



# Welcome to the Tidyverse

## An Introduction To Data Science

Teon Brooks



# Our Day

2:00-3:00 PM

Data Transformations

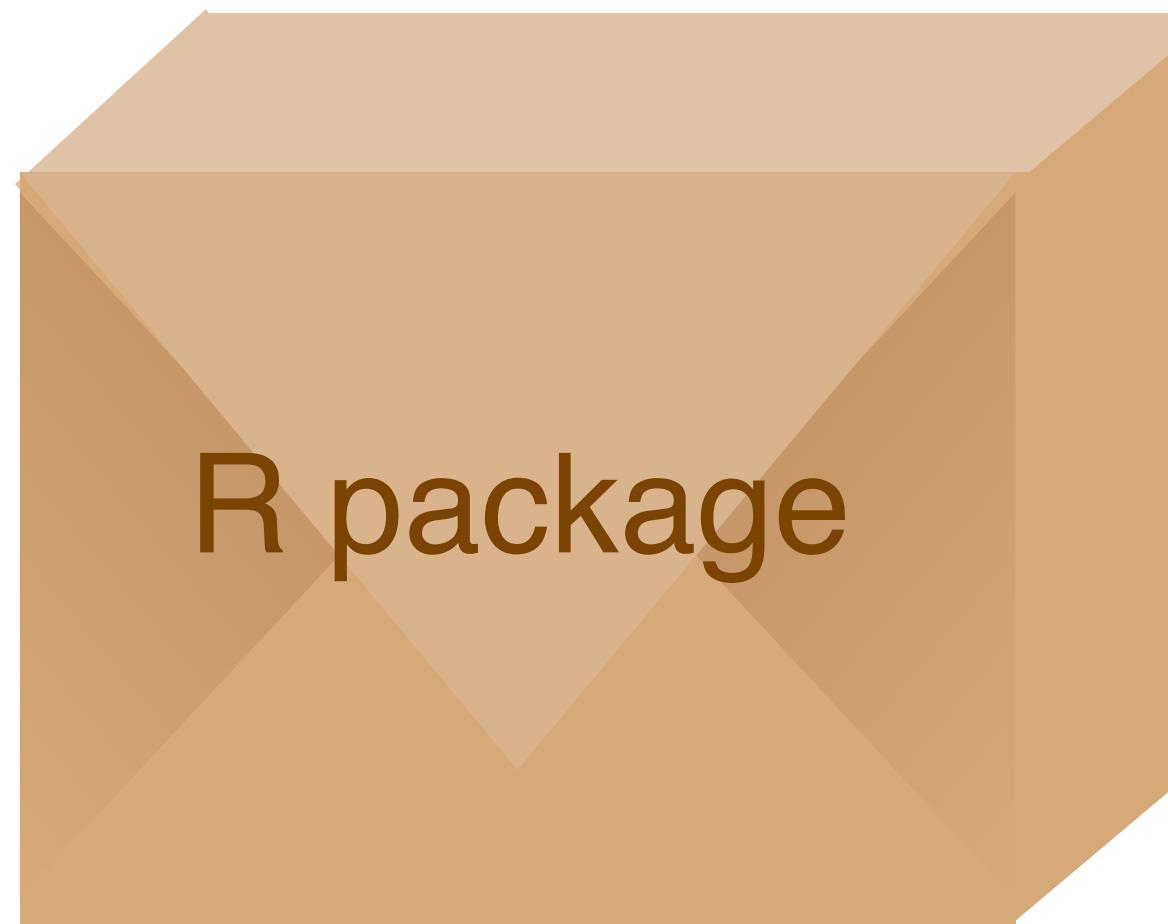
3:00-4:00 PM

Data Visualization 2

# Transform Data with



# gapminder



R package

A subset of Gapminder data: population, GDP per capita and life expectancy, for countries over time.

```
# install.packages("gapminder")  
library(gapminder)
```



# Your Turn 0

04-Transform-data.Rmd

Run the setup chunk

```
# install.packages("gapminder")
library(gapminder)
```



# gapminder

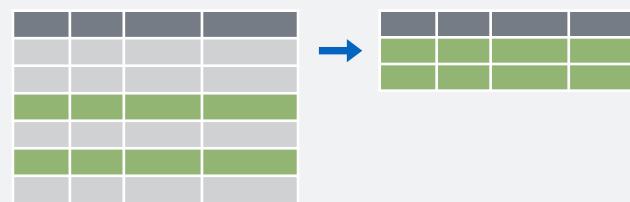
country	continent	year	lifeExp	pop	gdpPerCap
<fctr>	<fctr>	<int>	<dbl>	<int>	<dbl>
Afghanistan	Asia	1952	28.80100	8425333	779.4453
Afghanistan	Asia	1957	30.33200	9240934	820.8530
Afghanistan	Asia	1962	31.99700	10267083	853.1007
Afghanistan	Asia	1967	34.02000	11537966	836.1971
Afghanistan	Asia	1972	36.08800	13079460	739.9811
Afghanistan	Asia	1977	38.43800	14880372	786.1134
Afghanistan	Asia	1982	39.85400	12881816	978.0114
Afghanistan	Asia	1987	40.82200	13867957	852.3959
Afghanistan	Asia	1992	41.67400	16317921	649.3414
Afghanistan	Asia	1997	41.76300	22227415	635.3414

1-10 of 1,704 rows

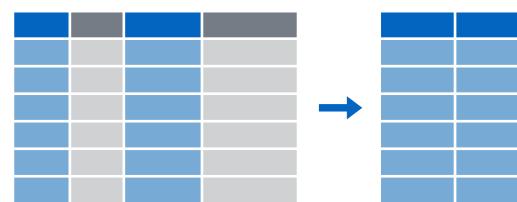
Previous [1](#) [2](#) [3](#) [4](#) [5](#) [6](#) ... [100](#) Next



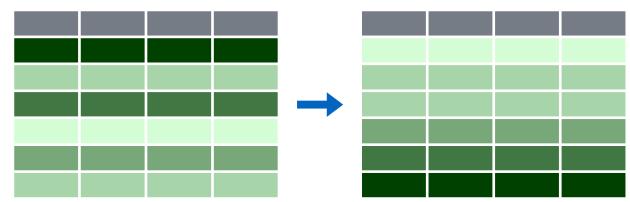
# dplyr: Data manipulation verbs



Extract cases with `filter()`



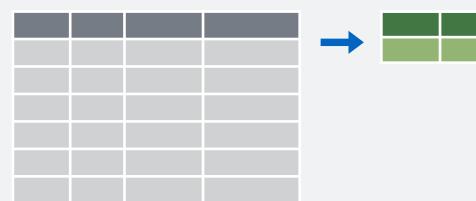
Extract variables with `select()`



Arrange cases, with `arrange()`.



Make new variables, with `mutate()`.



Make tables of summaries with `summarize()`.

along with `group_by()`



# filter()

# filter()

Extract rows that meet logical criteria.

```
filter(.data, ...)
```

data frame to transform

one or more logical tests  
(filter returns each row for which the test is TRUE)



# filter()

Extract rows that meet logical criteria.

```
filter(gapminder, country == "New Zealand")
```

gapminder

The diagram illustrates the use of the `filter()` function. On the left, a large grey arrow points from a full `gapminder` dataset to a smaller subset on the right. The `gapminder` dataset contains columns: `country`, `continent`, `year`, and `...`. The subset on the right retains the same structure but only includes rows where `country` is "New Zealand". The rows in the subset are highlighted in green, while the rest of the original data is shown in grey.

country	continent	year	...
Afghanistan	Asia	1952	
Afghanistan	Asia	1957	
...	...	...	
Netherlands	Europe	2007	
New	Oceania	1952	
New	Oceania	1957	

country continent year ...

New Oceania 1952

New Oceania 1957

New Oceania 1962

New Oceania 1967

... ... ... ...



# filter()

Extract rows that meet logical criteria.

```
filter(gapminder, country == "New Zealand")
```

gapminder			
country	continent	year	...
Afghanistan	Asia	1952	
Afghanistan	Asia	1957	
...	...	...	
Netherlands	Europe	2007	
New Zealand	Oceania	1952	
New Zealand	Oceania	1957	

= sets  
(returns nothing)  
== tests if equal  
(returns TRUE or

# Logical tests

## ?Comparison

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

# Your Turn 1

04-Transform-data.Rmd

See if you can use the logical operators to manipulate our code below to show:

1. The data for Canada
2. All data for countries in Oceania
3. Rows where the life expectancy is greater than 82



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

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

```
filter(gapminder, lifeExp > 82)
```

# Two common mistakes

## 1. Using `=` instead of `==`

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

## 2. Forgetting quotes

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



# filter()

Extract rows that meet every logical criteria.

```
filter(gapminder, country == "New Zealand", year > 2000)
```

gapminder

The diagram illustrates the use of the `filter()` function. On the left, a large gray arrow points from a full `gapminder` dataset to a smaller subset on the right. The original dataset has columns: `country`, `continent`, `year`, and `...`. The subset on the right includes only the rows where `country` is "New Zealand" and `year` is greater than 2000. The subset has columns: `country`, `continent`, `year`, and `...`.

country	continent	year	...
...	...	...	
New Zealand	Oceania	1952	
...	...	...	
New Zealand	Oceania	2002	
New Zealand	Oceania	2007	

# Boolean operators

?base::Logic

a & b	and
a   b	or
! a	not



# filter()

Extract rows that meet every logical criteria.

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

gapminder

The diagram illustrates the use of the `filter()` function. On the left, a large gray arrow points from a full `gapminder` dataset to a smaller subset on the right. The original dataset has columns: `country`, `continent`, `year`, and `...`. The subset on the right includes only the rows where `country == "New Zealand"` and `year > 2000`. The subset has columns: `country`, `continent`, `year`, and `...`. The rows in the subset are highlighted in green.

country	continent	year	...
...	...	...	
New Zealand	Oceania	1952	
...	...	...	
New Zealand	Oceania	2002	
New Zealand	Oceania	2007	

# Your Turn 2

Use Boolean operators to alter the code below to return only the rows that contain:

1. Canada before 1970
2. Countries where life expectancy in 2007 is below 50
3. Countries where life expectancy in 2007 is below 50, and are not in Africa.



```
filter(gapminder, country == "Canada", year < 1970)
```

```
filter(gapminder, year == 2007, lifeExp < 50)
```

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

# Two more common mistakes

## 3. Collapsing multiple tests into one

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

## 4. Stringing together many tests (when you could use %in%)

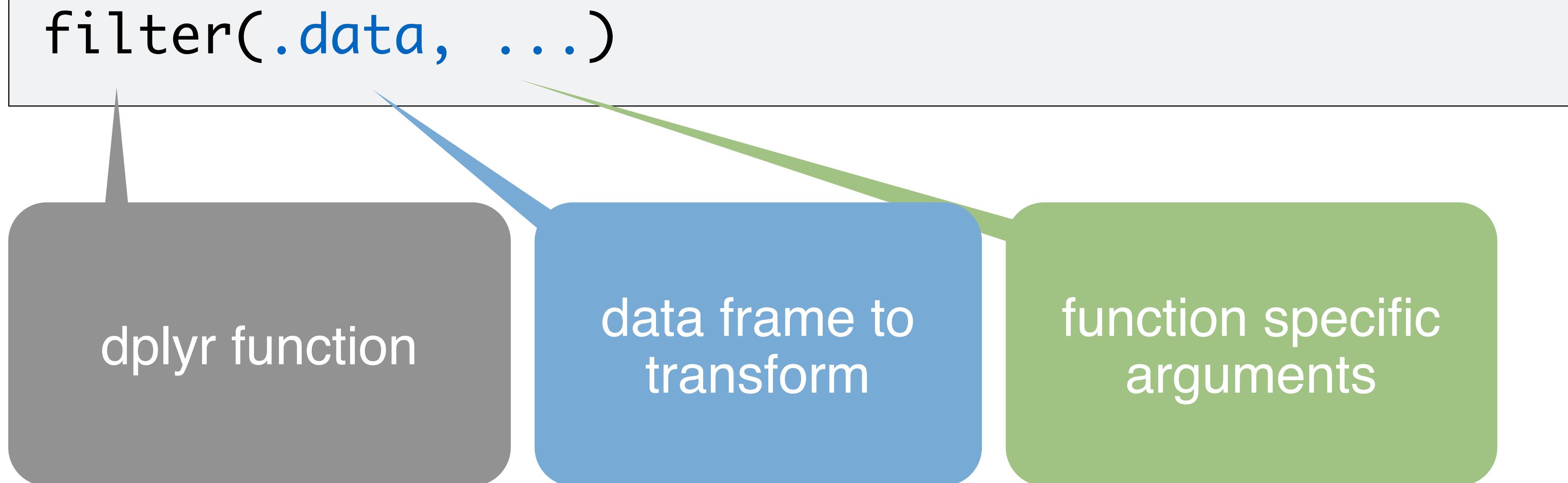
```
filter(gapminder, country == "New Zealand" |  
       country == "Canada" | country == "United States")  
filter(gapminder,  
       country %in% c("New Zealand", "Canada", "United States"))
```



# common syntax

Each function takes a data frame / tibble as its first argument and returns a data frame / tibble.

```
filter(.data, ...)
```



# mutate()

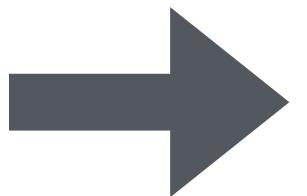
# mutate()

Create new columns.

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

gapminder

country	continent	year	...
Afghanistan	Asia	1952	
Afghanistan	Asia	1957	
Afghanistan	Asia	1962	
Afghanistan	Asia	1967	
Afghanistan	Asia	1972	
Afghanistan	Asia	1977	



country	continent	year	...	gdp
Afghanistan	Asia	1952		6567086330
Afghanistan	Asia	1957		7585448670
Afghanistan	Asia	1962		8758855797
Afghanistan	Asia	1967		9648014150
Afghanistan	Asia	1972		9678553274
Afghanistan	Asia	1977		11697659231



# mutate()

Create new columns.

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

gapminder	country	continent	year	pop	gdp
Afghanistan	Afghanistan	Asia	1955	2.957	6567086330
Afghanistan	Afghanistan	Asia	1957	3.057	7585448670
Afghanistan	Afghanistan	Asia	1962	3.157	8758855797
Afghanistan	Afghanistan	Asia	1967	3.257	9648014150
Afghanistan	Afghanistan	Asia	1972	3.357	9678553274
Afghanistan	Afghanistan	Asia	1977	3.457	11697659231

dplyr function

data frame to transform

function specific arguments



# round()

Round a number to a specified number of decimal digits (0 by default)

```
x <- c(1.2, 1/3, 10.01)
round(x)
[1] 1 0 10
round(x, digits = 2)
[1] 1.20 0.33 10.01
```

When used in  
mutate()  
the argument will be a

```
mutate(gapminder, gdp = gdpPerCap * pop,
       pop_mill = round(pop/1000000))
```



## Vectorized Functions

### TO USE WITH MUTATE ()

`dplyr::mutate()` and `transmute()` apply vectorized functions to columns to create new columns. Vectorized functions take vectors as input and return vectors of the same length as output.

vectorized function ➔

### OFFSETS

`dplyr::lag()` - Offset elements by 1  
`dplyr::lead()` - Offset elements by -1

### CUMULATIVE AGGREGATES

`dplyr::cumall()` - Cumulative all()  
`dplyr::cumany()` - Cumulative any()  
  `cummax()` - Cumulative max()  
`dplyr::cummean()` - Cumulative mean()  
  `cummin()` - Cumulative min()  
  `cumprod()` - Cumulative prod()  
  `cumsum()` - Cumulative sum()

### RANKINGS

`dplyr::cume_dist()` - Proportion of all values <=  
`dplyr::dense_rank()` - rank with ties = min, no gaps  
`dplyr::min_rank()` - rank with ties = min  
`dplyr::ntile()` - bins into n bins  
`dplyr::percent_rank()` - min\_rank scaled to [0,1]  
`dplyr::row_number()` - rank with ties = "first"

### MATH

+, -, \*, /, ^, %/%, %% - arithmetic ops  
`log()`, `log2()`, `log10()` - logs  
<, <=, >, >=, !=, == - logical comparisons

### MISC

`dplyr::between()` - x >= left & x <= right  
`dplyr::case_when()` - multi-case if\_else()  
`dplyr::coalesce()` - first non-NA values by element across a set of vectors  
`dplyr::if_else()` - element-wise if() + else()  
`dplyr::na_if()` - replace specific values with NA  
  `pmax()` - element-wise max()  
  `pmin()` - element-wise min()  
`dplyr::recode()` - Vectorized switch()  
`dplyr::recode_factor()` - Vectorized switch() for factors

# Vectorized functions

Take a vector as input.  
Return a vector of the same length as output.

**Vectorized Functions**  
TO USE WITH MUTATE ()  
`mutate()` and `transmute()` apply vectorized functions to columns to create new columns. Vectorized functions take vectors as input and return vectors of the same length as output.

vectorized function ➔

**OFFSETS**  
`dplyr::lag()` - Offset elements by 1  
`dplyr::lead()` - Offset elements by -1

**CUMULATIVE AGGREGATES**  
`dplyr::cumall()` - Cumulative all()  
`dplyr::cumany()` - Cumulative any()  
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`dplyr::cummean()` - Cumulative mean()  
  `cummin()` - Cumulative min()  
  `cumprod()` - Cumulative prod()  
  `cumsum()` - Cumulative sum()

**RANKINGS**  
`dplyr::cume_dist()` - Proportion of all values <=  
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`dplyr::row_number()` - rank with ties = "first"

**MATH**  
+, -, \*, /, ^, %/%, %% - arithmetic ops  
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`dplyr::between()` - x >= left & x <= right  
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**Summary Functions**  
TO USE WITH SUMMARISE ()  
`summarise()` applies summary functions to columns to create a new table. Summary functions take vectors as input and return single values as output.

summary function ➔

**COUNTS**  
`dplyr::n()` - number of values/rows  
`dplyr::n_distinct()` - # of uniques  
  `sum(is.na())` - # of non-NA's

**LOCATION**  
  `mean()` - mean, also `mean(is.na())`  
  `median()` - median

**LOGICALS**  
  `mean()` - Proportion of TRUE's  
  `sum()` - # of TRUE's

**POSITION/ORDER**  
`dplyr::first()` - first value  
`dplyr::last()` - last value  
`dplyr::nth()` - value in nth location of vector

**RANK**  
  `quantile()` - nth quantile  
  `min()` - minimum value  
  `max()` - maximum value

**SPREAD**  
  `IQR()` - Inter-Quartile Range  
  `mad()` - mean absolute deviation  
  `sd()` - standard deviation  
  `var()` - variance

**Row Names**  
Tidy data does not use rownames, which store a variable outside of the columns. To work with the rownames, first move them into a column.

rownames\_to\_column()  
Move row names into col.  
a <- rownames\_to\_column(iris, var = "C")

column\_to\_rownames()  
Move col in row names.  
column\_to\_rownames(a, var = "C")

Also has `rownames()`, `remove_rownames()`

**Combine Tables**  
COMBINE VARIABLES

Combine Tables

Combine Cases

**Extract Rows**

**dplyr**

Most useful:

• Math

• Misc



# min\_rank()

A goto ranking function (ties share the lowest rank)

```
min_rank(c(50, 100, 1000))  
# [1] 1 2 3
```

```
min_rank(desc(c(50, 100, 1000)))  
# [1] 3 2 1
```



# Your Turn 3

Add an `africa` column, which contains TRUE if the country is on the Africa continent.

Add a `rank_pop` column to rank each row in `gapminder` from largest pop to smallest pop.

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



```
mutate(gapminder,
       africa = continent == "Africa",
       rank_pop = min_rank(desc(pop)))

## # A tibble: 1,704 x 8
##   country continent year lifeExp      pop gdpPercap africa rank_pop
##   <fctr>    <fctr> <int>   <dbl>    <int>     <dbl>    <lgl>    <int>
## 1 Afghanistan    Asia  1952  28.801  8425333  779.4453 FALSE     762
## 2 Afghanistan    Asia  1957  30.332  9240934  820.8530 FALSE     706
## 3 Afghanistan    Asia  1962  31.997 10267083  853.1007 FALSE     638
## 4 Afghanistan    Asia  1967  34.020 11537966  836.1971 FALSE     576
## 5 Afghanistan    Asia  1972  36.088 13079460  739.9811 FALSE     536
```

%>%

# Multistep Operations

Consider the following:

Add a new column for rank, then

Extract rows with rank lower than 10

Use intermediate variables:

```
gapminder_ranked <- mutate(gapminder,  
                           rank = min_rank(desc(pop)))  
filter(gapminder_ranked, rank < 10)
```

# Multistep Operations

Consider the following:

Add a new column for rank, then

Extract rows with rank lower than 10

Do it all in one line:

```
filter(mutate(gapminder,  
             rank = min_rank(desc(pop))),  
       rank < 10)
```

# Multistep Operations

Consider the following:

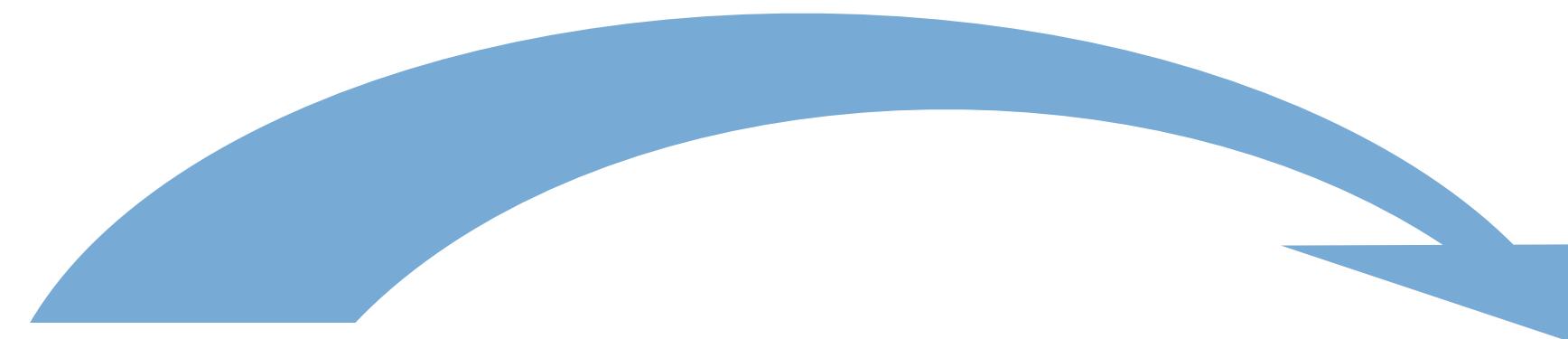
Add a new column for rank, then

Extract rows with rank lower than 10

Do it all in one line:

```
filter(mutate(gapminder, rank = min_rank(desc(pop))),  
rank < 10)
```

# The pipe operator %>%



```
gapminder %>% filter(_____, country == "Canada")
```

Passes result on left into first argument of function on right.  
So, for example, these do the same thing. Try it.

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



# Pipes

```
gapminder_ranked <- mutate(gapminder,  
                           rank = min_rank(desc(pop)))  
filter(gapminder_ranked, rank < 10)
```

```
gapminder %>%  
  mutate(rank = min_rank(desc(pop))) %>%  
  filter(rank < 10)
```

Consider the following:  
Add a new column for rank, then  
Extract rows with rank lower than 10

# Shortcut to type %>%

**Cmd** + **Shift** + **M** (Mac)

**Ctrl** + **Shift** + **M** (Windows)



# summarize()

# summarize()

Compute table of summaries.

```
gapminder %>% summarize(mean_life = mean(lifeExp))  
summarize(gapminder, mean_life = mean(lifeExp))
```

gapminder

country	continent	year	lifeExp	...	mean_life
Afghanistan	Asia	1952	28.801		59.47444
Afghanistan	Asia	1957	30.332		
Afghanistan	Asia	1962	31.997		
Afghanistan	Asia	1967	34.020		
Afghanistan	Asia	1972	36.088		



# summarize()

Compute table of summaries.

```
gapminder %>% summarize(mean_life = mean(lifeExp),  
                           min_life = min(lifeExp))
```

gapminder

country	continent	year	lifeExp	...
Afghanistan	Asia	1952	28.801	
Afghanistan	Asia	1957	30.332	
Afghanistan	Asia	1962	31.997	
Afghanistan	Asia	1967	34.020	
Afghanistan	Asia	1972	36.088	

→

mean_life	min_life
59.47444	23.599

# Summary functions

Take a vector as input.  
Return a single value as output.

## Summary Functions

**TO USE WITH SUMMARISE ()**

**summarise()** applies summary functions to columns to create a new table. Summary functions take vectors as input and return single values as output.

summary function →

**COUNTS**

dplyr::n() - number of values/rows  
dplyr::n\_distinct() - # of uniques  
sum(!is.na()) - # of non-NA's

**LOCATION**

mean() - mean, also mean(is.na())  
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mean() - Proportion of TRUE's  
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dplyr::nth() - value in nth location of vector

**RANK**

quantile() - nth quantile  
min() - minimum value  
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**SPREAD**

IQR() - Inter-Quartile Range  
mad() - mean absolute deviation  
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## Vectorized Functions

**TO USE WITH MUTATE ()**

**mutate()** and **transmute()** apply vectorized functions to columns to create new columns. Vectorized functions take vectors as input and return vectors of the same length as output.

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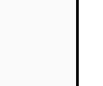
## Combine Tables

**COMBINE VARIABLES**

x      y  
 +  = 

Use bind\_cols() to paste tables beside each other as they are.

**COMBINE CASES**

x      y  
 +  = 

Use bind\_rows() to paste tables below each other as they are.

**COMBINE CASES**

**bind\_rows(..., .id = NULL)**  
Returns tables one on top of the other as a single table. Set .id to a column name to add a column of the original table names (as pictured)

**intersect(x, y, ...)**  
Rows that appear in both x and z.

**setdiff(x, y, ...)**  
Rows that appear in x but not z.

**union(x, y, ...)**  
Rows that appear in x or z. (Duplicates removed). union\_all() retains duplicates.

**use\_sequal()**  
Use setequal() to test whether two data sets contain the exact same rows (in any order).

**EXTRACT ROWS**

**left\_join(x, y, by = NULL, copy = FALSE, suffix = c("x", "y"), ...)**  
Join matching values from y to x.

**right\_join(x, y, by = NULL, copy = FALSE, suffix = c("x", "y"), ...)**  
Join matching values from x to y.

**inner\_join(x, y, by = NULL, copy = FALSE, suffix = c("x", "y"), ...)**  
Join data. Retain only rows with matches.

**full\_join(x, y, by = NULL, copy = FALSE, suffix = c("x", "y"), ...)**  
Join data. Retain all values, all rows.

**use\_by(c("col1", "col2"))**  
Use by = c("col1", "col2") to specify the column(s) to match on. left\_join(x, y, by = "A")

**use\_by(c("col1", "col2"))**  
Use a named vector, by = c("col1" = "col2"), to match on columns with different names in each data set. left\_join(x, y, by = c("C" = "D"))

**use\_by(c("col1", "col2"))**  
Use suffix to specify suffix to give to duplicate column names. left\_join(x, y, by = c("C" = "D"), suffix = c("1", "2"))

**semi\_join(x, y, by = NULL, ...)**  
Return rows of x that have a match in y. USEFUL TO SEE WHAT WILL BE JOINED.

**anti\_join(x, y, by = NULL, ...)**  
Return rows of x that do not have a match in y. USEFUL TO SEE WHAT WILL NOT BE JOINED.



# Your Turn 4

Use `summarize()` to compute three statistics about the data:

1. The first (minimum) year in the dataset
2. The last (maximum) year in the dataset
3. The number of countries represented in the data (Hint: use cheatsheet)



```
gapminder %>%  
  summarize(first = min(year),  
            last = max(year),  
            n_countries = n_distinct(country))  
  
# A tibble: 1 × 3  
#   first  last n_countries  
#   <dbl> <dbl>     <int>  
# 1 1952  2007     142
```



# Your Turn 5

Extract the rows where continent == "Africa" and year == 2007.

Then use summarize() and summary functions to find:

1. The number of unique countries
2. The median life expectancy



```
gapminder %>%  
  filter(continent == "Africa", year == 2007) %>%  
  summarize(n_countries = n_distinct(country), med_le = median(lifeExp))  
  
## # A tibble: 1 x 2  
#   n_countries med_life_exp  
#       <int>          <dbl>  
#1         52      52.9265
```



# Grouping cases

# group\_by()

Groups cases by common values of one or more columns.

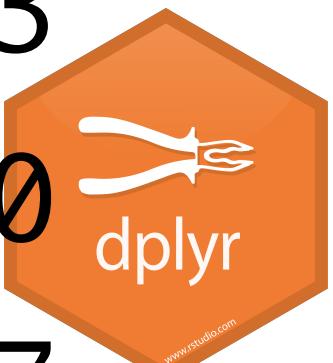
```
gapminder %>%  
  group_by(continent)
```

In console

# A tibble: 1,704 x 6

# Groups: continent [5]

	country	continent	year	lifeExp	pop	gdpPerCap
	<fctr>	<fctr>	<int>	<dbl>	<int>	<dbl>
1	Afghanistan	Asia	1952	28.801	8425333	779.4453
2	Afghanistan	Asia	1957	30.332	9240934	820.8530
3	Afghanistan	Asia	1962	31.997	10267083	853.1007



# group\_by()

Groups cases by common values, then summarize acts by group

```
gapminder %>%  
  group_by(continent) %>%  
  summarize(n_countries = n_distinct(country))
```

continent	n_countries
Africa	52
Americas	25
Asia	33
Europe	30
Oceania	2

# Manipulate Data Notebook

```
1 ---  
2 title: "Manipulate Data"  
3 output: html_document  
4 ---  
5  
6 ```{r setup}  
7 library(tidyverse)  
8 library(babynames)  
9  
10 # Toy dataset to use  
11 pollution <- tribble(  
12   ~city, ~size, ~amount,  
13   "New York", "large", 23,  
14   "New York", "small", 14,  
15   "London", "large", 22,  
16   "London", "small", 16,  
17   "Beijing", "large", 121,  
18   "Beijing", "small", 56  
19 )  
20 ...  
21  
22 ## babynames  
23  
24 ```{r}  
25 babynames
```

```
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  
)
```

Toy data set to practice with



## pollution

```
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  
)
```

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

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



mean	sum	n
42	252	6

pollution %>%

summarize(mean = mean(amount), sum = sum(amount), n = n())

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

mean	sum	n
42	252	6

city	particle size	amount ( $\mu\text{g}/\text{m}^3$ )
New York	large	23
New York	small	14



mean	sum	n
18.5	37	2

London	large	22
London	small	16



19.0	38	2
------	----	---

Beijing	large	121
Beijing	small	56



88.5	177	2
------	-----	---

# group\_by() + summarize()

# group\_by()

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

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

city	mean	sum	n
New York	18.5	37	2
London	19.0	38	2
Beijing	88.5	177	2

```
pollution %>%  
  group_by(city) %>%  
  summarize(mean = mean(amount), sum = sum(amount), n = n())
```

# Your Turn 6

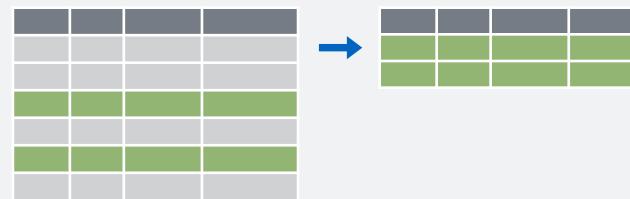
Find the median life expectancy by continent:

- \* For 2007,
- \* Or, over all years



```
gapminder %>%  
  group_by(continent) %>%  
  summarize(med_life_exp = median(lifeExp))  
  
# # A tibble: 5 × 2  
#   continent med_life_exp  
#   <fctr>        <dbl>  
# 1 Africa      47.7920  
# 2 Americas    67.0480  
# 3 Asia        61.7915  
# 4 Europe      72.2410  
# 5 Oceania     73.6650
```

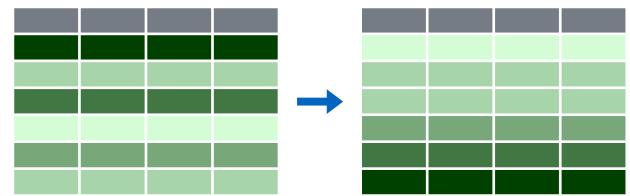
# dplyr: Data manipulation verbs



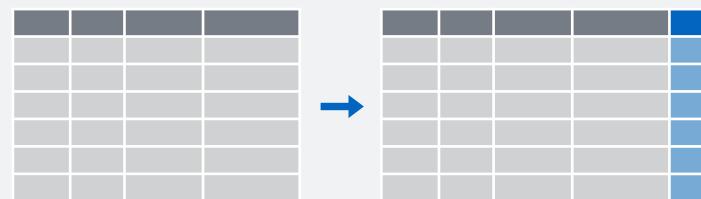
Extract cases with `filter()`



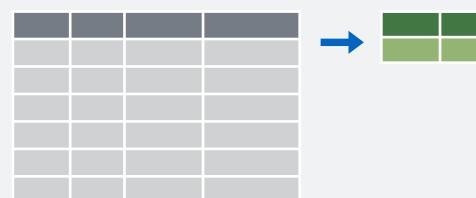
Extract variables with `select()`



Arrange cases, with `arrange()`.



Make new variables, with `mutate()`.



Make tables of summaries with `summarize()`.

along with `group_by()`

