

# Introduction to the Tidyverse

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# The tidyverse help

- <https://ggplot2.tidyverse.org/>
- <https://www.tidyverse.org/learn/>

# Using packages in R

```
install.packages('name')
```

- Downloads the files to your computer
  - **Do this once per computer**
- 

```
# How to use the package  
library('name')
```

- Loads the package
- **Do this once per session**

coherent system of packages for data manipulation

# The tidyverse

# The tidyverse

## Tidyverse?

- The tidyverse is an opinionated collection of R packages designed for data science.
- All packages share an underlying design philosophy, grammar, and data structures.
- The tidyverse makes data science faster, easier and more fun



"Tidyverse package"

## Task

- The tidyverse package is a shortcut for installing and loading all the key tidyverse packages

Install the `tidyverse` package

## Solution

```
install.packages('tidyverse')
```

```
install.packages("ggplot2")  
install.packages("dplyr")  
install.packages("tidyr")  
install.packages("readr")  
install.packages("purrr")  
install.packages("tibble")  
install.packages("stringr")  
install.packages("forcats")  
install.packages("lubridate")  
install.packages("hms")  
install.packages("DBI")  
install.packages("haven")  
install.packages("httr")  
install.packages("jsonlite")  
install.packages("readxl")  
install.packages("rvest")  
install.packages("xml2")  
install.packages("modelr")  
install.packages("broom")
```

# Data frames and tibbles

- Data frames are the most common kind of data objects; used for rectangular data (like spreadsheets)
- Data frames: R's native data object
- Tibbles (tbl): a fancier enhanced kind of data frame
- (You really won't notice a difference in this class)



# Vectors

- Vectors are a list of values of the same type (all text, or all numbers, etc.)
- **Make them with c():**

```
c(1, 4, 2, 5, 7)
```

```
## [1] 1 4 2 5 7
```

- You'll usually want to assign them to something: `:::{.task}` Create a vector `c(1, 4, 2, 5, 7)` and assign it `neat_numbers` object name `:::`

## Solution

```
neat_numbers <- c(1, 4, 2, 5, 7)
```

# Packages for importing data



Work with plain text data

```
my_data <-  
read_csv("file.csv")
```



Work with Excel files

```
my_data <-  
read_excel("file.xlsx")
```



Work with Stata, SPSS, and  
SAS data

```
my_data <-  
read_stata("file.dta")
```

- Hint use `read_csv` after loading `tidyverse`

Read in the `birthweight.csv` file and assign it to `bw_df`

## Solution

```
library(tidyverse)
bw_df <- read_csv('data/birthweight.csv')
```

Transform the data with dplyr

# The tidyverse: dplyr

# Dataset to use

- Excerpt of the `gapminder` data on life expectancy, GDP per capita, and population by country.
- The data frame `gapminder` has 1704 rows and 6 variables
  - Country -factor with 142 levels
  - Continent - factor with 5 levels
  - Year - ranges from 1952 to 2007 in increments of 5 years
  - lifeExp - life expectancy at birth, in years
  - Pop - population
  - gdpPercap - GDP per capita (US\$, inflation-adjusted)
- **Task: Install and load the gapminder package**



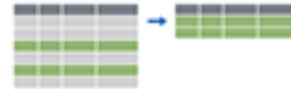
```
library(gapminder)
glimpse(gapminder)
```

```
## Rows: 1,704
## Columns: 7
## $ country   <fct> Afghanistan, Afghanistan, Afghanistan, Afghanistan, Afghanistan
## $ continent <fct> Asia, Asia, Asia, Asia, Asia, Asia, Asia, Asia, Asia, Asia, Asia, Asia
## $ year      <int> 1952, 1957, 1962, 1967, 1972, 1977, 1982, 1987, 1992, 1997, 2002, 2007
## $ lifeExp   <dbl> 28.801, 30.332, 31.997, 34.020, 36.088, 38.438, 39.854, 40.822, 41.674
## $ pop       <int> 8425333, 9240934, 10267083, 11537966, 13079460, 14880372, 12881816, 13
## $ gdpPercap <dbl> 779.4453, 820.8530, 853.1007, 836.1971, 739.9811, 786.1134, 978.0114,
## $ year_cat  <chr> "Before 1980", "Before 1980", "Before 1980", "Before 1980", "Before 19
```

# dplyr: verbs for manipulating data

---

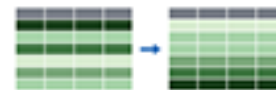
Extract rows with `filter()`



Extract columns with `select()`



Arrange/sort rows with `arrange()`



Make new columns with `mutate()`



Make group summaries with  
`group_by() %>% summarize()`



Select a subset of variables

```
select(data, vars(...))
```

- DATA = Data frame to transform
- ... = variables to select

# dplyr: select

# Our data

```
head(gapminder)
```

```
## # A tibble: 6 x 7
##   country      continent  year lifeExp      pop gdpPercap year_cat
##   <fct>        <fct>    <int>   <dbl>    <int>   <dbl> <chr>
## 1 Afghanistan Asia      1952    28.8  8425333    779. Before 1980
## 2 Afghanistan Asia      1957    30.3  9240934    821. Before 1980
## 3 Afghanistan Asia      1962    32.0 10267083    853. Before 1980
## 4 Afghanistan Asia      1967    34.0 11537966    836. Before 1980
## 5 Afghanistan Asia      1972    36.1 13079460    740. Before 1980
## 6 Afghanistan Asia      1977    38.4 14880372    786. Before 1980
```

# Subset country and life expectancy and year variables only

```
select(.data = gapminder, c(country, year , lifeExp))
```

```
## # A tibble: 1,704 x 3
##   country      year lifeExp
##   <fct>      <int>   <dbl>
## 1 Afghanistan  1952    28.8
## 2 Afghanistan  1957    30.3
## 3 Afghanistan  1962    32.0
## 4 Afghanistan  1967    34.0
## 5 Afghanistan  1972    36.1
## 6 Afghanistan  1977    38.4
## 7 Afghanistan  1982    39.9
## 8 Afghanistan  1987    40.8
## 9 Afghanistan  1992    41.7
## 10 Afghanistan 1997    41.8
## # ... with 1,694 more rows
```

# Task #1: Select

Use `select()`

- subset `country`, `year`, `gdpPercap`, and `pop` variables only
- create a new object called `population_gdp` assigned to subset data

## Select() solution

```
population_gdp <- select(gapminder, c(country, year, gdpPercap, pop))
population_gdp
```

```
## # A tibble: 1,704 x 4
##   country      year gdpPercap      pop
##   <fct>      <int>      <dbl>    <int>
## 1 Afghanistan 1952        779.  8425333
## 2 Afghanistan 1957        821.  9240934
## 3 Afghanistan 1962        853. 10267083
## 4 Afghanistan 1967        836. 11537966
## 5 Afghanistan 1972        740. 13079460
## 6 Afghanistan 1977        786. 14880372
## 7 Afghanistan 1982        978. 12881816
## 8 Afghanistan 1987        852. 13867957
## 9 Afghanistan 1992        649. 16317921
## 10 Afghanistan 1997        635. 22227415
## # ... with 1,694 more rows
```

Pick rows: filter()

**dplyr: filter**



# Extract rows that meet some sort of test

```
filter(.data = DATA, ...)
```

- `DATA` = Data frame to transform
- `...` = One or more tests
- `filter()` returns each row for which the test is TRUE

# Our data

```
head(gapminder)
```

```
## # A tibble: 6 x 7
##   country      continent  year lifeExp      pop gdpPercap year_cat
##   <fct>        <fct>    <int>  <dbl>    <int>    <dbl> <chr>
## 1 Afghanistan Asia      1952   28.8  8425333    779. Before 1980
## 2 Afghanistan Asia      1957   30.3  9240934    821. Before 1980
## 3 Afghanistan Asia      1962   32.0 10267083    853. Before 1980
## 4 Afghanistan Asia      1967   34.0 11537966    836. Before 1980
## 5 Afghanistan Asia      1972   36.1 13079460    740. Before 1980
## 6 Afghanistan Asia      1977   38.4 14880372    786. Before 1980
```

# Filtering only Tanzania data

- NB: We use == which tests if equal
  - One = sets an argument.

```
filter(.data = gapminder, country == "Tanzania")
```

```
## # A tibble: 12 x 7
##   country continent year lifeExp      pop gdpPercap year_cat
##   <fct>    <fct>    <int>   <dbl>    <int>   <dbl> <chr>
## 1 Tanzania Africa    1952   41.2  8322925   717. Before 1980
## 2 Tanzania Africa    1957   43.0  9452826   699. Before 1980
## 3 Tanzania Africa    1962   44.2 10863958   722. Before 1980
## 4 Tanzania Africa    1967   45.8 12607312   848. Before 1980
## 5 Tanzania Africa    1972   47.6 14706593   916. Before 1980
## 6 Tanzania Africa    1977   49.9 17129565   962. Before 1980
## 7 Tanzania Africa    1982   50.6 19844382   874. 1980-2000
## 8 Tanzania Africa    1987   51.5 23040630   832. 1980-2000
## 9 Tanzania Africa    1992   50.4 26605473   826. 1980-2000
## 10 Tanzania Africa    1997   48.5 30686889   789. 1980-2000
## 11 Tanzania Africa    2002   49.7 34593779   899. After 200
## 12 Tanzania Africa    2007   52.5 38139640  1107. After 200
```

# Logical tests

---

Test	Meaning	Test	Meaning
<code>x &lt; y</code>	Less than	<code>x %in% y</code>	In (group membership)
<code>x &gt; y</code>	Greater than	<code>is.na(x)</code>	Is missing
<code>==</code>	Equal to	<code>!is.na(x)</code>	Is not missing
<code>x &lt;= y</code>	Less than or equal to		
<code>x &gt;= y</code>	Greater than or equal to		
<code>x != y</code>	Not equal to		

---

# Task #1: Filtering

Use filter() and logical tests to show...

- The data for Kenya
- All data for countries in Oceania **Hint: Oceania is not a country**
- Rows where the life expectancy is less than 30

## filter the data for Kenya

```
filter(gapminder, country == "Kenya")
```

```
## # A tibble: 12 x 7
##   country continent  year lifeExp      pop gdpPercap year_cat
##   <fct>    <fct>    <int>   <dbl>    <int>   <dbl> <chr>
## 1 Kenya  Africa     1952   42.3  6464046    854. Before 1980
## 2 Kenya  Africa     1957   44.7  7454779    944. Before 1980
## 3 Kenya  Africa     1962   47.9  8678557    897. Before 1980
## 4 Kenya  Africa     1967   50.7 10191512   1057. Before 1980
## 5 Kenya  Africa     1972   53.6 12044785   1222. Before 1980
## 6 Kenya  Africa     1977   56.2 14500404   1268. Before 1980
## 7 Kenya  Africa     1982   58.8 17661452   1348. 1980-2000
## 8 Kenya  Africa     1987   59.3 21198082   1362. 1980-2000
## 9 Kenya  Africa     1992   59.3 25020539   1342. 1980-2000
## 10 Kenya Africa     1997   54.4 28263827   1360. 1980-2000
## 11 Kenya Africa     2002   51.0 31386842   1288. After 200
## 12 Kenya Africa     2007   54.1 35610177   1463. After 200
```

## filter all data for countries in Oceania

```
filter(gapminder, continent == "Oceania")
```

```
## # A tibble: 24 x 7
##   country    continent  year lifeExp      pop gdpPercap year_cat
##   <fct>      <fct>    <int>  <dbl>    <int>   <dbl> <chr>
## 1 Australia Oceania    1952   69.1  8691212  10040. Before 1980
## 2 Australia Oceania    1957   70.3  9712569  10950. Before 1980
## 3 Australia Oceania    1962   70.9 10794968  12217. Before 1980
## 4 Australia Oceania    1967   71.1 11872264  14526. Before 1980
## 5 Australia Oceania    1972   71.9 13177000  16789. Before 1980
## 6 Australia Oceania    1977   73.5 14074100  18334. Before 1980
## 7 Australia Oceania    1982   74.7 15184200  19477. 1980-2000
## 8 Australia Oceania    1987   76.3 16257249  21889. 1980-2000
## 9 Australia Oceania    1992   77.6 17481977  23425. 1980-2000
## 10 Australia Oceania    1997   78.8 18565243  26998. 1980-2000
## # ... with 14 more rows
```

filter rows where the life expectancy is less than 30

```
filter(gapminder, lifeExp <30)
```

```
## # A tibble: 2 x 7
##   country    continent  year lifeExp      pop gdpPercap year_cat
##   <fct>      <fct>    <int>   <dbl>   <int>   <dbl> <chr>
## 1 Afghanistan Asia      1952    28.8  8425333    779. Before 1980
## 2 Rwanda     Africa    1992    23.6  7290203    737. 1980-2000
```



# Common mistakes

- Using = instead of ==
- Forgetting quote

*## Wrong*

```
filter(gapminder, country = "Kenya")  
filter(gapminder, country = Kenya)
```

*## Correct*

```
filter(gapminder, country == "Kenya")
```

# filter() with multiple conditions

- Extract rows that meet every test
- Extract for Tanzania before year 2000

```
filter(.data = gapminder, country == "Tanzania" , year<2000)
```

```
## # A tibble: 10 x 7
##   country continent  year lifeExp      pop gdpPercap year_cat
##   <fct>      <fct>    <int>   <dbl>    <int>   <dbl> <chr>
## 1 Tanzania Africa    1952   41.2  8322925   717. Before 1980
## 2 Tanzania Africa    1957   43.0  9452826   699. Before 1980
## 3 Tanzania Africa    1962   44.2 10863958   722. Before 1980
## 4 Tanzania Africa    1967   45.8 12607312   848. Before 1980
## 5 Tanzania Africa    1972   47.6 14706593   916. Before 1980
## 6 Tanzania Africa    1977   49.9 17129565   962. Before 1980
## 7 Tanzania Africa    1982   50.6 19844382   874. 1980-2000
## 8 Tanzania Africa    1987   51.5 23040630   832. 1980-2000
## 9 Tanzania Africa    1992   50.4 26605473   826. 1980-2000
## 10 Tanzania Africa    1997   48.5 30686889   789. 1980-2000
```

# Boolean operators

<b>Operator Meaning</b>	
a & b	and
a   b	or
!a	not

# These do the same thing:

```
filter(.data = gapminder, country == "Tanzania" , year<2000)
```

---

```
filter(.data = gapminder, country == "Tanzania" & year<2000)
```

## Task #2: Filtering

- Use filter() and Boolean logical tests to show...
  - Kenya after 2000
  - Countries where life expectancy in 2002 is over 80
  - Countries where life expectancy in 2007 is below 50 and are not in Africa

## filter the data for Kenya after 200

```
filter(gapminder, country == "Kenya" & year>2000)
```

```
## # A tibble: 2 x 7
##   country continent  year lifeExp      pop gdpPercap year_cat
##   <fct>    <fct>    <int>   <dbl>   <int>   <dbl> <chr>
## 1 Kenya  Africa      2002    51.0 31386842  1288. After 200
## 2 Kenya  Africa      2007    54.1 35610177  1463. After 200
```

filter countries where life expectancy in 2002 is over 80

```
filter(gapminder, lifeExp >80 & year==2002)
```

```
## # A tibble: 7 x 7
```

##	country	continent	year	lifeExp	pop	gdpPercap	year_cat
##	<fct>	<fct>	<int>	<dbl>	<int>	<dbl>	<chr>
## 1	Australia	Oceania	2002	80.4	19546792	30688.	After 200
## 2	Hong Kong, China	Asia	2002	81.5	6762476	30209.	After 200
## 3	Iceland	Europe	2002	80.5	288030	31163.	After 200
## 4	Italy	Europe	2002	80.2	57926999	27968.	After 200
## 5	Japan	Asia	2002	82	127065841	28605.	After 200
## 6	Sweden	Europe	2002	80.0	8954175	29342.	After 200
## 7	Switzerland	Europe	2002	80.6	7361757	34481.	After 200

## Countries where life expectancy in 2007 is below 50 and are not in Africa

```
filter(gapminder, lifeExp <50 & continent!='Africa' & year==2007)
```

```
## # A tibble: 1 x 7
##   country    continent  year lifeExp      pop gdpPercap year_cat
##   <fct>      <fct>    <int>  <dbl>    <int>   <dbl> <chr>
## 1 Afghanistan Asia      2007   43.8 31889923   975. After 200
```



# Common mistakes

## Collapsing multiple tests into one

```
filter(gapminder, 1960 < year < 1980)
```

```
filter(gapminder,  
  year > 1960, year < 1980)
```

## Using multiple tests instead of %in%

```
filter(gapminder,  
  country == "Mexico",  
  country == "Canada",  
  country == "United States")
```

```
filter(gapminder,  
  country %in% c("Mexico", "Canada",  
    "United States"))
```

Create new columns: mutate()

**dplyr: mutate**

```
mutate(.data = DATA, ...)
```

- DATA = Data frame to transform
- ... = Columns to make

# mutate the gdp variable

```
mutate(gapminder, gdp = gdpPercap * pop)
```

```
## # A tibble: 1,704 x 8
##   country      continent year lifeExp      pop gdpPercap year_cat      gdp
##   <fct>        <fct>    <int>  <dbl>    <int>    <dbl> <chr>      <dbl>
## 1 Afghanistan Asia      1952   28.8  8425333    779. Before 1980  6567086330.
## 2 Afghanistan Asia      1957   30.3  9240934    821. Before 1980  7585448670.
## 3 Afghanistan Asia      1962   32.0 10267083    853. Before 1980  8758855797.
## 4 Afghanistan Asia      1967   34.0 11537966    836. Before 1980  9648014150.
## 5 Afghanistan Asia      1972   36.1 13079460    740. Before 1980  9678553274.
## 6 Afghanistan Asia      1977   38.4 14880372    786. Before 1980 11697659231.
## 7 Afghanistan Asia      1982   39.9 12881816    978. 1980-2000  12598563401.
## 8 Afghanistan Asia      1987   40.8 13867957    852. 1980-2000  11820990309.
## 9 Afghanistan Asia      1992   41.7 16317921    649. 1980-2000  10595901589.
## 10 Afghanistan Asia      1997   41.8 22227415    635. 1980-2000  14121995875.
## # ... with 1,694 more rows
```

# mutate 2 variables

```
mutate(gapminder, gdp = gdpPercap * pop,  
        pop_mil = round(pop / 1000000))
```

```
## # A tibble: 1,704 x 9  
##   country      continent year lifeExp      pop gdpPercap year_cat      gdp pop_mil  
##   <fct>        <fct>    <int>  <dbl>    <int>    <dbl> <chr>      <dbl>  <dbl>  
## 1 Afghanistan Asia      1952   28.8  8425333    779. Before 1980  6567086330.      8  
## 2 Afghanistan Asia      1957   30.3  9240934    821. Before 1980  7585448670.      9  
## 3 Afghanistan Asia      1962   32.0 10267083    853. Before 1980  8758855797.     10  
## 4 Afghanistan Asia      1967   34.0 11537966    836. Before 1980  9648014150.     12  
## 5 Afghanistan Asia      1972   36.1 13079460    740. Before 1980  9678553274.     13  
## 6 Afghanistan Asia      1977   38.4 14880372    786. Before 1980 11697659231.     15  
## 7 Afghanistan Asia      1982   39.9 12881816    978. 1980-2000  12598563401.     13  
## 8 Afghanistan Asia      1987   40.8 13867957    852. 1980-2000  11820990309.     14  
## 9 Afghanistan Asia      1992   41.7 16317921    649. 1980-2000  10595901589.     16  
## 10 Afghanistan Asia      1997   41.8 22227415    635. 1980-2000  14121995875.     22  
## # ... with 1,694 more rows
```

Do conditional tests within mutate()

**dplyr: ifelse()**

```
ifelse(TEST,  
       VALUE_IF_TRUE,  
       VALUE_IF_FALSE)
```

- TEST = A logical test
- VALUE\_IF\_TRUE = What happens if test is true
- VALUE\_IF\_FALSE = What happens if test is false

# Create a variable to show before and after 2000

```
mutate(gapminder,  
  after_1960 = ifelse(year > 1960, TRUE, FALSE))
```

```
mutate(gapminder,  
  after_2000 = ifelse(year > 2000,  
    "After 2000",  
    "Before 2000"))
```

```
## # A tibble: 1,704 x 8  
##   country      continent year lifeExp      pop gdpPercap year_cat  after_2000  
##   <fct>        <fct>    <int>   <dbl>    <int>    <dbl> <chr>      <chr>  
## 1 Afghanistan Asia      1952    28.8  8425333    779. Before 1980 Before 2000  
## 2 Afghanistan Asia      1957    30.3  9240934    821. Before 1980 Before 2000  
## 3 Afghanistan Asia      1962    32.0 10267083    853. Before 1980 Before 2000  
## 4 Afghanistan Asia      1967    34.0 11537966    836. Before 1980 Before 2000  
## 5 Afghanistan Asia      1972    36.1 13079460    740. Before 1980 Before 2000  
## 6 Afghanistan Asia      1977    38.4 14880372    786. Before 1980 Before 2000  
## 7 Afghanistan Asia      1982    39.9 12881816    978. 1980-2000 Before 2000  
## 8 Afghanistan Asia      1987    40.8 13867957    852. 1980-2000 Before 2000  
## 9 Afghanistan Asia      1992    41.7 16317921    649. 1980-2000 Before 2000  
## 10 Afghanistan Asia      1997    41.8 22227415    635. 1980-2000 Before 2000  
## # ... with 1,694 more rows
```



# Task #1: Mutate

- Use `mutate()` and `if_else()` to...
  - Add an `africa` column that is TRUE if the country is on the African continent
  - Add a column for logged GDP per capita (hint: use `log()`)
  - Add a column `life_exp_asia` for life expectancy that is TRUE if the country is in Asia and life expectancy is greater than 80
  - Add an `africa_asia` column that says "Africa or Asia" if the country is in Africa or Asia, and "Not Africa or Asia" if it's not

## Add an africa column

```
mutate(gapminder, africa = ifelse(continent == "Africa",  
                                  TRUE, FALSE))
```

```
## # A tibble: 1,704 x 8  
##   country      continent  year lifeExp      pop gdpPercap year_cat  africa  
##   <fct>        <fct>    <int>  <dbl>    <int>    <dbl> <chr>    <lgl>  
## 1 Afghanistan Asia      1952   28.8  8425333    779. Before 1980 FALSE  
## 2 Afghanistan Asia      1957   30.3  9240934    821. Before 1980 FALSE  
## 3 Afghanistan Asia      1962   32.0 10267083    853. Before 1980 FALSE  
## 4 Afghanistan Asia      1967   34.0 11537966    836. Before 1980 FALSE  
## 5 Afghanistan Asia      1972   36.1 13079460    740. Before 1980 FALSE  
## 6 Afghanistan Asia      1977   38.4 14880372    786. Before 1980 FALSE  
## 7 Afghanistan Asia      1982   39.9 12881816    978. 1980-2000  FALSE  
## 8 Afghanistan Asia      1987   40.8 13867957    852. 1980-2000  FALSE  
## 9 Afghanistan Asia      1992   41.7 16317921    649. 1980-2000  FALSE  
## 10 Afghanistan Asia      1997   41.8 22227415    635. 1980-2000  FALSE  
## # ... with 1,694 more rows
```

## Add a column for logged GDP per capita

```
mutate(gapminder, log_gdpPercap = log(gdpPercap))
```

```
## # A tibble: 1,704 x 8
##   country      continent  year lifeExp      pop gdpPercap year_cat  log_gdpPercap
##   <fct>        <fct>    <int>  <dbl>    <int>    <dbl> <chr>        <dbl>
## 1 Afghanistan Asia      1952   28.8  8425333    779. Before 1980    6.66
## 2 Afghanistan Asia      1957   30.3  9240934    821. Before 1980    6.71
## 3 Afghanistan Asia      1962   32.0 10267083    853. Before 1980    6.75
## 4 Afghanistan Asia      1967   34.0 11537966    836. Before 1980    6.73
## 5 Afghanistan Asia      1972   36.1 13079460    740. Before 1980    6.61
## 6 Afghanistan Asia      1977   38.4 14880372    786. Before 1980    6.67
## 7 Afghanistan Asia      1982   39.9 12881816    978. 1980-2000    6.89
## 8 Afghanistan Asia      1987   40.8 13867957    852. 1980-2000    6.75
## 9 Afghanistan Asia      1992   41.7 16317921    649. 1980-2000    6.48
## 10 Afghanistan Asia      1997   41.8 22227415    635. 1980-2000    6.45
## # ... with 1,694 more rows
```

Add a column `life_exp_asia` for Asian countries with `lifeExp>80`

```
mutate(gapminder, life_exp_asia=ifelse(continent=="Asia" & lifeExp>80,  
                                         TRUE, FALSE) )
```

```
## # A tibble: 1,704 x 8  
##   country      continent year lifeExp      pop gdpPercap year_cat life_exp_asia  
##   <fct>        <fct>    <int>   <dbl>    <int>    <dbl> <chr>      <lgl>  
## 1 Afghanistan Asia      1952    28.8  8425333    779. Before 1980 FALSE  
## 2 Afghanistan Asia      1957    30.3  9240934    821. Before 1980 FALSE  
## 3 Afghanistan Asia      1962    32.0 10267083    853. Before 1980 FALSE  
## 4 Afghanistan Asia      1967    34.0 11537966    836. Before 1980 FALSE  
## 5 Afghanistan Asia      1972    36.1 13079460    740. Before 1980 FALSE  
## 6 Afghanistan Asia      1977    38.4 14880372    786. Before 1980 FALSE  
## 7 Afghanistan Asia      1982    39.9 12881816    978. 1980-2000 FALSE  
## 8 Afghanistan Asia      1987    40.8 13867957    852. 1980-2000 FALSE  
## 9 Afghanistan Asia      1992    41.7 16317921    649. 1980-2000 FALSE  
## 10 Afghanistan Asia      1997    41.8 22227415    635. 1980-2000 FALSE  
## # ... with 1,694 more rows
```

## Add an africa\_asia column

```
mutate(gapminder,  
  africa_asia = ifelse(continent %in% c("Africa", "Asia"),  
    "Africa or Asia",  
    "Not Africa or Asia"))
```

```
## # A tibble: 1,704 x 8  
##   country      continent  year lifeExp      pop gdpPercap year_cat africa_asia  
##   <fct>        <fct>    <int>  <dbl>    <int>    <dbl> <chr>      <chr>  
## 1 Afghanistan Asia      1952   28.8  8425333    779. Before 1980 Africa or Asia  
## 2 Afghanistan Asia      1957   30.3  9240934    821. Before 1980 Africa or Asia  
## 3 Afghanistan Asia      1962   32.0 10267083    853. Before 1980 Africa or Asia  
## 4 Afghanistan Asia      1967   34.0 11537966    836. Before 1980 Africa or Asia  
## 5 Afghanistan Asia      1972   36.1 13079460    740. Before 1980 Africa or Asia  
## 6 Afghanistan Asia      1977   38.4 14880372    786. Before 1980 Africa or Asia  
## 7 Afghanistan Asia      1982   39.9 12881816    978. 1980-2000 Africa or Asia  
## 8 Afghanistan Asia      1987   40.8 13867957    852. 1980-2000 Africa or Asia  
## 9 Afghanistan Asia      1992   41.7 16317921    649. 1980-2000 Africa or Asia  
## 10 Afghanistan Asia      1997   41.8 22227415    635. 1980-2000 Africa or Asia  
## # ... with 1,694 more rows
```

# What if you have multiple conditions?

- Make a dataset for just 2002 and calculate logged GDP per capita

## Solution 1

```
gapminder_2002 <- filter(gapminder, year == 2002)
gapminder_2002_log <- mutate(gapminder_2002,
                             log_gdpPercap = log(gdpPercap))
```

## Solution 2: Pipes

- The %>% operator (pipe) takes an object on the left
- Then passes it as the first argument of the function on the right

```
gapminder %>% filter(country == "Kenya")
```

- These two commands do the same thing

```
filter(gapminder, country == "Kenya")  
gapminder %>% filter(country == "Kenya")
```

- Make a dataset for just 2002 and calculate logged GDP per capita

```
gapminder_2002_log <- gapminder %>%  
  filter(year == 2002) %>%  
  mutate(log_gdpPercap = log(gdpPercap))  
gapminder_2002_log
```



%>%

```
leave_house(  
  take_breakfast(  
    get_dressed(  
      wake_up(  
        me, ## start here  
        time = "8:00"),  
        trouser = TRUE, shirt = TRUE , socks=FALSE),  
        mayai = TRUE, viazi = TRUE , chai=TRUE),  
        nduthi = TRUE, car = FALSE)
```

```
me %>%  
  wake_up(time = "8:00") %>%  
  get_dressed(trouser = TRUE, shirt = TRUE , socks=FALSE) %>%  
  take_breakfast( mayai = TRUE, viazi = TRUE , chai=TRUE, ukwanju=FALSE) %>%  
  leave_house( nduthi = TRUE, car = FALSE)
```



Questions()

# Data wrangling with R 1 - Done()

Compute a table of summaries: summarize()

**dplyr: summarize()**

# Summarize

`pivot_longer()` and `pivot_wider()`

**tidyr: reshape data**

# pivot\_wider()

- `pivot_wider()` “widens” data, increasing the number of columns and decreasing the number of rows.
- `pivot_wider()` is an updated approach to `spread()`

```
DATA %>%  
  pivot_wider(names_from,  
              values_from ,  
              ....)
```

- DATA = A data frame to pivot
- names\_from = Column(s) to pivot into wider format.
- values\_from = Column(s) to get the cell values from to be into wider format.
- ... = other specifications (check help)

- Filter data after 1992
- Select the continent, country, year and gdpPercap and pivot\_wider the values of gdpPercap by country

```
gapminder_sub <- gapminder %>%
  select(continent, country, year, gdpPercap) %>%
  filter(year>1992 )
gapminder_sub %>%
  pivot_wider(names_from =country , values_from =gdpPercap )
```

```
## # A tibble: 15 x 144
```

	continent	year	Afghanistan	Albania	Algeria	Angola	Argentina	Australia	Austria	Bahrain	Bangladesh	Belgium	Beni
	<fct>	<int>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>
## 1	Asia	1997	635.	NA	NA	NA	NA	NA	NA	20292.	973.	NA	NA
## 2	Asia	2002	727.	NA	NA	NA	NA	NA	NA	23404.	1136.	NA	NA
## 3	Asia	2007	975.	NA	NA	NA	NA	NA	NA	29796.	1391.	NA	NA
## 4	Europe	1997	NA	3193.	NA	NA	NA	NA	29096.	NA	NA	27561.	NA
## 5	Europe	2002	NA	4604.	NA	NA	NA	NA	32418.	NA	NA	30486.	NA
## 6	Europe	2007	NA	5937.	NA	NA	NA	NA	36126.	NA	NA	33693.	NA
## 7	Africa	1997	NA	NA	4797.	2277.	NA	NA	NA	NA	NA	NA	1233
## 8	Africa	2002	NA	NA	5288.	2773.	NA	NA	NA	NA	NA	NA	1373
## 9	Africa	2007	NA	NA	6223.	4797.	NA	NA	NA	NA	NA	NA	1441
## 10	Americas	1997	NA	NA	NA	NA	10967.	NA	NA	NA	NA	NA	NA
## 11	Americas	2002	NA	NA	NA	NA	8798.	NA	NA	NA	NA	NA	NA
## 12	Americas	2007	NA	NA	NA	NA	12779.	NA	NA	NA	NA	NA	NA
## 13	Oceania	1997	NA	NA	NA	NA	NA	26998.	NA	NA	NA	NA	NA
## 14	Oceania	2002	NA	NA	NA	NA	NA	30688.	NA	NA	NA	NA	NA
## 15	Oceania	2007	NA	NA	NA	NA	NA	34435.	NA	NA	NA	NA	NA

## # ... with 131 more variables: Bolivia <dbl>, Bosnia and Herzegovina <dbl>, Botswana <dbl>, Brazil <dbl>, Bulgaria <dbl>, Burkina Faso <dbl>, Burundi <dbl>, Cambodia <dbl>, Cameroon <dbl>, Canada <dbl>, Central African Republic <dbl>, Chad <dbl>, Chile <dbl>, China <dbl>, Colombia <dbl>, Comoros <dbl>, Congo, Dem. Rep. <dbl>, Congo, Rep. <dbl>, Costa Rica <dbl>, Cote d'Ivoire <dbl>, Croatia <dbl>, Cuba <dbl>, Czech Republic <dbl>, Denmark <dbl>, Djibouti <dbl>, Dominican Republic <dbl>, Ecuador <dbl>, Egypt <dbl>,



- What if I remove the continent?

```
country_gdp_wider <- gapminder_sub %>%
  select(-continent) %>%
  pivot_wider(names_from =country , values_from =gdpPercap )
country_gdp_wider
```

```
## # A tibble: 3 x 143
##   year Afghanistan Albania Algeria Angola Argentina Australia Austria Bahrain Bangladesh Belgium Benin Bolivia
##   <int>          <dbl>   <dbl>   <dbl>   <dbl>       <dbl>       <dbl>   <dbl>   <dbl>   <dbl> <dbl> <dbl>
## 1  1997           635.    3193.   4797.   2277.    10967.    26998.  29096.  20292.    973.  27561. 1233.  3326.
## 2  2002           727.   4604.   5288.   2773.     8798.    30688.  32418.  23404.   1136.  30486. 1373.  3413.
## 3  2007           975.   5937.   6223.   4797.   12779.    34435.  36126.  29796.   1391.  33693. 1441.  3822.
## # ... with 130 more variables: Bosnia and Herzegovina <dbl>, Botswana <dbl>, Brazil <dbl>, Bulgaria <dbl>,
## #   Burkina Faso <dbl>, Burundi <dbl>, Cambodia <dbl>, Cameroon <dbl>, Canada <dbl>,
## #   Central African Republic <dbl>, Chad <dbl>, Chile <dbl>, China <dbl>, Colombia <dbl>, Comoros <dbl>,
## #   Congo, Dem. Rep. <dbl>, Congo, Rep. <dbl>, Costa Rica <dbl>, Cote d'Ivoire <dbl>, Croatia <dbl>, Cuba <dbl>,
## #   Czech Republic <dbl>, Denmark <dbl>, Djibouti <dbl>, Dominican Republic <dbl>, Ecuador <dbl>, Egypt <dbl>,
## #   El Salvador <dbl>, Equatorial Guinea <dbl>, Eritrea <dbl>, Ethiopia <dbl>, Finland <dbl>, France <dbl>,
## #   Gabon <dbl>, Gambia <dbl>, Germany <dbl>, Ghana <dbl>, Greece <dbl>, Guatemala <dbl>, Guinea <dbl>,
## #   Guinea-Bissau <dbl>, Haiti <dbl>, Honduras <dbl>, Hong Kong, China <dbl>, Hungary <dbl>, Iceland <dbl>,
## #   India <dbl>, Indonesia <dbl>, Iran <dbl>, Iraq <dbl>, Ireland <dbl>, Israel <dbl>, Italy <dbl>,
## #   Jamaica <dbl>, Japan <dbl>, Jordan <dbl>, Kenya <dbl>, Korea, Dem. Rep. <dbl>, Korea, Rep. <dbl>,
## #   Kuwait <dbl>, Lebanon <dbl>, Lesotho <dbl>, Liberia <dbl>, Libya <dbl>, Madagascar <dbl>, Malawi <dbl>,
## #   Malaysia <dbl>, Mali <dbl>, Mauritania <dbl>, Mauritius <dbl>, Mexico <dbl>, Mongolia <dbl>,
## #   Montenegro <dbl>, Morocco <dbl>, Mozambique <dbl>, Myanmar <dbl>, Namibia <dbl>, Nepal <dbl>,
## #   Netherlands <dbl>, New Zealand <dbl>, Nicaragua <dbl>, Niger <dbl>, Nigeria <dbl>, Norway <dbl>, Oman <dbl>,
## #   Pakistan <dbl>, Panama <dbl>, Paraguay <dbl>, Peru <dbl>, Philippines <dbl>, Poland <dbl>, Portugal <dbl>,
## #   Puerto Rico <dbl>, Reunion <dbl>, Romania <dbl>, Rwanda <dbl>, Sao Tome and Principe <dbl>,
## #   Saudi Arabia <dbl>, Senegal <dbl>, Serbia <dbl>, ...
```

# Task #1: pivot\_wider

- Use `pivot_wider()` to...
  - Show the population data only for African countries before 1992. (hint: `pivot_wider()` population values from countries)
  - Create a `year_cat` variable that is "Before 1980" if year in 1952, 1957, 1962, 1967, 1972, 1977, "1980-2000" if year in 1982, 1987, 1992, 1997 and "After 2000" if year in 2002 and 2007
  - Summarize the median `gdpPerCap` for each `year_cat` by country
  - Pivot\_wider the median values from the countries

## Show the population data only for African countries before 1992

```
africa_before_1992 <- gapminder %>%  
  filter(continent=="Africa") %>%  
  select(country, year, pop) %>%  
  filter(year<1992 )  
africa_before_1992_wide <- africa_before_1992 %>%  
  pivot_wider(names_from =country , values_from =pop )  
africa_before_1992_wide
```

```
## # A tibble: 8 x 53  
##   year  Algeria  Angola  Benin Botswana `Burkina Faso` Burundi Cameroon `Central African Republi~  Chad Comoro  
##   <int>    <int>    <int>    <int>    <int>         <int>    <int>    <int>         <int> <int>    <int>  
## 1  1952   9279525  4232095  1738315   442308         4469979  2445618   5009067         1291695  2.68e6  15393  
## 2  1957  10270856  4561361  1925173   474639         4713416  2667518   5359923         1392284  2.89e6  17092  
## 3  1962  11000948  4826015  2151895   512764         4919632  2961915   5793633         1523478  3.15e6  19168  
## 4  1967  12760499  5247469  2427334   553541         5127935  3330989   6335506         1733638  3.50e6  21737  
## 5  1972  14760787  5894858  2761407   619351         5433886  3529983   7021028         1927260  3.90e6  25002  
## 6  1977  17152804  6162675  3168267   781472         5889574  3834415   7959865         2167533  4.39e6  30473  
## 7  1982  20033753  7016384  3641603   970347         6634596  4580410   9250831         2476971  4.88e6  34864  
## 8  1987  23254956  7874230  4243788  1151184         7586551  5126023  10780667         2840009  5.50e6  39511  
## # ... with 42 more variables: Congo, Dem. Rep. <int>, Congo, Rep. <int>, Cote d'Ivoire <int>, Djibouti <int>,  
## # Egypt <int>, Equatorial Guinea <int>, Eritrea <int>, Ethiopia <int>, Gabon <int>, Gambia <int>, Ghana <int>,  
## # Guinea <int>, Guinea-Bissau <int>, Kenya <int>, Lesotho <int>, Liberia <int>, Libya <int>, Madagascar <int>,  
## # Malawi <int>, Mali <int>, Mauritania <int>, Mauritius <int>, Morocco <int>, Mozambique <int>, Namibia <int>,  
## # Niger <int>, Nigeria <int>, Reunion <int>, Rwanda <int>, Sao Tome and Principe <int>, Senegal <int>,  
## # Sierra Leone <int>, Somalia <int>, South Africa <int>, Sudan <int>, Swaziland <int>, Tanzania <int>,  
## # Togo <int>, Tunisia <int>, Uganda <int>, Zambia <int>, Zimbabwe <int>
```

## Add a column `year_cat`

```
gapminder <- gapminder %>%
  mutate(year_cat = ifelse(year %in% c(1952, 1957, 1962, 1967, 1972, 1977),
                             "Before 1980",
                             ifelse(year %in% c(1982, 1987, 1992, 1997),
                                      "1980-2000", "After 2000")))
glimpse(gapminder)
```

[illegible]

```
table(gapminder$year_cat)
```

##			
##	1980-2000	After 200	Before 1980
##	568	284	852

## Summarize the median of each country by `year_cat`

```
gapminder_yearcat <- gapminder %>%  
  group_by(year_cat, country) %>%  
  summarise(med_gdpPercap=median(gdpPercap))
```

```
## `summarise()` has grouped output by 'year_cat'. You can override using the `.groups` argument.
```

```
head(gapminder_yearcat)
```

```
## # A tibble: 6 x 3  
## # Groups:   year_cat [1]  
##   year_cat country      med_gdpPercap  
##   <chr>      <fct>          <dbl>  
## 1 1980-2000 Afghanistan      751.  
## 2 1980-2000 Albania        3412.  
## 3 1980-2000 Algeria        5352.  
## 4 1980-2000 Angola         2529.  
## 5 1980-2000 Argentina       9224.  
## 6 1980-2000 Australia      22657.
```

## pivot\_wider year\_cat values over country

```
median_gdpPercap <- gapminder_yearcat %>%  
  pivot_wider(names_from =country , values_from =med_gdpPercap )  
median_gdpPercap
```

```
## # A tibble: 3 x 143  
## # Groups:   year_cat [3]  
##   year_cat Afghanistan Albania Algeria Angola Argentina Australia Austria Bahrain Bangladesh Belgium Benin Bolivia  
##   <chr>          <dbl>   <dbl>   <dbl>   <dbl>      <dbl>      <dbl>   <dbl>   <dbl>      <dbl>   <dbl> <dbl>   <dbl>  
## 1 1980-20~      751.    3412.   5352.   2529.     9224.    22657.  25365.  19123.     795.   24051. 1229.  3059  
## 2 After 2~      851.    5271.   5756.   3785.    10789.    32562.  34272.  26600.    1264.   32089. 1407.  3618  
## 3 Before ~      803.    2537.   3130.   4049.     7593.    13372.  11793.  13779.     673.   12070. 1032.  2632  
## # ... with 130 more variables: Bosnia and Herzegovina <dbl>, Botswana <dbl>, Brazil <dbl>, Bulgaria <dbl>,  
## #   Burkina Faso <dbl>, Burundi <dbl>, Cambodia <dbl>, Cameroon <dbl>, Canada <dbl>,  
## #   Central African Republic <dbl>, Chad <dbl>, Chile <dbl>, China <dbl>, Colombia <dbl>, Comoros <dbl>,  
## #   Congo, Dem. Rep. <dbl>, Congo, Rep. <dbl>, Costa Rica <dbl>, Cote d'Ivoire <dbl>, Croatia <dbl>, Cuba <dbl>,  
## #   Czech Republic <dbl>, Denmark <dbl>, Djibouti <dbl>, Dominican Republic <dbl>, Ecuador <dbl>, Egypt <dbl>,  
## #   El Salvador <dbl>, Equatorial Guinea <dbl>, Eritrea <dbl>, Ethiopia <dbl>, Finland <dbl>, France <dbl>,  
## #   Gabon <dbl>, Gambia <dbl>, Germany <dbl>, Ghana <dbl>, Greece <dbl>, Guatemala <dbl>, Guinea <dbl>,  
## #   Guinea-Bissau <dbl>, Haiti <dbl>, Honduras <dbl>, Hong Kong, China <dbl>, Hungary <dbl>, Iceland <dbl>,  
## #   India <dbl>, Indonesia <dbl>, Iran <dbl>, Iraq <dbl>, Ireland <dbl>, Israel <dbl>, Italy <dbl>,  
## #   Jamaica <dbl>, Japan <dbl>, Jordan <dbl>, Kenya <dbl>, Korea, Dem. Rep. <dbl>, Korea, Rep. <dbl>,  
## #   Kuwait <dbl>, Lebanon <dbl>, Lesotho <dbl>, Liberia <dbl>, Libya <dbl>, Madagascar <dbl>, Malawi <dbl>,  
## #   Malaysia <dbl>, Mali <dbl>, Mauritania <dbl>, Mauritius <dbl>, Mexico <dbl>, Mongolia <dbl>,  
## #   Montenegro <dbl>, Morocco <dbl>, Mozambique <dbl>, Myanmar <dbl>, Namibia <dbl>, Nepal <dbl>,  
## #   Netherlands <dbl>, New Zealand <dbl>, Nicaragua <dbl>, Niger <dbl>, Nigeria <dbl>, Norway <dbl>, Oman <dbl>,  
## #   Pakistan <dbl>, Panama <dbl>, Paraguay <dbl>, Peru <dbl>, Philippines <dbl>, Poland <dbl>, Portugal <dbl>,  
## #   Puerto Rico <dbl>, Reunion <dbl>, Romania <dbl>, Rwanda <dbl>, Sao Tome and Principe <dbl>,  
## #   Saudi Arabia <dbl>, Senegal <dbl>, Serbia <dbl>, ...
```

# pivot\_longer()

- `pivot_longer()` “lengthens” data, increasing the number of rows and decreasing the number of columns.
- `pivot_longer()` is an updated approach to `gather()`

```
DATA %>%  
  pivot_longer(cols,  
               names_to = ,  
               values_to = ,  
               ....)
```

- DATA = A data frame to pivot
- cols = Columns to pivot into longer format.
- names\_to = name of the column to create from the data stored in the column names
- values\_to = string specifying the name of the column to create from the data stored in cell values
- ... = other specifications (check help)

- We use the `country_gdp` pivot\_longer the values of `gdpPercap`

```
country_gdp_longer <- country_gdp_wider %>%  
  pivot_longer(cols =!year, names_to="country", values_to = "gdpPercap")  
country_gdp_longer
```

```
## # A tibble: 426 x 3  
##   year country      gdpPercap  
##   <int> <chr>      <dbl>  
## 1  1997 Afghanistan    635.  
## 2  1997 Albania       3193.  
## 3  1997 Algeria       4797.  
## 4  1997 Angola        2277.  
## 5  1997 Argentina    10967.  
## 6  1997 Australia    26998.  
## 7  1997 Austria      29096.  
## 8  1997 Bahrain      20292.  
## 9  1997 Bangladesh     973.  
## 10 1997 Belgium     27561.  
## # ... with 416 more rows
```



# Task #1: pivot\_longer

- Use `pivot_longer()` to...
  - `pivot_longer` the values of `median_gdpPerCap` into `country` and `med_gdpPerCap`

## Solution

```
median_gdpPercap_longer <- median_gdpPercap %>%  
  pivot_longer(cols =!year_cat , names_to="country", values_to = "med_gdpPercap")  
median_gdpPercap_longer
```

```
## # A tibble: 426 x 3  
## # Groups:   year_cat [3]  
##   year_cat country      med_gdpPercap  
##   <chr>      <chr>          <dbl>  
## 1 1980-2000 Afghanistan      751.  
## 2 1980-2000 Albania        3412.  
## 3 1980-2000 Algeria        5352.  
## 4 1980-2000 Angola         2529.  
## 5 1980-2000 Argentina      9224.  
## 6 1980-2000 Australia     22657.  
## 7 1980-2000 Austria       25365.  
## 8 1980-2000 Bahrain       19123.  
## 9 1980-2000 Bangladesh      795.  
## 10 1980-2000 Belgium      24051.  
## # ... with 416 more rows
```

## How to export data to CSV

```
write_csv(x = median_gdpPercap_longer, file = "median_gdpPercap_longer.csv")
```

## How to export data to Stata

```
## bad code  
library(haven)  
haven::write_dta(x = median_gdpPercap_longer, file = "median_gdpPercap_longer.dta")
```

Questions()

# Data wrangling with R - Done()