

## **6. Exploratory Data Analysis - dplyr**

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# Overview

- Visualisation is an important tool for insight generation, but it's rare that you get the data in exactly the right form you need" (Wickham and Grolemund 2017)
  - Create new variables
  - Create summaries
  - Order data
- dplyr package is designed for data transformation

# dplyr

- All verbs (functions) work similarly
- The first argument is a data frame/tibble
- The subsequent arguments decide what to do with the data frame
- The result is a data frame (supports chaining of steps)

Function	Purpose
<b>filter()</b>	Pick observations by their values
<b>arrange()</b>	Reorder the rows
<b>select()</b>	Pick variables by their names
<b>mutate()</b>	Create new variables with functions of existing variables
<b>summarise()</b>	Collapse many values down to a single summary

## Sample Data set ggplot2::mpg

```
## Observations: 234
## Variables: 11
## $ manufacturer <chr> "audi", "audi", "audi", "audi", "audi"
## $ model         <chr> "a4", "a4", "a4", "a4", "a4", "a4", "a4"
## $ displ         <dbl> 1.8, 1.8, 2.0, 2.0, 2.8, 2.8, 3.1, 1.8
## $ year          <int> 1999, 1999, 2008, 2008, 1999, 1999, 2008
## $ cyl           <int> 4, 4, 4, 4, 6, 6, 6, 4, 4, 4, 4, 6, 6
## $ trans         <chr> "auto(l5)", "manual(m5)", "manual(m6)"
## $ drv           <chr> "f", "f", "f", "f", "f", "f", "f", "f", "f", "f", "f", "f", "f"
## $ cty           <int> 18, 21, 20, 21, 16, 18, 18, 18, 16, 20, 19, 17, 18
## $ hwy           <int> 29, 29, 31, 30, 26, 26, 27, 26, 25, 28, 26, 23, 26
## $ fl            <chr> "p", "p", "p", "p", "p", "p", "p", "p", "p", "p", "p", "p", "p"
## $ class         <chr> "compact", "compact", "compact", "compact", "compact", "compact", "compact", "compact", "compact", "compact", "compact", "compact", "compact"
```

# (1) filter()

- Subset observations based on their values.
- First argument the name of the data frame
- Subsequent arguments are expressions that filter the data frame
- Only includes rows that have no missing values

```
filter(mpg,manufacturer=="audi",year==1999,model=="a4")
```

```
## # A tibble: 4 x 11
```

```
##   manufacturer model displ  year   cyl trans  drv      cty  
##   <chr>          <chr> <dbl> <int> <int> <chr> <chr> <int> <chr>  
## 1 audi          a4      1.8  1999     4 auto(~ f      18  
## 2 audi          a4      1.8  1999     4 manua~ f      21  
## 3 audi          a4      2.8  1999     6 auto(~ f      16  
## 4 audi          a4      2.8  1999     6 manua~ f      18
```

# Cars with highest mpg, lowest mpg?

```
filter(mpg,hwy==max(hwy))
```

```
## # A tibble: 2 x 11
```

##	manufacturer	model	displ	year	cyl	trans	drv	cty
##	<chr>	<chr>	<dbl>	<int>	<int>	<chr>	<chr>	<int>
## 1	volkswagen	jetta	1.9	1999	4	manu~	f	33
## 2	volkswagen	new b~	1.9	1999	4	manu~	f	35

```
filter(mpg,hwy==min(hwy))
```

```
## # A tibble: 5 x 11
```

##	manufacturer	model	displ	year	cyl	trans	drv	cty
##	<chr>	<chr>	<dbl>	<int>	<int>	<chr>	<chr>	<int>
## 1	dodge	dakot~	4.7	2008	8	auto~	4	9
## 2	dodge	duran~	4.7	2008	8	auto~	4	9
## 3	dodge	ram 1~	4.7	2008	8	auto~	4	9
## 4	dodge	ram 1~	4.7	2008	8	manu~	4	9

## Challenge 2.1

- List the cars with an average city mpg greater than the median.
- Show the cars with the maximum displacement

## (2) arrange()

- Changes the order of rows.
- Takes a data frame and a set of column names to order by

```
arrange(mpg,displ)
```

```
## # A tibble: 234 x 11
```

```
##   manufacturer model displ  year   cyl trans  drv    cty
##   <chr>          <chr> <dbl> <int> <int> <chr> <chr> <int> <chr>
## 1 honda         civic   1.6  1999     4 manu~ f      28
## 2 honda         civic   1.6  1999     4 auto~ f      24
## 3 honda         civic   1.6  1999     4 manu~ f      25
## 4 honda         civic   1.6  1999     4 manu~ f      23
## 5 honda         civic   1.6  1999     4 auto~ f      24
## 6 audi          a4      1.8  1999     4 auto~ f      18
## 7 audi          a4      1.8  1999     4 manu~ f      21
## 8 audi          a4 q~    1.8  1999     4 manu~ 4      18
## 9 audi          a4 q~    1.8  1999     4 auto~ 4      16
```



# Show in descending order

```
arrange(mpg, desc(displ))
```

```
## # A tibble: 234 x 11
```

```
##   manufacturer model displ  year   cyl trans drv      cty  
##   <chr>          <chr> <dbl> <int> <int> <chr> <chr> <int> <chr>  
## 1 chevrolet     corv~    7    2008     8 manu~ r      15  
## 2 chevrolet     k150~   6.5  1999     8 auto~ 4      14  
## 3 chevrolet     corv~   6.2  2008     8 manu~ r      16  
## 4 chevrolet     corv~   6.2  2008     8 auto~ r      15  
## 5 jeep          gran~   6.1  2008     8 auto~ 4      11  
## 6 chevrolet     c150~    6    2008     8 auto~ r      12  
## 7 dodge         dura~   5.9  1999     8 auto~ 4      11  
## 8 dodge         ram ~   5.9  1999     8 auto~ 4      11  
## 9 chevrolet     c150~   5.7  1999     8 auto~ r      13  
## 10 chevrolet    corv~   5.7  1999     8 manu~ r      16  
## # ... with 224 more rows
```

# Add an extra sort column

```
arrange(mpg, desc(year), desc(displ))
```

```
## # A tibble: 234 x 11
##   manufacturer model displ  year   cyl trans  drv      cty
##   <chr>          <chr> <dbl> <int> <int> <chr> <chr> <int> <chr>
## 1 chevrolet      corv~    7   2008     8 manu~ r      15
## 2 chevrolet      corv~   6.2  2008     8 manu~ r      16
## 3 chevrolet      corv~   6.2  2008     8 auto~ r      15
## 4 jeep           gran~   6.1  2008     8 auto~ 4      11
## 5 chevrolet      c150~    6   2008     8 auto~ r      12
## 6 dodge          dura~   5.7  2008     8 auto~ 4      13
## 7 dodge          ram ~   5.7  2008     8 auto~ 4      13
## 8 jeep           gran~   5.7  2008     8 auto~ 4      13
## 9 toyota         land~   5.7  2008     8 auto~ 4      13
## 10 nissan         path~   5.6  2008     8 auto~ 4      12
## # ... with 224 more rows
```

### (3) select()

- It is not uncommon to get datasets with hundreds, or even thousands, of variables
- A challenge is to narrow down on the variables of you're interested in
- `select()` allows you to rapidly zoom in on a useful subset using operations based on the variable names

```
select(mpg,model,year,displ, cty, hwy)
```

```
## # A tibble: 234 x 5
```

```
##      model      year displ   cty   hwy
##      <chr>    <int> <dbl> <int> <int>
##  1 a4        1999   1.8    18    29
##  2 a4        1999   1.8    21    29
##  3 a4        2008    2     20    31
##  4 a4        2008    2     21    30
##  5 a4        1999   2.8    16    26
##  6 a4        1999   2.8    18    26
```

# Special Function with select

## Special functions

As well as using existing functions like `:` and `c`, there are a number of special functions that only work inside `select`

- `starts_with(x, ignore.case = TRUE)`: names starts with `x`
- `ends_with(x, ignore.case = TRUE)`: names ends in `x`
- `contains(x, ignore.case = TRUE)`: selects all variables whose name contains `x`
- `matches(x, ignore.case = TRUE)`: selects all variables whose name matches the regular expression `x`
- `num_range("x", 1:5, width = 2)`: selects all variables (numerically) from `x01` to `x05`.
- `one_of("x", "y", "z")`: selects variables provided in a character vector.
- `everything()`: selects all variables.

## (4) mutate()

- It is often useful to add new columns that are functions of existing columns
- `mutate()` always adds new columns at the end of your data set.

```
sml <- select(mpg,model,displ,cty)
sml <- mutate(sml,Category=ifelse(cty>mean(cty),"AboveAvr","Be
sml
```

```
## # A tibble: 234 x 4
##   model      displ   cty Category
##   <chr>      <dbl> <int> <chr>
## 1 a4          1.8    18 AboveAvr
## 2 a4          1.8    21 AboveAvr
## 3 a4          2      20 AboveAvr
## 4 a4          2      21 AboveAvr
## 5 a4          2.8    16 BelowAvr
## 6 a4          2.8    18 AboveAvr
```

# Useful creation functions

- There are many functions for creating new variables that can be used with `mutate()`
- The key property is that the function must be vectorised:
  - It must take a vector of values as input, and,
  - Return a vector with the same number of values as output

Grouping	Examples
<b>Arithmetic Operators</b>	<code>+</code> , <code>-</code> , <code>*</code> , <code>/</code> , <code>^</code>
<b>Modular Arithmetic</b>	<code>%/%</code> - Integer division <code>&amp;&amp;</code> - Remainder
<b>Logs</b>	<code>log()</code> , <code>log2()</code> , <code>log10()</code>
<b>Offsets</b>	<code>lead()</code> and <code>lag()</code> Find when values change <code>x!=lag(x)</code>
<b>Cumulative and rolling aggregates</b>	<code>cumsum()</code> , <code>cumprod()</code> , <code>cummin()</code> , <code>cummax()</code> , <code>cummean()</code>
<b>Logical comparisons</b>	<code>&lt;</code> , <code>&lt;=</code> , <code>&gt;</code> , <code>&gt;=</code> , <code>!=</code>
<b>Ranking</b>	<code>min_rank()</code>

## (5) summarise()

- The last key verb is summarise()
- It collapses a data frame into a single row
- Not very useful unless paired with group\_by()
- Very useful to combine with the pipe operator %>%
- The pipe %>% comes from the magrittr package (Stefan Milton Bache)
- Helps to write code that is easier to read and understand
  - $x \%>\% f(y)$  turns into  $f(x, y)$

```
mpg %>% select(model, displ, cty) %>% slice(1:2)
```

```
## # A tibble: 2 x 3
##   model displ  cty
##   <chr> <dbl> <int>
## 1 a4     1.8     18
## 2 a4     1.8     21
```

# The function `group_by()`

- Most summary data operations are useful done on groups defined by variables in the the dataset.
- The `group_by` function takes an existing `tbl` and converts it into a grouped `tbl` where operations can then performed “by group”.

```
gr <- group_by(mpg,year)
agg <- summarise(gr,AverageCty=mean(cty))
agg
```

```
## # A tibble: 2 x 2
##   year AverageCty
##   <int>      <dbl>
## 1  1999      17.0
## 2  2008      16.7
```

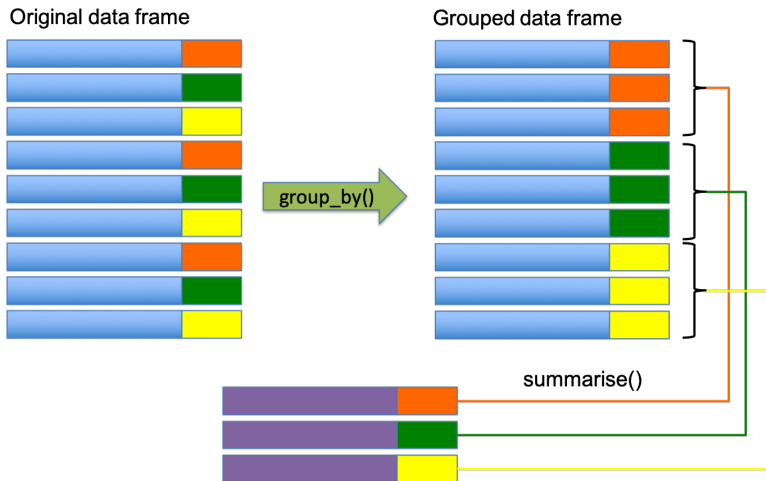


# Using %>%

```
mpg %>% group_by(manufacturer) %>%  
  summarise(AvrCty=mean(cty),N=n()) %>%  
  arrange(desc(AvrCty)) %>%  
  slice(1:5)
```

```
## # A tibble: 5 x 3  
##   manufacturer AvrCty      N  
##   <chr>         <dbl> <int>  
## 1 honda         24.4      9  
## 2 volkswagen    20.9     27  
## 3 subaru        19.3     14  
## 4 hyundai       18.6     14  
## 5 toyota        18.5     34
```

# Overall idea



# Useful Summary Functions

Grouping	Examples
<b>Measures of location</b>	mean(), median()
<b>Measures of spread</b>	sd(), IQR(),mad()
<b>Measures of rank</b>	min(), quantile(), max()
<b>Measures of position</b>	first(), nth(), last()
<b>Counts</b>	n(), n_distinct()
<b>Counts and proportions of logical values</b>	sum(x>0) when used with numeric functions, (T,F) converted to (1,0)

# The package nycflights13

```
glimpse(nycflights13::flights)
```

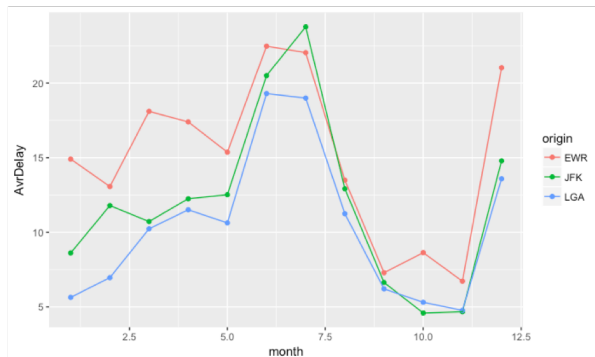
```
## Observations: 336,776
## Variables: 19
## $ year          <int> 2013, 2013, 2013, 2013, 2013, 2013,
## $ month         <int> 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,
## $ day           <int> 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,
## $ dep_time      <int> 517, 533, 542, 544, 554, 554, 555, 5
## $ sched_dep_time <int> 515, 529, 540, 545, 600, 558, 600, 6
## $ dep_delay     <dbl> 2, 4, 2, -1, -6, -4, -5, -3, -3, -2,
## $ arr_time      <int> 830, 850, 923, 1004, 812, 740, 913,
## $ sched_arr_time <int> 819, 830, 850, 1022, 837, 728, 854,
## $ arr_delay     <dbl> 11, 20, 33, -18, -25, 12, 19, -14, -
## $ carrier       <chr> "UA", "UA", "AA", "B6", "DL", "UA",
## $ flight        <int> 1545, 1714, 1141, 725, 461, 1696, 50
## $ tailnum       <chr> "N14228", "N24211", "N619AA", "N804J
```

## Challenge 6.1 | nycflights13::flights

Generate the following graph. Use the variable **dep\_delay**. The variable **origin** indicates the departure airport.

```
unique(nycflights13::flights$origin)
```

```
## [1] "EWR" "LGA" "JFK"
```



# Summary

- dplyr - a grammar of data manipulation
- Five verbs
  - **filter()**
  - **arrange()**
  - **select()**
  - **mutate()**
  - **summarise()** (along with **group\_by()**)
- Usefully combined with **%>%** operator