      2000 population 174504898
## 9 China       1999 cases      212258
## 10 China      1999 population 1272915272
## 11 China      2000 cases      213766
## 12 China      2000 population 1280428583
```

See table3

```
table3
```

```
## # A tibble: 6 x 3
##   country      year rate
## * <chr>      <int> <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
```

See table4a

```
table4a
```

```
## # A tibble: 3 x 3
##   country      `1999` `2000`
## * <chr>      <int> <int>
## 1 Afghanistan    745    2666
## 2 Brazil        37737  80488
## 3 China         212258 213766
```

See table4b

```
table4b
```

```
## # A tibble: 3 x 3
##   country      `1999`      `2000`
## * <chr>      <int>      <int>
## 1 Afghanistan 19987071 20595360
## 2 Brazil      172006362 174504898
## 3 China       1272915272 1280428583
```

Diffrent ways to work with table 1. Compute rate per 10,000

```
table1 %>%
```

```
  mutate(rate = cases / population * 10000)
```

```
## # A tibble: 6 x 5
##   country      year cases population rate
##   <chr>      <int> <int>      <int> <dbl>
## 1 Afghanistan 1999    745  19987071 0.373
## 2 Afghanistan 2000   2666  20595360 1.29
## 3 Brazil      1999  37737  172006362 2.19
## 4 Brazil      2000  80488  174504898 4.61
## 5 China       1999 212258 1272915272 1.67
## 6 China       2000 213766 1280428583 1.67
```

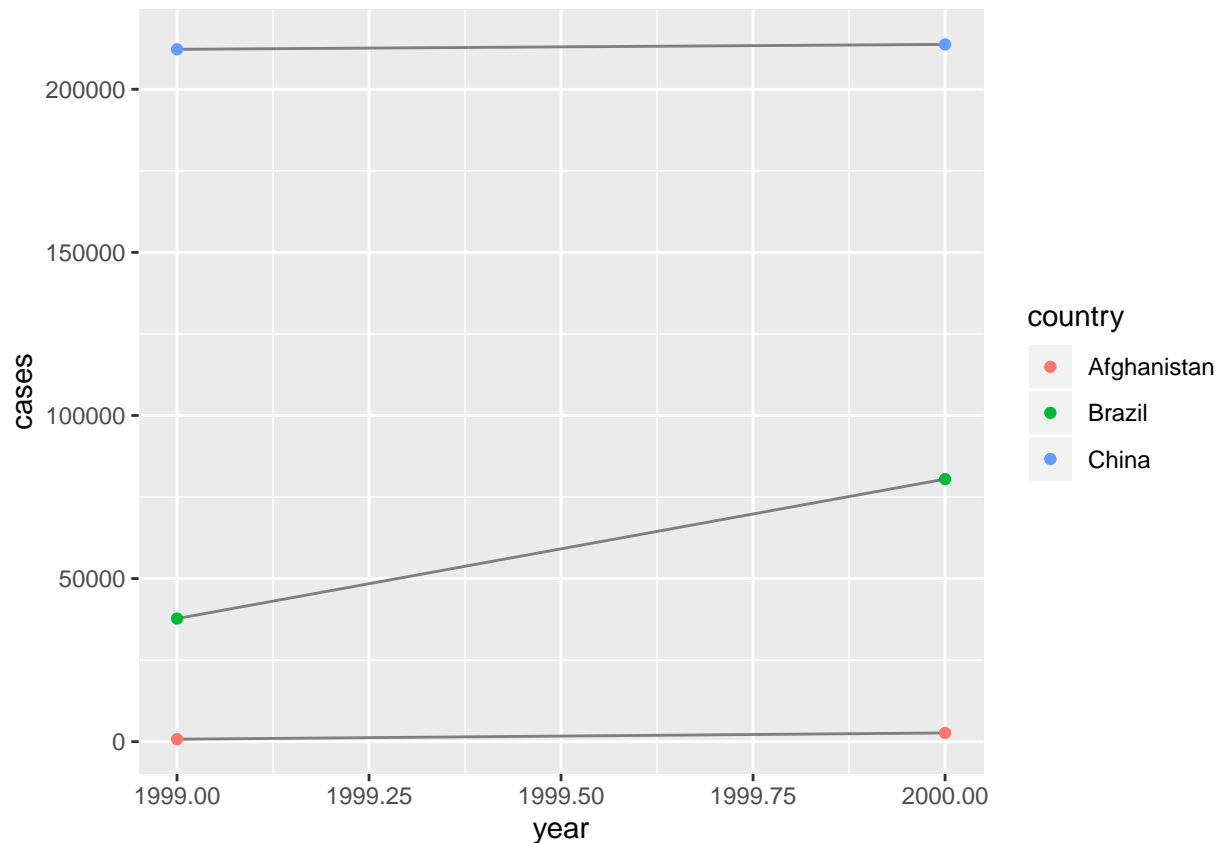
Compute cases per year

```
table1 %>%  
  count(year, wt = cases)
```

```
## # A tibble: 2 x 2  
##   year      n  
##   <int> <int>  
## 1  1999 250740  
## 2  2000 296920
```

Visualise changes over time

```
library(ggplot2)  
ggplot(table1, aes(year, cases)) +  
  geom_line(aes(group = country), colour = "grey50") +  
  geom_point(aes(colour = country))
```



Spreading and gathering

Gathering

some times in data the column names are not names of variables, but values of a variable. See table4

```
table4a
```

```
## # A tibble: 3 x 3  
##   country `1999` `2000`  
## * <chr>   <int> <int>
```

```
## 1 Afghanistan      745    2666
## 2 Brazil            37737   80488
## 3 China             212258  213766
```

Generate the call to `gather()` To tidy a dataset like this, we need to gather those columns into a new pair of variables.

```
table4a %>%
  gather(`1999`, `2000`, key = "year", value = "cases")
```

```
## # A tibble: 6 x 3
##   country    year  cases
##   <chr>      <chr> <int>
## 1 Afghanistan 1999     745
## 2 Brazil      1999   37737
## 3 China       1999  212258
## 4 Afghanistan 2000     2666
## 5 Brazil      2000   80488
## 6 China       2000  213766
```

Generate the call to `gather()`

```
table4b %>%
  gather(`1999`, `2000`, key = "year", value = "population")
```

```
## # A tibble: 6 x 3
##   country    year  population
##   <chr>      <chr>      <int>
## 1 Afghanistan 1999   19987071
## 2 Brazil      1999  172006362
## 3 China       1999 1272915272
## 4 Afghanistan 2000   20595360
## 5 Brazil      2000  174504898
## 6 China       2000 1280428583
```

Combine the tidied versions of table4a and table4b into a single tibble,

```
tidy4a <- table4a %>%
  gather(`1999`, `2000`, key = "year", value = "cases")
tidy4b <- table4b %>%
  gather(`1999`, `2000`, key = "year", value = "population")
left_join(tidy4a, tidy4b)
```

```
## Joining, by = c("country", "year")
```

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

Separate

In spreading opposite to what is done gathering.

```
table2
```

```
## # A tibble: 12 x 4
##   country      year type      count
##   <chr>      <int> <chr>    <int>
## 1 Afghanistan 1999 cases      745
## 2 Afghanistan 1999 population 19987071
## 3 Afghanistan 2000 cases      2666
## 4 Afghanistan 2000 population 20595360
## 5 Brazil      1999 cases      37737
## 6 Brazil      1999 population 172006362
## 7 Brazil      2000 cases      80488
## 8 Brazil      2000 population 174504898
## 9 China       1999 cases      212258
## 10 China      1999 population 1272915272
## 11 China      2000 cases      213766
## 12 China      2000 population 1280428583
```

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

```
## # A tibble: 6 x 4
##   country      year cases population
##   <chr>      <int> <int>    <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
```

Separating and uniting

Separate

```
table3
```

```
## # A tibble: 6 x 3
##   country      year rate
##   * <chr>      <int> <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
```

Function `separate()` pulls apart one column into multiple columns, by splitting wherever a separator character appears

```
table3 %>%
  separate(rate, into = c("cases", "population"))
```

```
## # A tibble: 6 x 4
##   country      year cases population
##   <chr>      <int> <chr>    <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
```

rate column contains both cases and population variables which need to be split it into two variables.

```
table3 %>%
  separate(rate, into = c("cases", "population"), sep = "/")
```

```
## # A tibble: 6 x 4
##   country      year cases population
##   <chr>      <int> <chr>   <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
```

separate() split the values of rate at the forward slash characters

```
table3 %>%
  separate(rate, into = c("cases", "population"), convert = TRUE)
```

```
## # A tibble: 6 x 4
##   country      year cases population
##   <chr>      <int> <int>      <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
```

```
table3 %>%
  separate(year, into = c("century", "year"), sep = 2)
```

```
## # A tibble: 6 x 4
##   country      century year  rate
##   <chr>      <chr>   <chr> <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
```

Unite

```
table5 %>%
  unite(new, century, year)
```

```
## # A tibble: 6 x 3
##   country      new  rate
```

```
##   <chr>      <chr> <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
```

```
table5 %>%
  unite(new, century, year, sep = "")
```

```
## # A tibble: 6 x 3
##   country    new    rate
##   <chr>      <chr> <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
```

Missing values

A value can be missing in one of two possible ways. 1. Explicitly 2. Implicitly

```
stocks <- tibble(
  year = c(2015, 2015, 2015, 2015, 2016, 2016, 2016),
  qtr  = c( 1,   2,   3,   4,   2,   3,   4),
  return = c(1.88, 0.59, 0.35, NA, 0.92, 0.17, 2.66)
)
```

Make the implicit missing value explicit by putting years in the columns

```
stocks %>%
  spread(year, return)
```

```
## # A tibble: 4 x 3
##   qtr `2015` `2016`
##   <dbl> <dbl> <dbl>
## 1     1  1.88  NA
## 2     2  0.59  0.92
## 3     3  0.35  0.17
## 4     4  NA    2.66
```

Turn explicit missing values implicit by setting na.rm = TRUE

```
stocks %>%
  spread(year, return) %>%
  gather(year, return, `2015`:`2016`, na.rm = TRUE)
```

```
## # A tibble: 6 x 3
##   qtr year return
##   <dbl> <chr> <dbl>
## 1     1 2015  1.88
## 2     2 2015  0.59
## 3     3 2015  0.35
## 4     2 2016  0.92
## 5     3 2016  0.17
```

```
## 6      4 2016      2.66
```

Make missing values explicit in tidy data by using function complete

```
stocks %>%  
  complete(year, qtr)
```

```
## # A tibble: 8 x 3  
##   year   qtr return  
##   <dbl> <dbl> <dbl>  
## 1  2015     1   1.88  
## 2  2015     2   0.59  
## 3  2015     3   0.35  
## 4  2015     4    NA  
## 5  2016     1    NA  
## 6  2016     2   0.92  
## 7  2016     3   0.17  
## 8  2016     4   2.66
```

```
treatment <- tribble(  
  ~ person,      ~ treatment, ~response,  
  "Derrick Whitmore", 1,      7,  
  NA,                2,      10,  
  NA,                3,      9,  
  "Katherine Burke", 1,      4  
)
```

```
treatment %>%  
  fill(person)
```

```
## # A tibble: 4 x 3  
##   person      treatment response  
##   <chr>      <dbl>    <dbl>  
## 1 Derrick Whitmore      1      7  
## 2 Derrick Whitmore      2     10  
## 3 Derrick Whitmore      3      9  
## 4 Katherine Burke       1      4
```

Case Study

```
who
```

```
## # A tibble: 7,240 x 60  
##   country iso2 iso3 year new_sp_m014 new_sp_m1524 new_sp_m2534  
##   <chr>   <chr> <chr> <int>    <int>    <int>    <int>  
## 1 Afghan~ AF   AFG  1980      NA      NA      NA  
## 2 Afghan~ AF   AFG  1981      NA      NA      NA  
## 3 Afghan~ AF   AFG  1982      NA      NA      NA  
## 4 Afghan~ AF   AFG  1983      NA      NA      NA  
## 5 Afghan~ AF   AFG  1984      NA      NA      NA  
## 6 Afghan~ AF   AFG  1985      NA      NA      NA  
## 7 Afghan~ AF   AFG  1986      NA      NA      NA  
## 8 Afghan~ AF   AFG  1987      NA      NA      NA  
## 9 Afghan~ AF   AFG  1988      NA      NA      NA  
## 10 Afghan~ AF   AFG  1989      NA      NA      NA  
## # ... with 7,230 more rows, and 53 more variables: new_sp_m3544 <int>,
```



```
## # new_sp_m4554 <int>, new_sp_m5564 <int>, new_sp_m65 <int>,
## # new_sp_f014 <int>, new_sp_f1524 <int>, new_sp_f2534 <int>,
## # new_sp_f3544 <int>, new_sp_f4554 <int>, new_sp_f5564 <int>,
## # new_sp_f65 <int>, new_sn_m014 <int>, new_sn_m1524 <int>,
## # new_sn_m2534 <int>, new_sn_m3544 <int>, new_sn_m4554 <int>,
## # new_sn_m5564 <int>, new_sn_m65 <int>, new_sn_f014 <int>,
## # new_sn_f1524 <int>, new_sn_f2534 <int>, new_sn_f3544 <int>,
## # new_sn_f4554 <int>, new_sn_f5564 <int>, new_sn_f65 <int>,
## # new_ep_m014 <int>, new_ep_m1524 <int>, new_ep_m2534 <int>,
## # new_ep_m3544 <int>, new_ep_m4554 <int>, new_ep_m5564 <int>,
## # new_ep_m65 <int>, new_ep_f014 <int>, new_ep_f1524 <int>,
## # new_ep_f2534 <int>, new_ep_f3544 <int>, new_ep_f4554 <int>,
## # new_ep_f5564 <int>, new_ep_f65 <int>, newrel_m014 <int>,
## # newrel_m1524 <int>, newrel_m2534 <int>, newrel_m3544 <int>,
## # newrel_m4554 <int>, newrel_m5564 <int>, newrel_m65 <int>,
## # newrel_f014 <int>, newrel_f1524 <int>, newrel_f2534 <int>,
## # newrel_f3544 <int>, newrel_f4554 <int>, newrel_f5564 <int>,
## # newrel_f65 <int>
```

Focus on the values that are present.

```
who1 <- who %>%
  gather(new_sp_m014:newrel_f65, key = "key", value = "cases", na.rm = TRUE)
who1
```

```
## # A tibble: 76,046 x 6
##   country    iso2 iso3   year key      cases
##   <chr>      <chr> <chr> <int> <chr>    <int>
## 1 Afghanistan AF    AFG   1997 new_sp_m014    0
## 2 Afghanistan AF    AFG   1998 new_sp_m014   30
## 3 Afghanistan AF    AFG   1999 new_sp_m014    8
## 4 Afghanistan AF    AFG   2000 new_sp_m014   52
## 5 Afghanistan AF    AFG   2001 new_sp_m014  129
## 6 Afghanistan AF    AFG   2002 new_sp_m014   90
## 7 Afghanistan AF    AFG   2003 new_sp_m014  127
## 8 Afghanistan AF    AFG   2004 new_sp_m014  139
## 9 Afghanistan AF    AFG   2005 new_sp_m014  151
## 10 Afghanistan AF    AFG   2006 new_sp_m014  193
## # ... with 76,036 more rows
```

Count using key.

```
who1 %>%
  count(key)
```

```
## # A tibble: 56 x 2
##   key      n
##   <chr>    <int>
## 1 new_ep_f014  1032
## 2 new_ep_f1524 1021
## 3 new_ep_f2534 1021
## 4 new_ep_f3544 1021
## 5 new_ep_f4554 1017
## 6 new_ep_f5564 1017
## 7 new_ep_f65   1014
## 8 new_ep_m014  1038
## 9 new_ep_m1524 1026
```

```
## 10 new_ep_m2534 1020
## # ... with 46 more rows
```

Makes all variable names consistent.

```
who2 <- who1 %>%
  mutate(key = stringr::str_replace(key, "newrel", "new_rel"))
who2
```

```
## # A tibble: 76,046 x 6
##   country    iso2 iso3  year key      cases
##   <chr>      <chr> <chr> <int> <chr>    <int>
## 1 Afghanistan AF    AFG  1997 new_sp_m014 0
## 2 Afghanistan AF    AFG  1998 new_sp_m014 30
## 3 Afghanistan AF    AFG  1999 new_sp_m014 8
## 4 Afghanistan AF    AFG  2000 new_sp_m014 52
## 5 Afghanistan AF    AFG  2001 new_sp_m014 129
## 6 Afghanistan AF    AFG  2002 new_sp_m014 90
## 7 Afghanistan AF    AFG  2003 new_sp_m014 127
## 8 Afghanistan AF    AFG  2004 new_sp_m014 139
## 9 Afghanistan AF    AFG  2005 new_sp_m014 151
## 10 Afghanistan AF    AFG  2006 new_sp_m014 193
## # ... with 76,036 more rows
```

Separate the values in each code with two passes using function separate()

```
who3 <- who2 %>%
  separate(key, c("new", "type", "sexage"), sep = "_")
who3
```

```
## # A tibble: 76,046 x 8
##   country    iso2 iso3  year new  type sexage cases
##   <chr>      <chr> <chr> <int> <chr> <chr> <chr>    <int>
## 1 Afghanistan AF    AFG  1997 new  sp   m014 0
## 2 Afghanistan AF    AFG  1998 new  sp   m014 30
## 3 Afghanistan AF    AFG  1999 new  sp   m014 8
## 4 Afghanistan AF    AFG  2000 new  sp   m014 52
## 5 Afghanistan AF    AFG  2001 new  sp   m014 129
## 6 Afghanistan AF    AFG  2002 new  sp   m014 90
## 7 Afghanistan AF    AFG  2003 new  sp   m014 127
## 8 Afghanistan AF    AFG  2004 new  sp   m014 139
## 9 Afghanistan AF    AFG  2005 new  sp   m014 151
## 10 Afghanistan AF    AFG  2006 new  sp   m014 193
## # ... with 76,036 more rows
```

```
who3 %>%
  count(new)
```

```
## # A tibble: 1 x 2
##   new      n
##   <chr> <int>
## 1 new  76046
```

```
who4 <- who3 %>%
  select(-new, -iso2, -iso3)
```

Separate sexage into sex and age by splitting after the first character.

```
who5 <- who4 %>%
  separate(sexage, c("sex", "age"), sep = 1)
who5
```

```
## # A tibble: 76,046 x 6
##   country      year type  sex  age  cases
##   <chr>        <int> <chr> <chr> <chr> <int>
## 1 Afghanistan 1997 sp    m    014     0
## 2 Afghanistan 1998 sp    m    014    30
## 3 Afghanistan 1999 sp    m    014     8
## 4 Afghanistan 2000 sp    m    014    52
## 5 Afghanistan 2001 sp    m    014   129
## 6 Afghanistan 2002 sp    m    014    90
## 7 Afghanistan 2003 sp    m    014   127
## 8 Afghanistan 2004 sp    m    014   139
## 9 Afghanistan 2005 sp    m    014   151
## 10 Afghanistan 2006 sp    m    014   193
## # ... with 76,036 more rows
```

who dataset is now tidy!

```
who %>%
  gather(key, value, new_sp_m014:newrel_f65, na.rm = TRUE) %>%
  mutate(key = stringr::str_replace(key, "newrel", "new_rel")) %>%
  separate(key, c("new", "var", "sexage")) %>%
  select(-new, -iso2, -iso3) %>%
  separate(sexage, c("sex", "age"), sep = 1)
```

```
## # A tibble: 76,046 x 6
##   country      year var  sex  age  value
##   <chr>        <int> <chr> <chr> <chr> <int>
## 1 Afghanistan 1997 sp    m    014     0
## 2 Afghanistan 1998 sp    m    014    30
## 3 Afghanistan 1999 sp    m    014     8
## 4 Afghanistan 2000 sp    m    014    52
## 5 Afghanistan 2001 sp    m    014   129
## 6 Afghanistan 2002 sp    m    014    90
## 7 Afghanistan 2003 sp    m    014   127
## 8 Afghanistan 2004 sp    m    014   139
## 9 Afghanistan 2005 sp    m    014   151
## 10 Afghanistan 2006 sp    m    014   193
## # ... with 76,036 more rows
```