

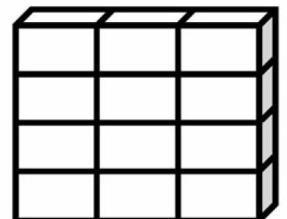
数据类型 (Data Types)

(a) 向量



```
a <- c(1, 2, 5, 3, 6, -2, 4)
b <- c("one", "two", "three")
c <- c(TRUE, TRUE)
```

(d) 数据框

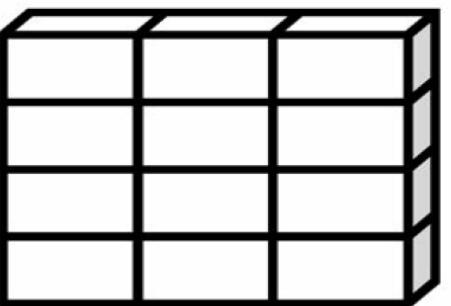


各列的模式 (modes) 可以不同

```
patientID <- c(1, 2, 3, 4)
age <- c(25, 34, 28, 52)
diabetes <- c("Type1", "Type2", "Type1", "Type1")
status <- c("Poor", "Improved", "Excellent", "Poor")
patientdata <- data.frame(patientID, age, diabetes, status)
patientdata
```

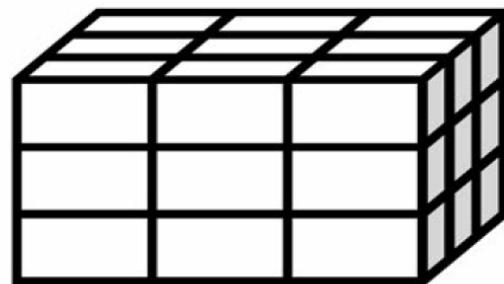
	patientID	age	diabetes	status
1	1	25	Type1	Poor
2	2	34	Type2	Improved
3	3	28	Type1	Excellent
4	4	52	Type1	Poor

(b) 矩阵



```
y <- matrix(1:12, nrow=3, ncol=4)
y
[1] [2] [3] [4]
[1,] 1   4   7   10
[2,] 2   5   8   11
[3,] 3   6   9   12
```

(c) 数组



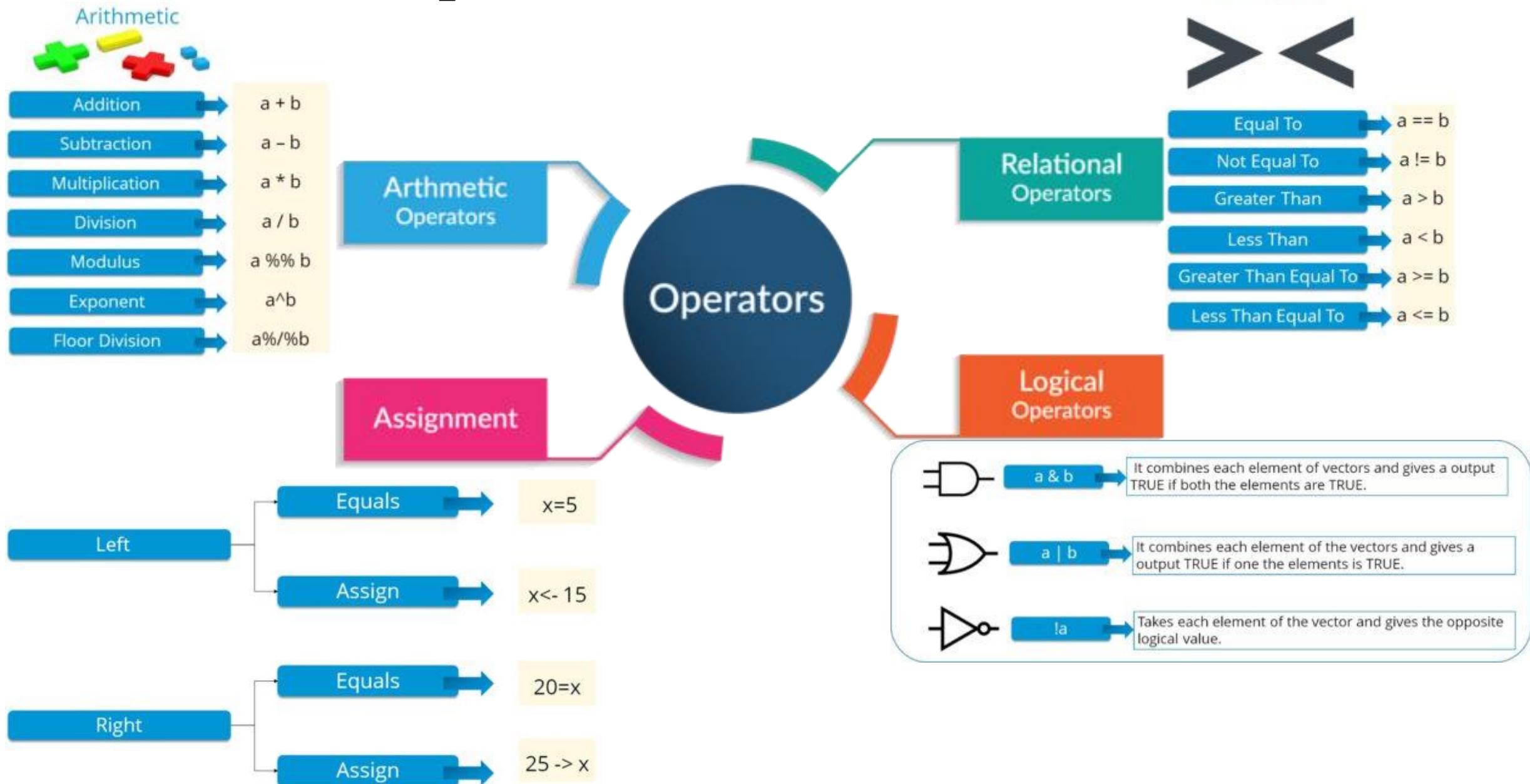
```
dim1 <- c("A1", "A2")
dim2 <- c("B1", "B2", "B3")
dim3 <- c("C1", "C2")
z <- array(1:12, c(2, 3, 2),
           dimnames=list(dim1, dim2, dim3))
z
,, C1
  B1 B2 B3
A1 1   3   5
A2 2   4   6
,, C2
  B1 B2 B3
A1 7   9   11
A2 8   10  12
```

(e) 列表

向量
数组
数据框
列表

```
g <- "My First List"
h <- c(25, 26, 18, 39)
j <- matrix(1:4, nrow=2)
k <- c("one", "two", "three")
mylist <- list(title=g, ages=h, j, k)
mylist
$title
[1] "My First List"
$ages
[1] 25 26 18 39
[[3]]
[1] [2]
[1,] 1   3
[2,] 2   4
[[4]]
[1] "one" "two" "three"
```

数据操作 (Data Operators)



函数(Functions)

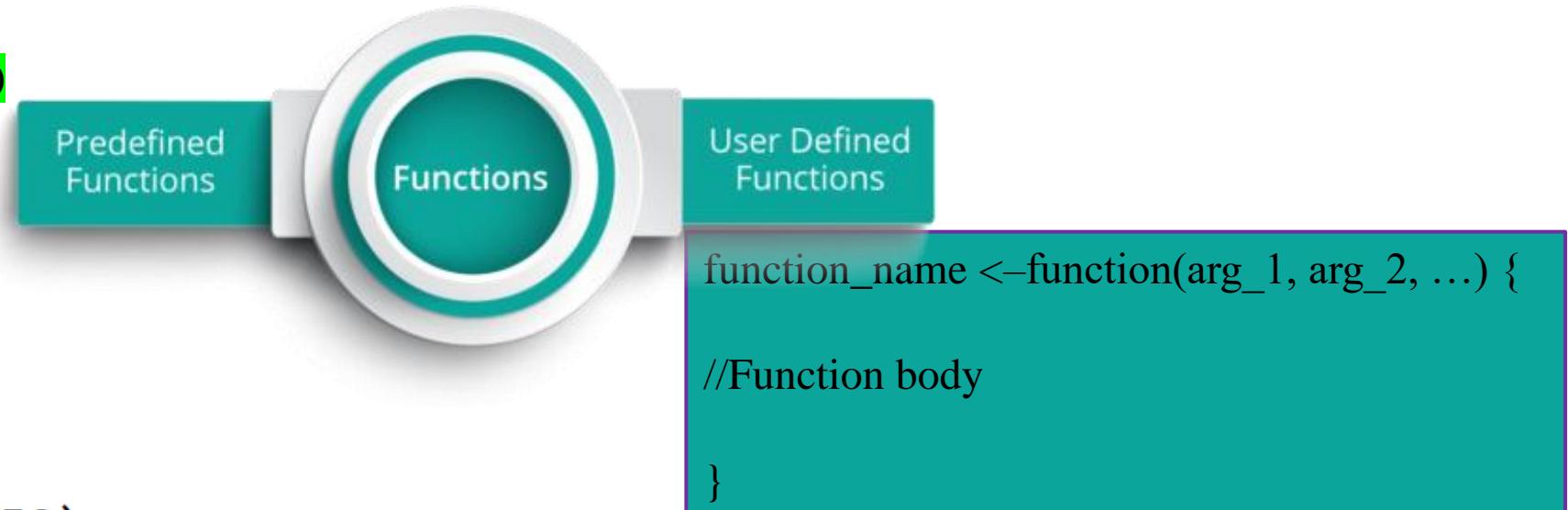
求平均: mean(x)

求 和: sum(x)

开平方: sqrt(x)

.....

```
1 > example("mean")
2
3 mean> x <- c(0:10, 50)
4
5 mean> xm <- mean(x)
6
7 mean> c(xm, mean(x, trim = 0.10))
8 [1] 8.75 5.50
9
```



```
1 □ sum_of_square <- function(x,y) {  
2 | x^2 + y^2  
3 |}  
4 sum_of_squares(3,4)  
5
```

R 包 (R packages)

R packages are collections of functions and data sets developed by the community.



CRAN

```
install.packages("XXX")
```



Bioconductor

```
source("https://bioconductor.org/biocLite.R")  
biocLite("XXX")
```

library(XXX)



```
install.packages("devtools")  
devtools::install_github("hadley/babynames")
```

help(XXX)

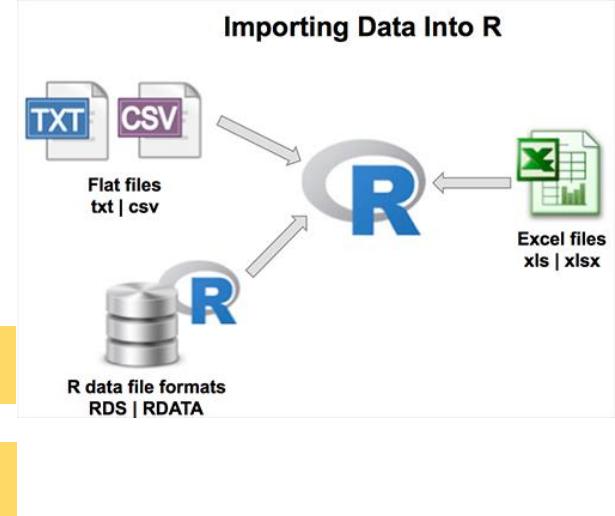


search()

读取数据 (Import from files)

R base functions:

```
1 # Command to copy&paste tables from Excel or other programs into R.  
2 read.delim("clipboard", header=T)  
3  
4 # Reads in table and assigns it to data frame, with info on column headers and field separators.  
5 read.table(file="file", header=TRUE, sep="\t")  
6  
7 # Reads a file in table format and creates a data frame from  
8 read.csv("file", header = T)  
9
```



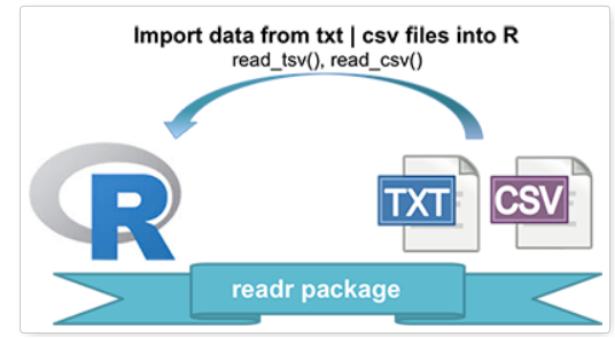
- file参数：这是必须的，可以是相对路径或者绝对路径（注意：Windows下路径要用斜杠'/'或者双反斜杠'\\'）。
- 如果数据集中含有中文，直接导入很有可能不识别中文，这时加上参fileEncoding='utf-8'

R package (readr):

```
1 install.packages("readr")  
2 # Loading  
3 library("readr")  
4 read_csv(file, col_names = TRUE)  
5 read_tsv(file, col_names = TRUE)  
6
```

Compared to R base functions, readr functions are:

- ❖ much faster (X10),
- ❖ have a helpful progress bar if loading is going to take a while
- ❖ all functions work exactly the same way.



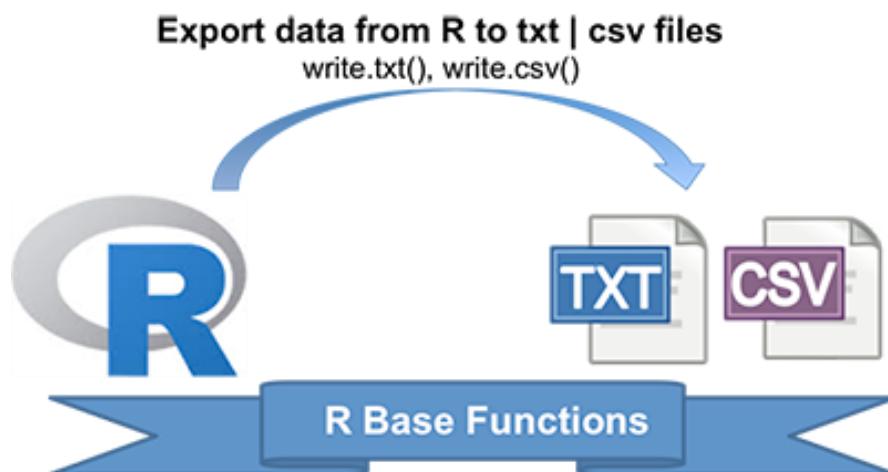
导出数据 (Export data from R)

R base functions:

```
1 data("mtcars")
2 write.table(mtcars, file = "mtcars.txt", sep = "\t",
3             row.names = TRUE, col.names = NA)
4 -----
5 write.csv(mtcars, file = "mtcars.csv")
6
7
```

R package (readr):

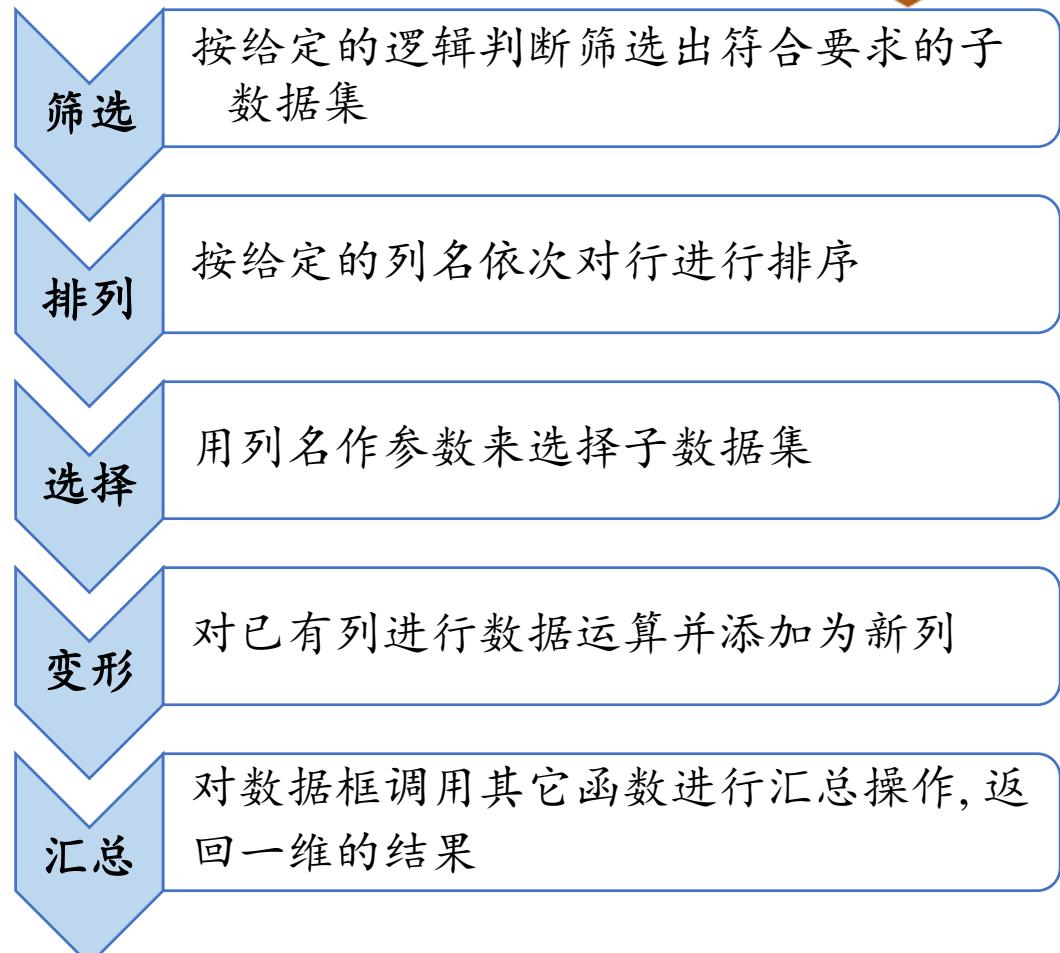
```
1 write_csv(challenge, "challenge.csv")
2
```



数据转换 (Data transformation)

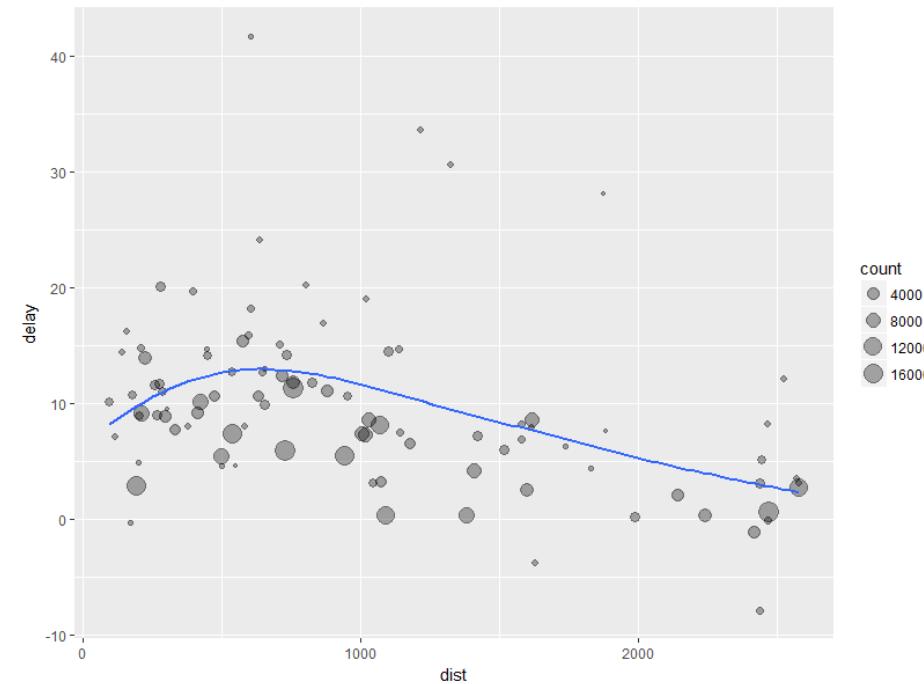
清洗和整理

```
1 library(nycflights13)
2 library(tidyverse)
3 # 1.1 筛选: filter()
4 (jan1 <- filter(flights, month == 1, day == 1))
5 # 1.2 排列: arrange()
6 arrange(flights, year, month, day)
7 arrange(flights, desc(arr_delay))
8 # 1.3 选择: select()
9 select(flights, year, month, day)
10 # 1.4 变形: mutate()
11 flights_sml <- select(flights,
12   year:day,
13   ends_with("delay"),
14   distance,
15   air_time )
16 ##### 新添加的列可以用于后续计算
17 mutate(flights_sml,
18   gain = arr_delay - dep_delay,
19   hours = air_time / 60,
20   gain_per_hour = gain / hours )
21 ##### 只保留变形后的列
22 transmute(flights,
23   gain = arr_delay - dep_delay,
24   hours = air_time / 60,
25   gain_per_hour = gain / hours )
26 # 1.5 汇总: summarise()
27 summarise(flights, delay = mean(dep_delay, na.rm = TRUE))
28 # 1.6 分组: group_by()
29 by_day <- group_by(flights, year, month, day)
30 summarise(by_day, delay = mean(dep_delay, na.rm = TRUE))
```

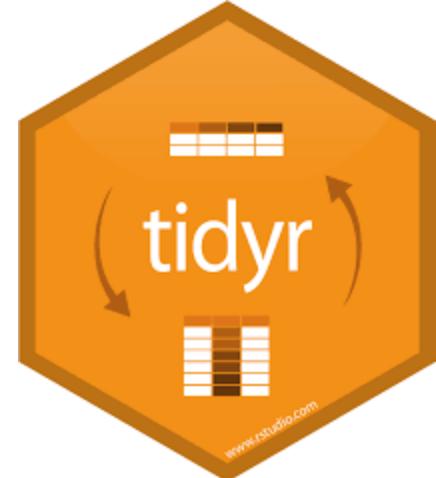


管道函数(%>%) 和 绘图

```
1  delays <- flights %>%
2    | group_by(dest) %>%
3    | summarise(
4      |   count = n(),
5      |   dist = mean(distance, na.rm = TRUE),
6      |   delay = mean(arr_delay, na.rm = TRUE)
7    ) %>%
8    | filter(count > 20, dest != "HNL")
9  ggplot(data = delay, mapping = aes(x = dist, y = delay)) +
10    | geom_point(aes(size = count), alpha = 1/3) +
11    | geom_smooth(se = FALSE)
```



数据整形 (Reshaping Data)



```
1 install.packages("tibble")
2 library("tibble")
3 # Create a new tibble use data_frame()
4 friends_data <- data_frame(
5   name = c("Nicolas", "Thierry", "Bernard", "Jerome"),
6   age = c(27, 25, 29, 26),
7   height = c(180, 170, 185, 169),
8   married = c(TRUE, FALSE, TRUE, TRUE)
9 )
10 friends_data
11 #> #> Source: local data frame [4 x 4]
12 #> #>   name    age height married
13 #>
14 #> 1 Nicolas    27    180     TRUE
15 #> 2 Thierry    25    170    FALSE
16 #> 3 Bernard    29    185     TRUE
17 #> 4 Jerome     26    169     TRUE
18 #> #> Convert your data as a tibble
19 data("iris")
20 class(iris)
21 [1] "data.frame"
22 my_data <- as_data_frame(iris)
23 class(my_data)
24 [1] "tbl_df"     "tbl"        "data.frame"
25 #> #> Turn a tibble back to a data frame
26 my_data2 <- as.data.frame(my_data)
27
```

Tibble Data Format in R: Best and Modern Way to Work with Data

```
> flights
Source: local data frame [336,776 x 16]
```

	year	month	day	dep_time	dep_delay	arr_time	arr_delay	carrier	tailnum						
	<int>	<int>	<int>	<int>	<dbl>	<int>	<dbl>	<chr>	<chr>						
1	2013	1	1	517	2	630	11	UA	N1422E						
2	2013	1	1	533	4	850	20	UA	N24211						
3	2013	1	1	542	2	923	33	AA	N619AA						
4	2013	1	1	54	4	-18	86	B6	N804JE						
5	2013	1	1	55	2	-25	DL	N668DN							
6	2013	1	1	55	0	12	UA	N39463							
7	2013	1	1	55:	3	19	B6	N516JB							
8	2013	1	1	55:	9	-14	EV	N829AS							
9	2013	1	1	55:	18	-8	B6	N593JB							
10	2013	1	1	558	-2	753	8	AA	N3ALAA						
...															

Variables not shown: flight <int>, origin <chr>, dest <chr>, arr_time <dbl>, distance <dbl>, hour <dbl>, minute <dbl>.

- `data_frame()`: create a new tibble data frame (`tbl_df`)
- `as_data_frame()`: convert your data as `tbl_df`

tibble R package

Column types

First 10 rows
printed by default
(Useful with large
data sets)

Columns that don't
fit to the screen are
not shown

```

1 > library("tidyverse")
2 > my_data <- USArrests[c(1, 10, 20, 30), ]
3 > my_data
4   Murder Assault UrbanPop Rape
5 Alabama    13.2     236      58 21.2
6 Georgia    17.4     211      60 25.8
7 Maryland   11.3     300      67 27.8
8 New Jersey  7.4      159      89 18.8
9 > my_data <- cbind(state = rownames(my_data), my_data)
10 > my_data
11   state Murder Assault UrbanPop Rape
12 Alabama    13.2     236      58 21.2
13 Georgia    17.4     211      60 25.8
14 Maryland   11.3     300      67 27.8
15 New Jersey  7.4      159      89 18.8
16

```

Tidy data

Variables

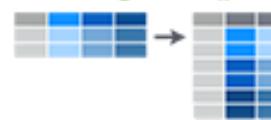
A	B	C	D	E
1 State	Murder	Assault	UrbanPop	Rape
2 Alabama	13.2	236	58	21.2
3 Alaska	10	263	48	44.5
4 Araria	8.1	294	80	1
5 Arkansas	8.8	150	50	19.5
6 California	9	276	91	40.6
7 Colorado	7.9	204	78	38.7
8 Connecticut	3.3	110	77	11.1
9 Delaware	5.9	238	72	15.8
10 Florida	15.4	335	80	31.9
11 Georgia	17.4	211	60	25.8
12 Hawaii	5.3	46	83	20.8
13 Idaho	2.6	120	54	14.2
14 Illinois	10.4	249	83	24
15 Indiana	7.2	113	65	21
16 Iowa	2.2	56	57	11.3
17 Kansas	6	115	66	18

tidy data set:

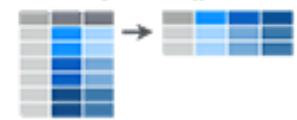
- each column represents a variable
- each row represents an observation
- ❖ The opposite of tidy is messy data

Organize Your Data for Easier Analyses in R

gather()



spread()



separate()



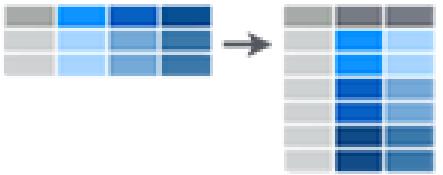
unite()



- **gather():** collapse multiple columns into key-pair values
- **spread():** reverse of gather. Separate one column into multiple
- **separate():** separate one column into multiple
- **unite():** unite multiple columns into one

tidy R package

gather() 宽数据转为长数据



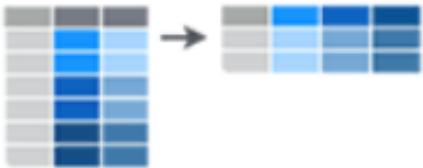
gather(data, key, value, ...)

- **data:** A data frame
- **key, value:** Names of key and value columns to create in output
- **...:** Specification of columns to gather. Allowed values are:
 - variable names
 - if you want to select all variables between a and e, use a:e
 - if you want to exclude a column name y use -y
 - for more options, see: dplyr::select()

```
1 my_data2 <- gather(my_data,
2                     key = "arrest_attribute",
3                     value = "arrest_estimate",
4                     -state)
5 # Only Murder and Assault columns,
6 # remaining columns have been duplicated.
7 my_data2 <- gather(my_data,
8                     key = "arrest_attribute",
9                     value = "arrest_estimate",
10                    Murder, Assault)
```

```
1 > my_data
2   state Murder Assault UrbanPop Rape
3   Alabama    13.2     236      58 21.2
4   Georgia    17.4     211      60 25.8
5   Maryland   11.3     300      67 27.8
6   New Jersey  7.4     159      89 18.8
7
8 > my_data2
9   state arrest_attribute arrest_estimate
10  1   Alabama           Murder          13.2
11  2   Georgia           Murder          17.4
12  3   Maryland          Murder          11.3
13  4   New Jersey        Murder          7.4
14  5   Alabama           Assault         236.0
15  6   Georgia           Assault         211.0
16  7   Maryland          Assault         300.0
17  8   New Jersey        Assault         159.0
18  9   Alabama           UrbanPop       58.0
19  10  Georgia           UrbanPop       60.0
20  11  Maryland          UrbanPop       67.0
21  12  New Jersey        UrbanPop       89.0
22  13  Alabama           Rape            21.2
23  14  Georgia           Rape            25.8
24  15  Maryland          Rape            27.8
25  16  New Jersey        Rape            18.8
26
```

spread() 长数据转为宽数据



spread(data, key, value)

- **data:** A data frame
- **key:** The (unquoted) name of the column whose values will be used as column headings.
- **value:** The (unquoted) names of the column whose values will populate the cells.

```
1 my_data3 <- spread(my_data2,
2   key = "arrest_attribute",
3   value = "arrest_estimate"
4 )
5
```

```
1 > my_data
2   state Murder Assault UrbanPop Rape
3   Alabama    Alabama  13.2    236     58 21.2
4   Georgia    Georgia  17.4    211     60 25.8
5   Maryland   Maryland 11.3    300     67 27.8
6   New Jersey New Jersey 7.4    159     89 18.8
7
8 > my_data2
9   state arrest_attribute arrest_estimate
10  1   Alabama           Murder          13.2
11  2   Georgia           Murder          17.4
12  3   Maryland          Murder          11.3
13  4   New Jersey        Murder          7.4
14  5   Alabama           Assault         236.0
15  6   Georgia           Assault         211.0
16  7   Maryland          Assault         300.0
17  8   New Jersey        Assault         159.0
18  9   Alabama           UrbanPop       58.0
19  10  Georgia           UrbanPop       60.0
20  11  Maryland          UrbanPop       67.0
21  12  New Jersey        UrbanPop       89.0
22  13  Alabama           Rape            21.2
23  14  Georgia           Rape            25.8
24  15  Maryland          Rape            27.8
25  16  New Jersey        Rape            18.8
26
27 > my_data3
28   state Assault Murder Rape UrbanPop
29  1   Alabama    236  13.2 21.2      58
30  2   Georgia    211  17.4 25.8      60
31  3   Maryland   300  11.3 27.8      67
32  4   New Jersey 159  7.4 18.8      89
33
```

unite() 多列合并为一列



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

- ✓ **data**: A data frame
- ✓ **col**: The new (unquoted) name of column to add.
- ✓ **sep**: Separator to use between values

```
1 □ my_data4 <- unite(my_data,
2 |           col = "Murder_Assault",
3 |           Murder, Assault,
4 |           sep = "_")
5
```

```
1 □ > my_data
2 |           Murder Assault UrbanPop Rape
3 |   Alabama      13.2     236      58 21.2
4 |   Georgia      17.4     211      60 25.8
5 |   Maryland     11.3     300      67 27.8
6 |   New Jersey    7.4     159      89 18.8
7 □ > my_data4
8 |           Murder_Assault UrbanPop Rape
9 |   Alabama          13.2_236      58 21.2
10 |  Georgia          17.4_211      60 25.8
11 |  Maryland          11.3_300      67 27.8
12 |  New Jersey        7.4_159      89 18.8
13
```

separate() 将一列分离为多列



```
separate(data, col, into, sep = "[^[:alnum:]]+")
```

- ✓ **data:** A data frame
- ✓ **col:** Unquoted column names
- ✓ **into:** Character vector specifying the names of new variables to be created.
- ✓ **sep:** Separator between columns:
 - ✓ If character, is interpreted as a regular expression.
 - ✓ If numeric, interpreted as positions to split at. Positive values start at 1 at the far-left of the string; negative value start at -1 at the far-right of the string.

```
1 my_data5 <- separate(my_data4,  
2   col = "Murder_Assault",  
3   into = c("Murder", "Assault"),  
4   sep = "_")  
5
```

```
1 > my_data  
2 |   Murder Assault UrbanPop Rape  
3 |   Alabama    13.2    236     58 21.2  
4 |   Georgia    17.4    211     60 25.8  
5 |   Maryland    11.3    300     67 27.8  
6 |   New Jersey  7.4     159     89 18.8  
7 > my_data4  
8 |   Murder_Assault UrbanPop Rape  
9 |   Alabama    13.2_236     58 21.2  
10 |  Georgia    17.4_211     60 25.8  
11 |  Maryland    11.3_300     67 27.8  
12 |  New Jersey  7.4_159     89 18.8  
13 > my_data5  
14 |   Murder Assault UrbanPop Rape  
15 |   Alabama    13.2    236     58 21.2  
16 |  Georgia    17.4    211     60 25.8  
17 |  Maryland    11.3    300     67 27.8  
18 |  New Jersey  7.4     159     89 18.8  
19
```

管道函数(%>%)

gather() + unite()

```
1 └ my_data6 <- my_data %>% gather(key = "arrest_attribute",
2                                     value = "arrest_estimate",
3                                     Murder:UrbanPop) %>%
4 └   unite(col = "attribute_estimate",
5           arrest_attribute, arrest_estimate)
6 -----
7 └ > my_data
8   └             Murder Assault UrbanPop Rape
9     Alabama      13.2    236      58 21.2
10    Georgia      17.4    211      60 25.8
11    Maryland      11.3    300      67 27.8
12    New Jersey    7.4    159      89 18.8
13 └ > my_data6
14   └   Rape attribute_estimate
15     1 21.2          Murder_13.2
16     2 25.8          Murder_17.4
17     3 27.8          Murder_11.3
18     4 18.8          Murder_7.4
19     5 21.2          Assault_236
20     6 25.8          Assault_211
21     7 27.8          Assault_300
22     8 18.8          Assault_159
23     9 21.2          UrbanPop_58
24    10 25.8          UrbanPop_60
25    11 27.8          UrbanPop_67
26    12 18.8          UrbanPop_89
27
```

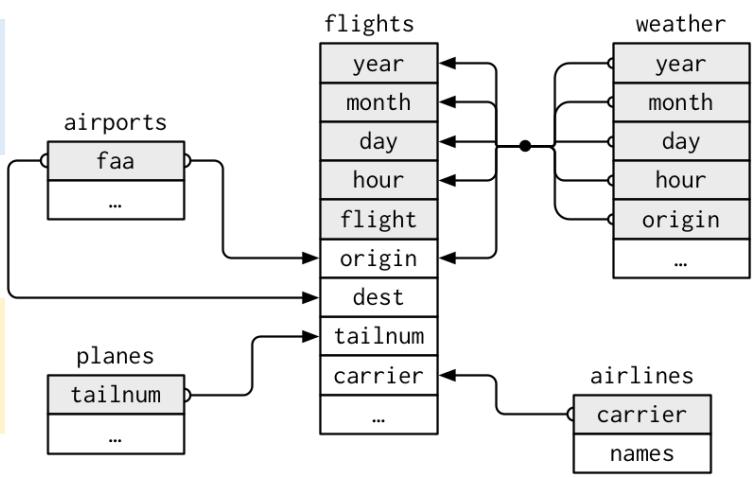


关系型数据 (Relational data)

```
1 library(nycflights13)
2 > class(flights)
3 [1] "tbl_df"     "tbl"        "data.frame"
4 > flights
5 # A tibble: 336,776 x 19
6   year month   day dep_time sched_dep_time dep_delay arr_time sched_arr_time arr_delay carrier flight tailnum origin dest
7   <int> <int> <int>    <int>      <dbl>    <int>    <int>      <dbl>    <chr>   <int>   <chr>   <chr>   <chr>
8   1  2013     1     1      517       515      2     830      819     11     UA  1545 N14228  EWR  IAH
9   2  2013     1     1      533       529      4     850      830     20     UA  1714 N24211  LGA  IAH
10  3  2013     1     1      542       540      2     923      850     33     AA  1141 N619AA  JFK  MIA
11  4  2013     1     1      544       545     -1    1004     1022    -18     B6   725  N804JB  JFK  BQN
12  5  2013     1     1      554       600     -6     812      837     -25     DL   461  N668DN  LGA  ATL
13  6  2013     1     1      554       558     -4     740      728     12     UA  1696 N39463  EWR  ORD
14  7  2013     1     1      555       600     -5     913      854     19     B6   507  N516JB  EWR  FLL
15  8  2013     1     1      557       600     -3     709      723     -14     EV   5708 N829AS  LGA  IAD
16  9  2013     1     1      557       600     -3     838      846     -8     B6    79  N593JB  JFK  MCO
17 10  2013     1     1      558       600     -2     753      745      8     AA   301  N3ALAA  LGA  ORD
18 # ... with 336,766 more rows, and 5 more variables: air_time <dbl>, distance <dbl>, hour <dbl>, minute <dbl>, time_hour <dttm>
19 planes %>% count(tailnum) %>% filter(n > 1) # verify the key
20
```

key primary key: uniquely identifies an observation in its own table.
planes\$tailnum

foreign key: uniquely identifies an observation in another table.
flights\$tailnum

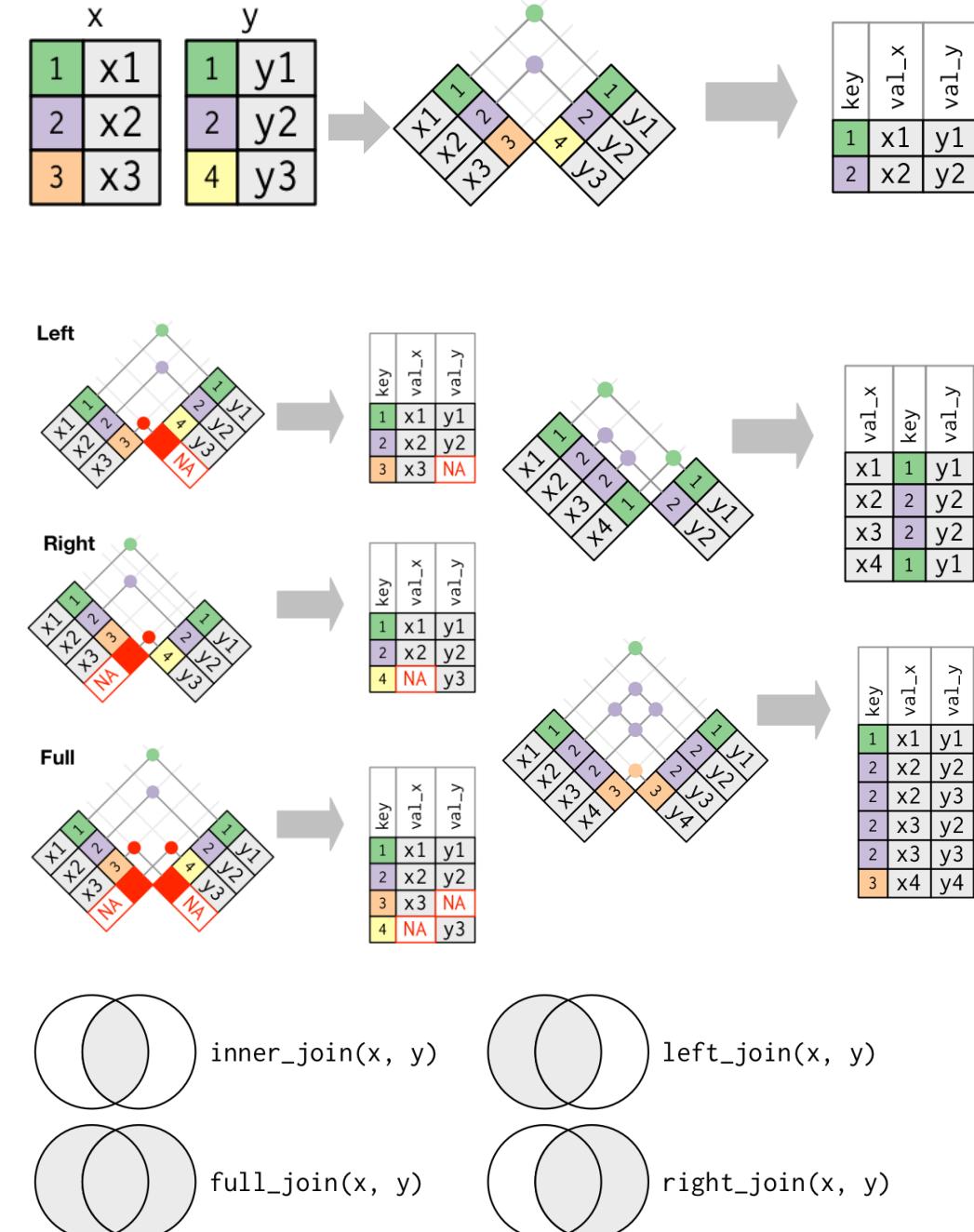


Mutating joins

```

1 > flights2 <- flights %>% select(year:day, hour, origin, dest, tailnum, carrier)
2 > flights2
3 # A tibble: 336,776 x 8
4   year month day hour origin dest tailnum carrier
5   <int> <int> <int> <dbl> <chr> <chr> <chr>
6 1 2013     1    1     5   EWR   IAH   N14228   UA
7 2 2013     1    1     5   LGA   IAH   N24211   UA
8 3 2013     1    1     5   JFK   MIA   N619AA   AA
9 4 2013     1    1     5   JFK   BQN   N804JB   B6
10 5 2013     1    1     6   LGA   ATL   N668DN   DL
11 6 2013     1    1     5   EWR   ORD   N39463   UA
12 7 2013     1    1     6   EWR   FLL   N516JB   B6
13 8 2013     1    1     6   LGA   IAD   N829AS   EV
14 9 2013     1    1     6   JFK   MCO   N593JB   B6
15 10 2013    1    1     6   LGA   ORD   N3ALAA   AA
16 # ... with 336,766 more rows
17 > flights2 %>% select(-origin, -dest) %>% left_join(airlines, by = "carrier")
18 # A tibble: 336,776 x 7
19   year month day hour tailnum carrier          name
20   <int> <int> <int> <dbl> <chr> <chr> <chr>
21 1 2013     1    1     5   N14228   UA  United Air Lines Inc.
22 2 2013     1    1     5   N24211   UA  United Air Lines Inc.
23 3 2013     1    1     5   N619AA   AA  American Airlines Inc.
24 4 2013     1    1     5   N804JB   B6  JetBlue Airways
25 5 2013     1    1     6   N668DN   DL  Delta Air Lines Inc.
26 6 2013     1    1     5   N39463   UA  United Air Lines Inc.
27 7 2013     1    1     6   N516JB   B6  JetBlue Airways
28 8 2013     1    1     6   N829AS   EV  ExpressJet Airlines Inc.
29 9 2013     1    1     6   N593JB   B6  JetBlue Airways
30 10 2013    1    1     6   N3ALAA   AA  American Airlines Inc.
31 # ... with 336,766 more rows
32

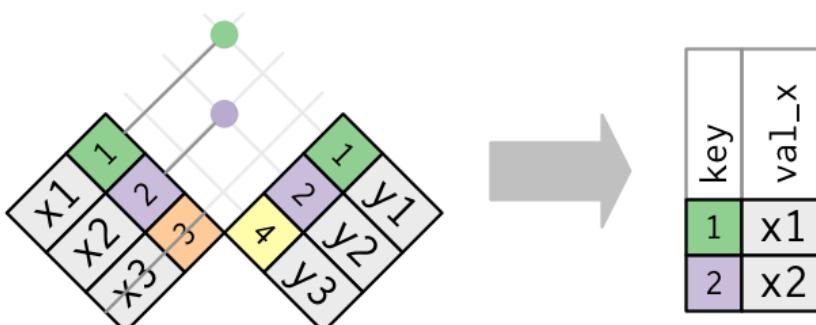
```



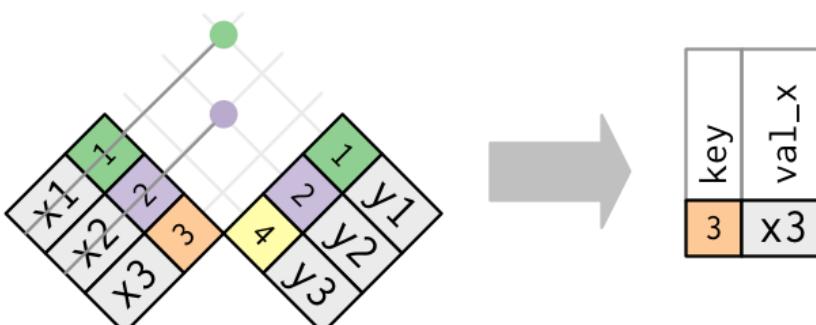
Filtering joins

```
1 > top_dest <- flights %>% count(dest, sort = TRUE) %>% head(10)
2 > flights %>% semi_join(top_dest)
3 > dim(flights)
4 [1] 336776    19
5 > dim(flights %>% semi_join(top_dest))
6 [1] 80262    19
7 > flights %>% anti_join(planes, by = "tailnum") %>% count(tailnum, sort = TRUE)
8 # A tibble: 722 x 2
9   tailnum     n
10  <chr> <int>
11  1      <NA>  2512
12  2      N725MQ  575
13  3      N722MQ  513
14  4      N723MQ  507
15  5      N713MQ  483
16  6      N735MQ  396
17 # ... with 712 more rows
18
```

semi_join(x, y): keeps all observations in x that have a match in y



anti_join(x, y): drops all observations in x that have a match in y



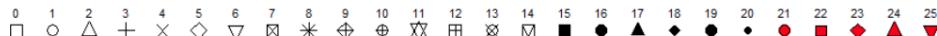
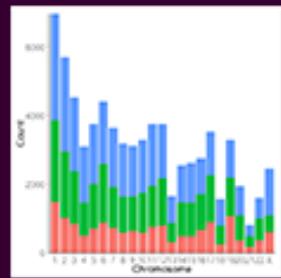
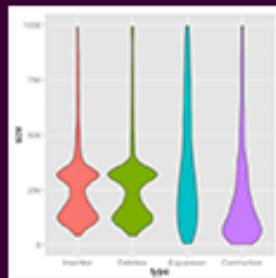
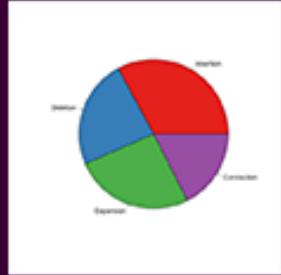
Set operations

```
1  df1 <- tribble(  
2    ~x, ~y,  
3    1, 1,  
4    2, 1  
5  )  
6  df2 <- tribble(  
7    ~x, ~y,  
8    1, 1,  
9    1, 2  
10 )  
11 > df1  
12 # A tibble: 2 x 2  
13 |   x     |     y |  
14 | <dbl> <dbl> |  
15 | 1     |     1 |  
16 | 2     |     1 |  
17 > df2  
18 # A tibble: 2 x 2  
19 |   x     |     y |  
20 | <dbl> <dbl> |  
21 | 1     |     1 |  
22 | 2     |     2 |  
23
```

- ✓ **intersect(x, y):** return only observations in both x and y.
- ✓ **union(x, y):** return unique observations in x and y.
- ✓ **setdiff(x, y):** return observations in x, but not in y.

```
1  > intersect(df1, df2)  
2  # A tibble: 1 x 2  
3  |   x     |     y |  
4  | <dbl> <dbl> |  
5  | 1     |     1 |  
6  |-----|  
7  > union(df1, df2)  
8  # A tibble: 3 x 2  
9  |   x     |     y |  
10 | <dbl> <dbl> |  
11 | 1     |     1 |  
12 | 2     |     2 |  
13 | 3     |     1 |  
14 |-----|  
15 > setdiff(df1, df2)  
16 # A tibble: 1 x 2  
17 |   x     |     y |  
18 | <dbl> <dbl> |  
19 | 1     |     2 |  
20
```

Plotting in R for Biologists



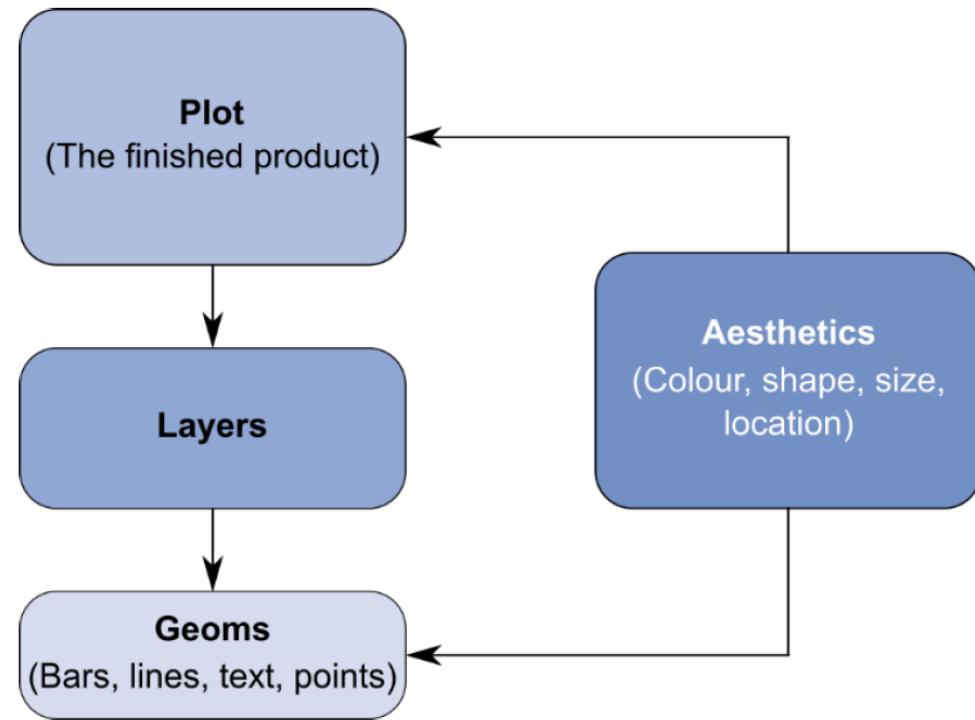
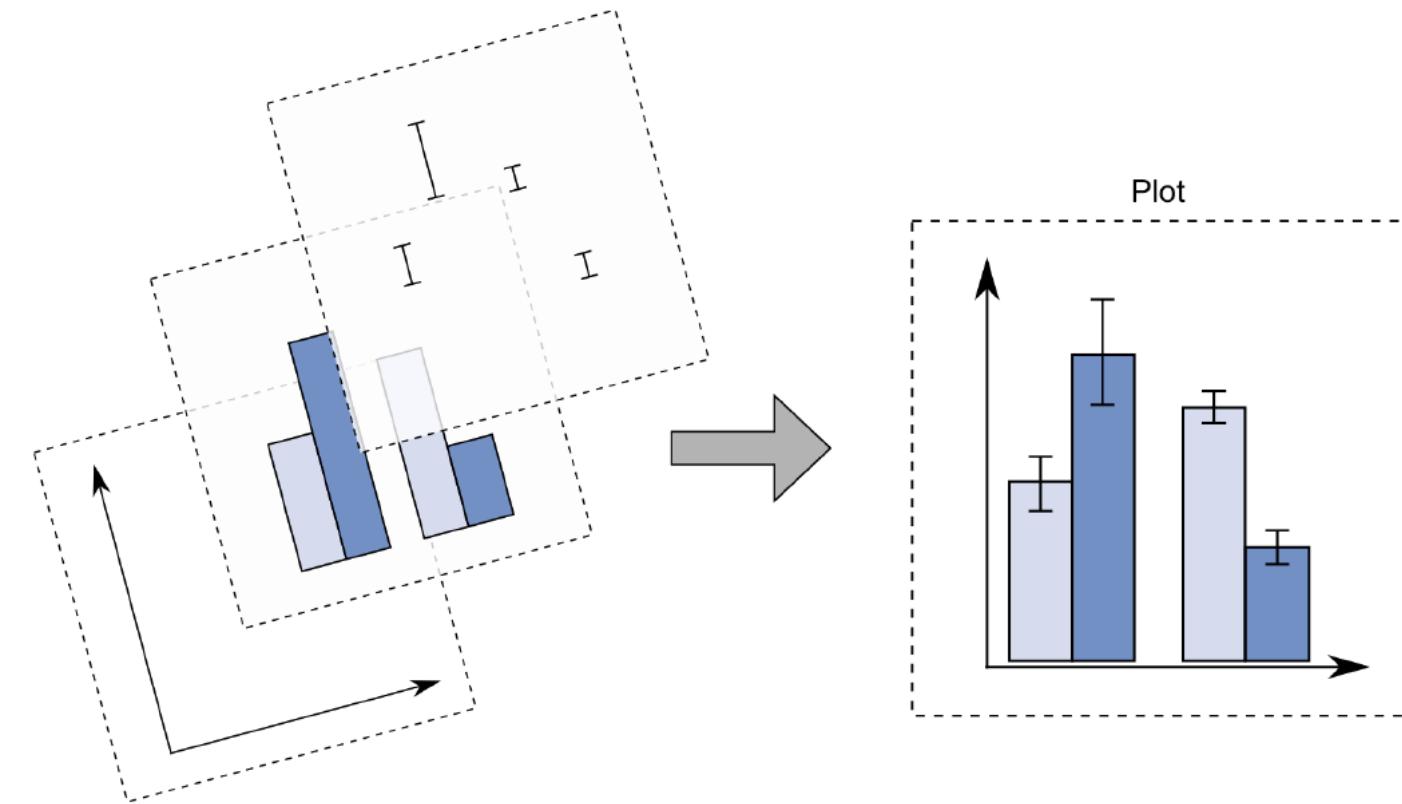
ggplot2

<http://hadley.nz/>

<http://ggplot2.org/>

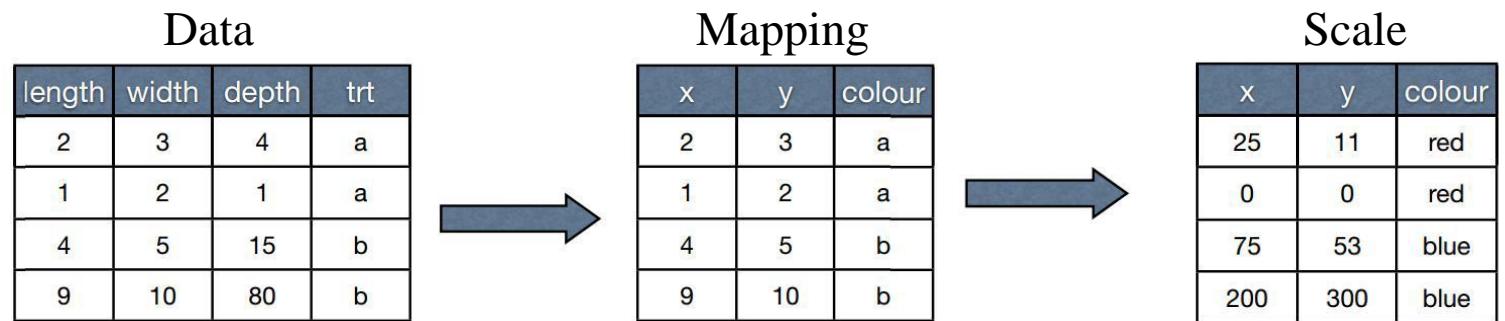


图层 (layers)



ggplot2的基本概念

➤ 数据 (Data) 和映射 (Mapping)



➤ 标度 (Scale)

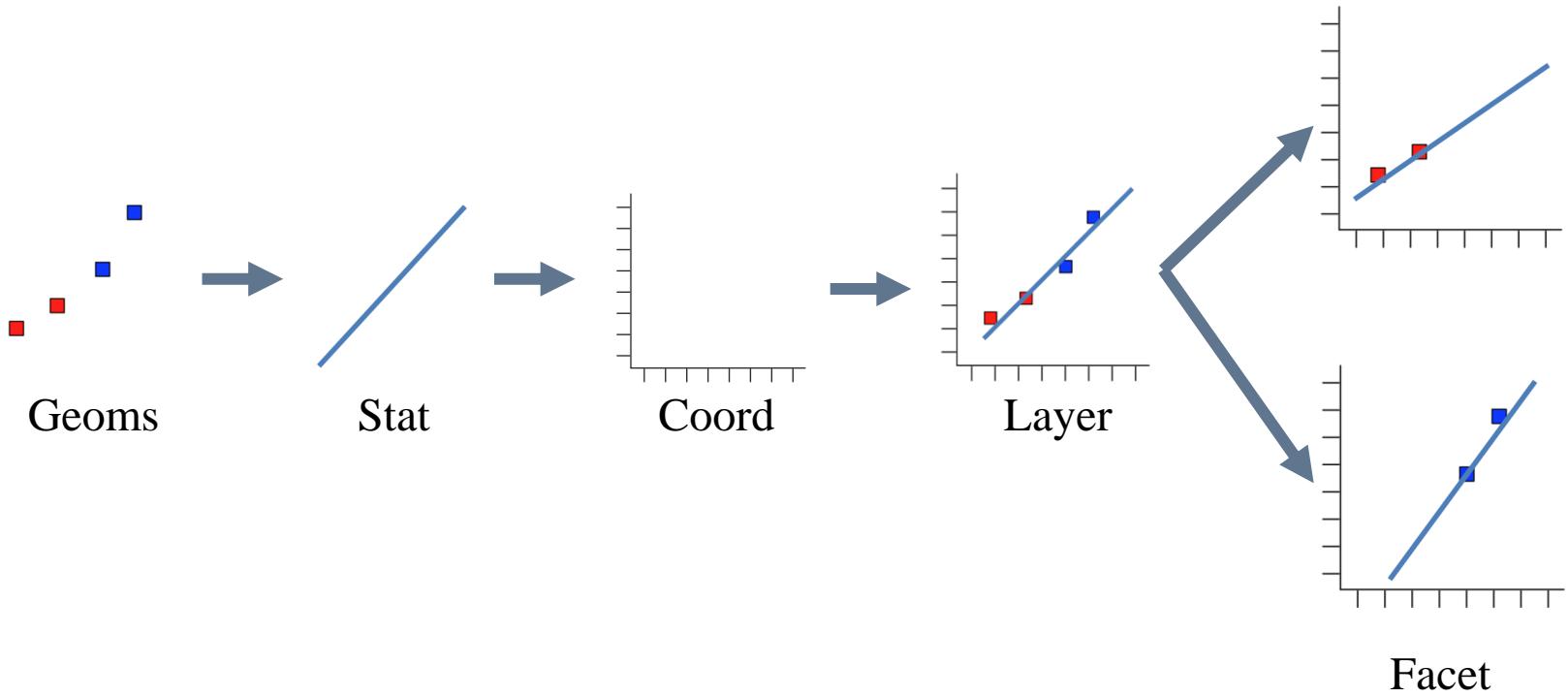
➤ 几何对象 (Geometric)

➤ 统计变换 (Statistics)

➤ 坐标系统 (Coordinate)

➤ 图层 (Layer)

➤ 分面 (Facet)



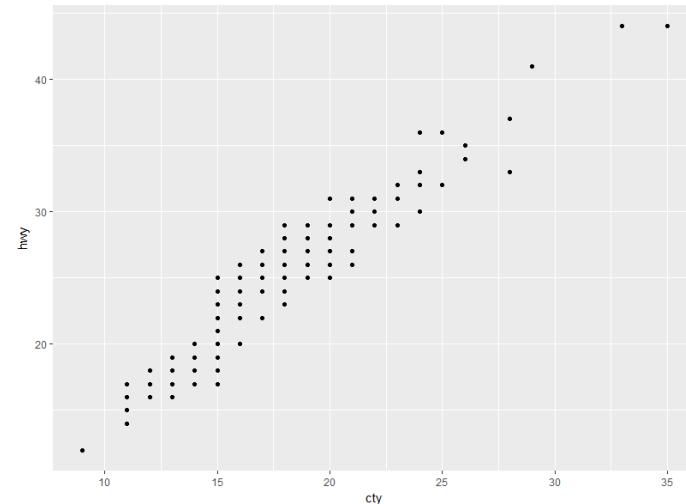
实例数据

```
1 > library(ggplot2)
2 > str(mpg)
3 #> Classes 'tbl_df', 'tbl' and 'data.frame': 234 obs. of 11 variables:
4 #>   $ manufacturer: chr "audi" "audi" "audi" "audi" ...
5 #>   $ model       : chr "a4" "a4" "a4" "a4" ...
6 #>   $ displ        : num 1.8 1.8 2 2 2.8 2.8 3.1 1.8 1.8 2 ...
7 #>   $ year         : int 1999 1999 2008 2008 1999 1999 2008 1999 1999 2008 ...
8 #>   $ cyl          : int 4 4 4 4 6 6 6 4 4 4 ...
9 #>   $ trans        : chr "auto(15)" "manual(m5)" "manual(m6)" "auto(av)" ...
10 #>  $ drv           : chr "f" "f" "f" "f" ...
11 #>  $ cty           : int 18 21 20 21 16 18 18 18 16 20 ...
12 #>  $ hwy           : int 29 29 31 30 26 26 27 26 25 28 ...
13 #>  $ fl            : chr "p" "p" "p" "p" ...
14 #>  $ class         : chr "compact" "compact" "compact" "compact" ...
15
16 > head(mpg)
17 #> # A tibble: 6 x 11
18 #>   manufacturer model  displ  year   cyl     trans   drv   cty   hwy     fl   class
19 #>   <chr> <chr> <dbl> <int> <int>    <chr> <chr> <int> <int> <chr> <chr>
20 1      audi    a4     1.8  1999     4  auto(15)    f     18    29     p compact
21 2      audi    a4     1.8  1999     4  manual(m5)   f     21    29     p compact
22 3      audi    a4     2.0  2008     4  manual(m6)   f     20    31     p compact
23 4      audi    a4     2.0  2008     4  auto(av)     f     21    30     p compact
24 5      audi    a4     2.8  1999     6  auto(15)     f     16    26     p compact
25 6      audi    a4     2.8  1999     6  manual(m5)   f     18    26     p compact
26
```

散点图

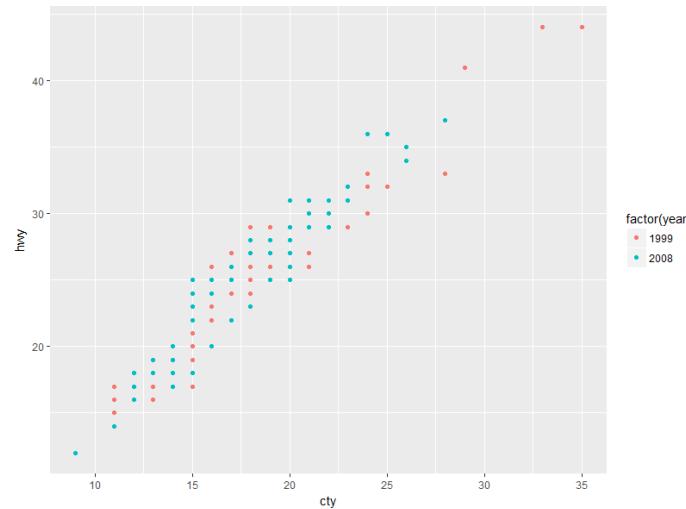
aesthetics

```
1 p <- ggplot(data=mpg, mapping=aes(x=cty, y=hwy))  
2 p + geom_point()  
3
```



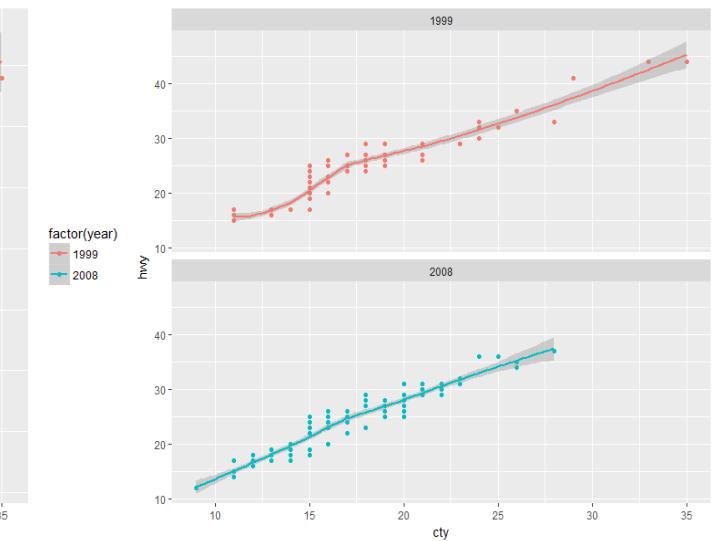
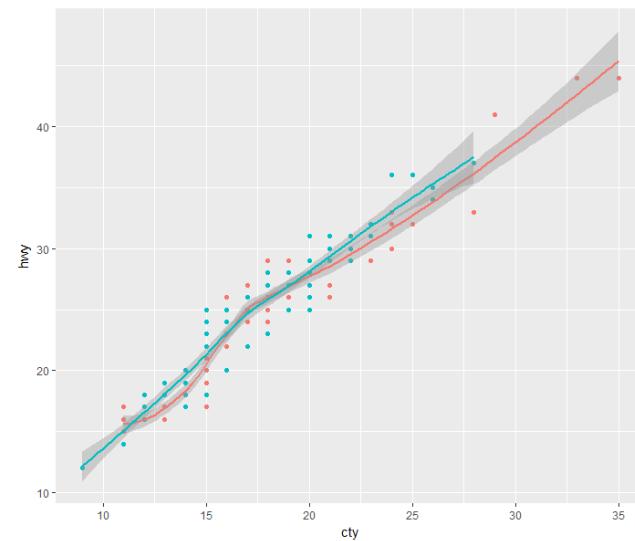
将年份映射到颜色属性 (Mapping):

```
1 p <- ggplot(mpg,aes(x=cty, y=hwy, colour=factor(year)))  
2 p + geom_point()  
3
```



增加平滑曲线 (Statistics):

```
1 p + geom_point() + stat_smooth()  
2 > summary(p + geom_point() + stat_smooth())  
3 geom_smooth: na.rm = FALSE  
4 stat_smooth: method = auto, formula = y ~ x,  
5     se = TRUE, n = 80, fullrange = FALSE,  
6     level = 0.95, na.rm = FALSE,  
7     method.args = list(), span = 0.75  
8 position_identity  
9
```

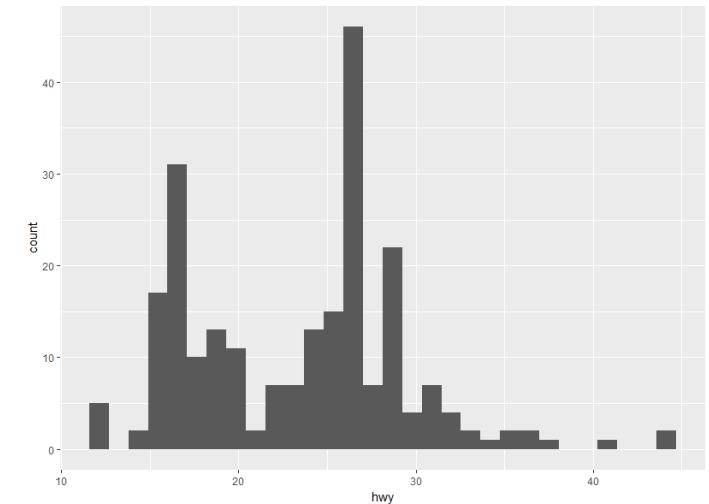


分面 (Facet):

```
1 p + geom_point() + stat_smooth()+facet_wrap(~ year, ncol=1)  
2
```

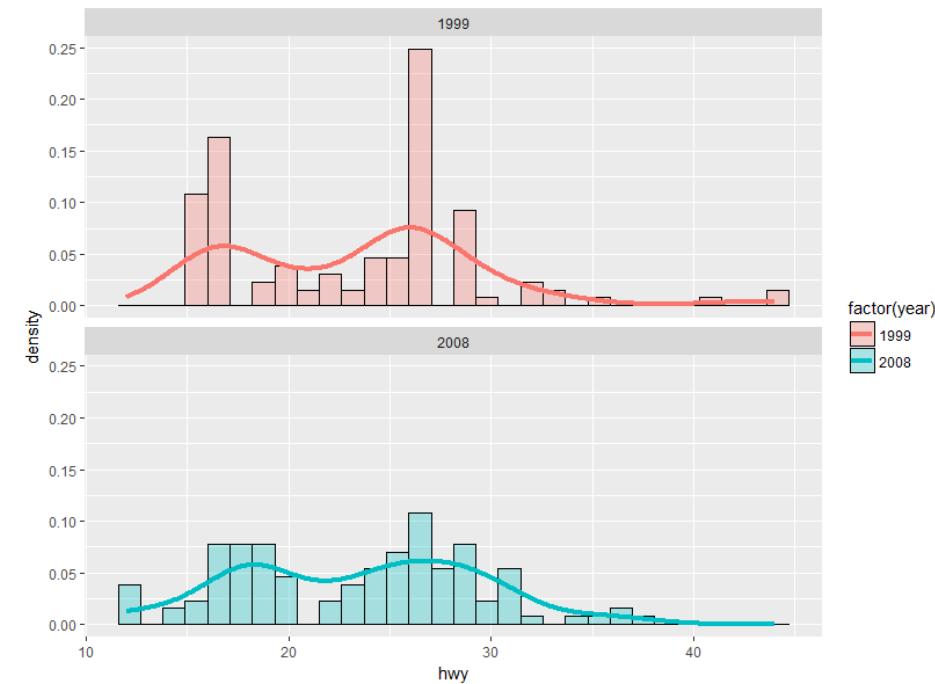
直方图

```
1 p <- ggplot(mpg, aes(x=hwy))  
2 p + geom_histogram()  
3
```



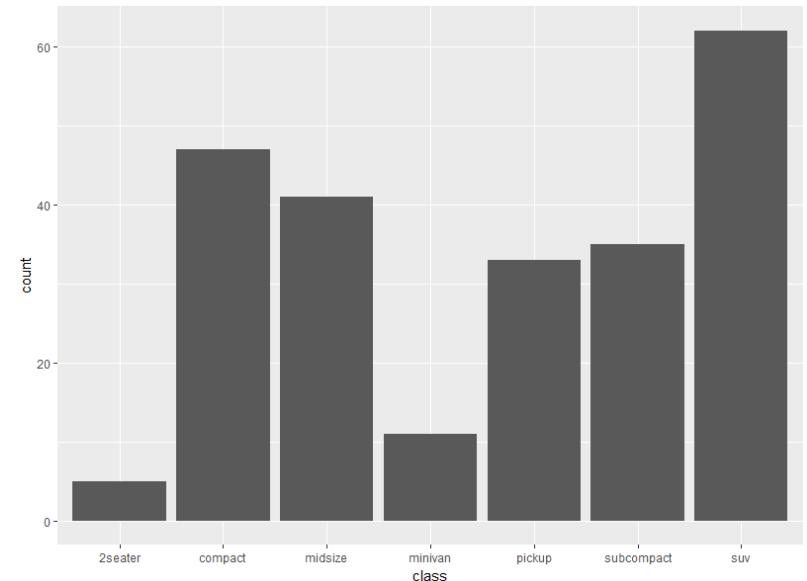
统计变换 + 分面：

```
1 # Histogram with density plot  
2 p + geom_histogram(aes(fill=factor(year),y=..density..), alpha=0.3, colour='black') +  
3   stat_density(geom='line', position='identity', size=1.5, aes(colour=factor(year))) +  
4   facet_wrap(~year, ncol=1)  
5
```



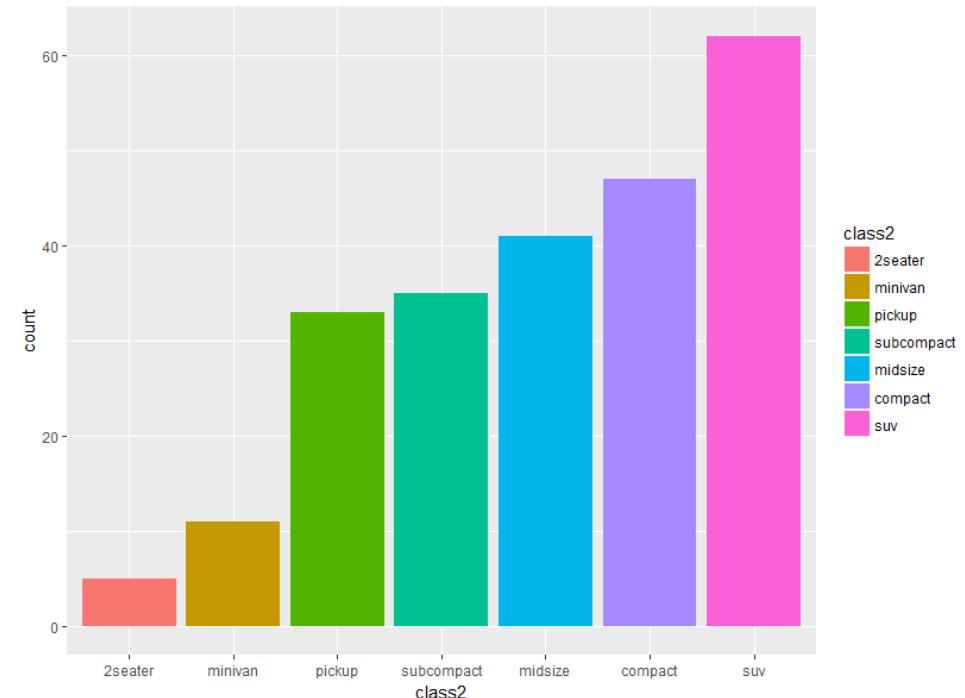
条形图

```
1 p <- ggplot(mpg, aes(x=class))  
2 p + geom_bar()  
3
```



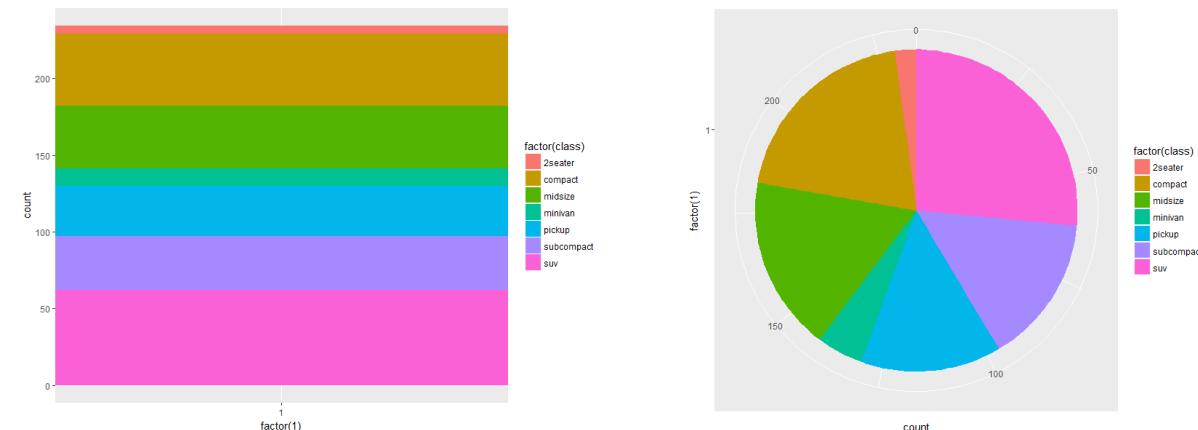
根据计数排序后绘制的条形图：

```
1 class2 <- mpg$class  
2 class2 <- reorder(class2,class2,length)  
3 mpg$class2 <- class2  
4 p <- ggplot(mpg, aes(x=class2))  
5 p + geom_bar(aes(fill=class2))  
6  
7 help(reorder)  
8  
9 Reorder Levels of a Factor  
10 Usage  
11 reorder(x, X, FUN = mean, ...,  
12   order = is.ordered(x))  
13  
14 class(class2)  
15 [1] "factor"  
16
```



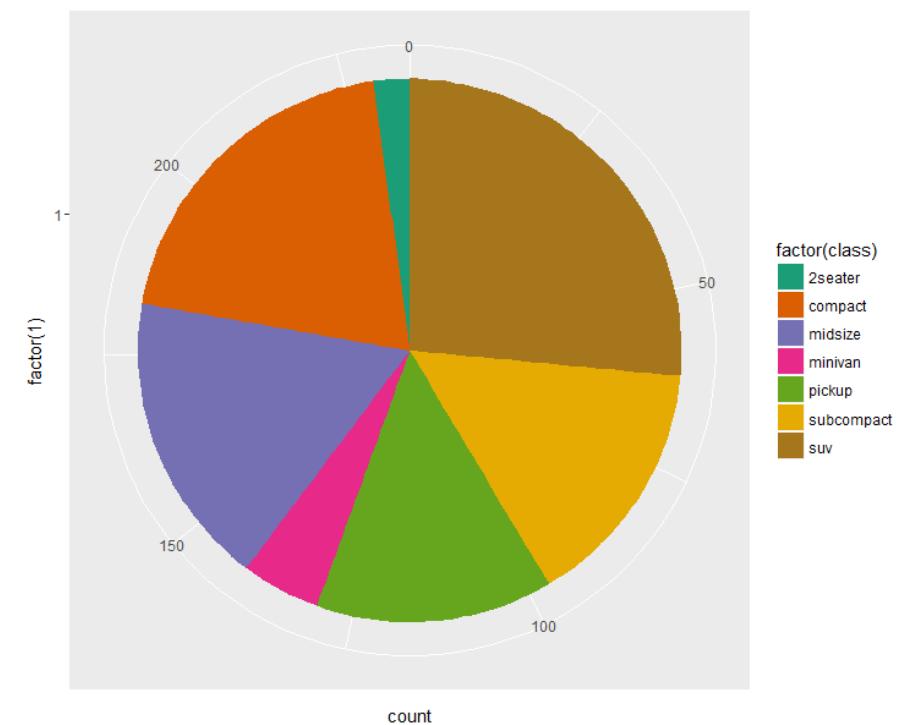
饼图

```
1 p <- ggplot(mpg, aes(x = factor(1), fill = factor(class))) +  
2   geom_bar(width = 1)  
3 p  
4  
5 p + coord_polar(theta = "y")  
6  
7
```



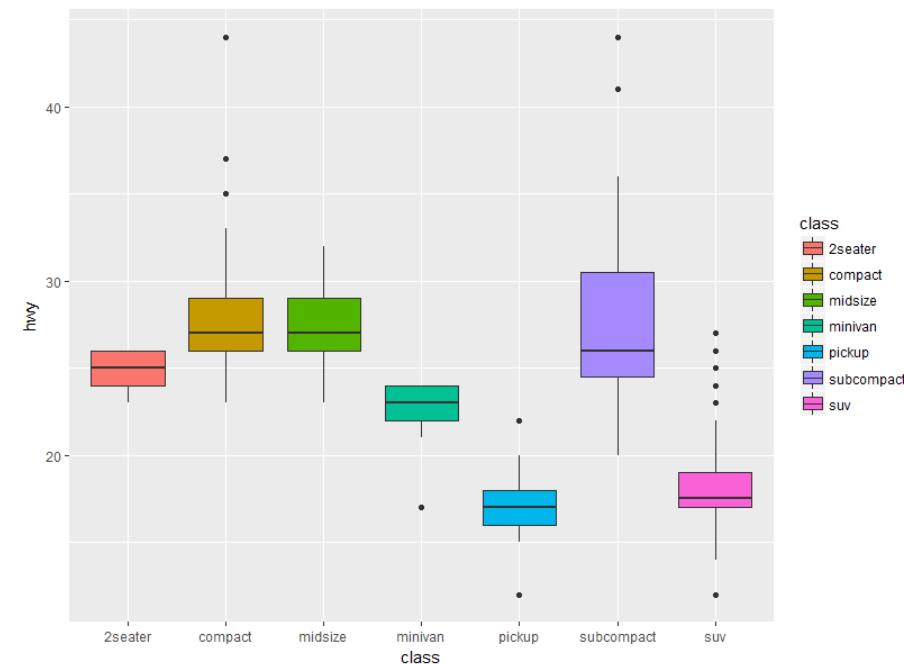
改变填充颜色:

```
1 p + coord_polar(theta = "y") +  
2   scale_fill_brewer(palette="Dark2")  
3
```



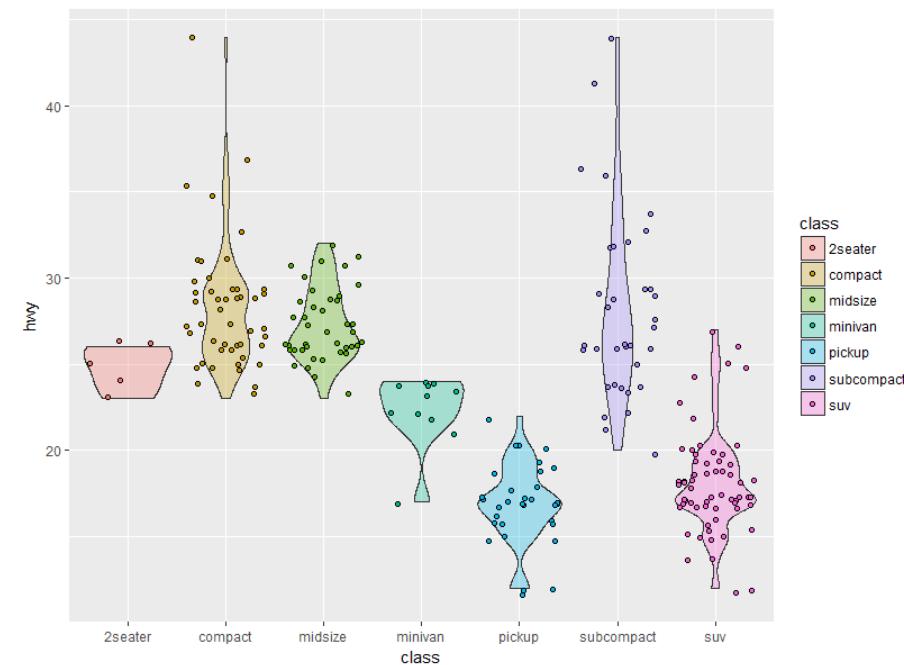
箱线图

```
1 p <- ggplot(mpg, aes(class,hwy,fill=class))  
2 p + geom_boxplot()  
3
```



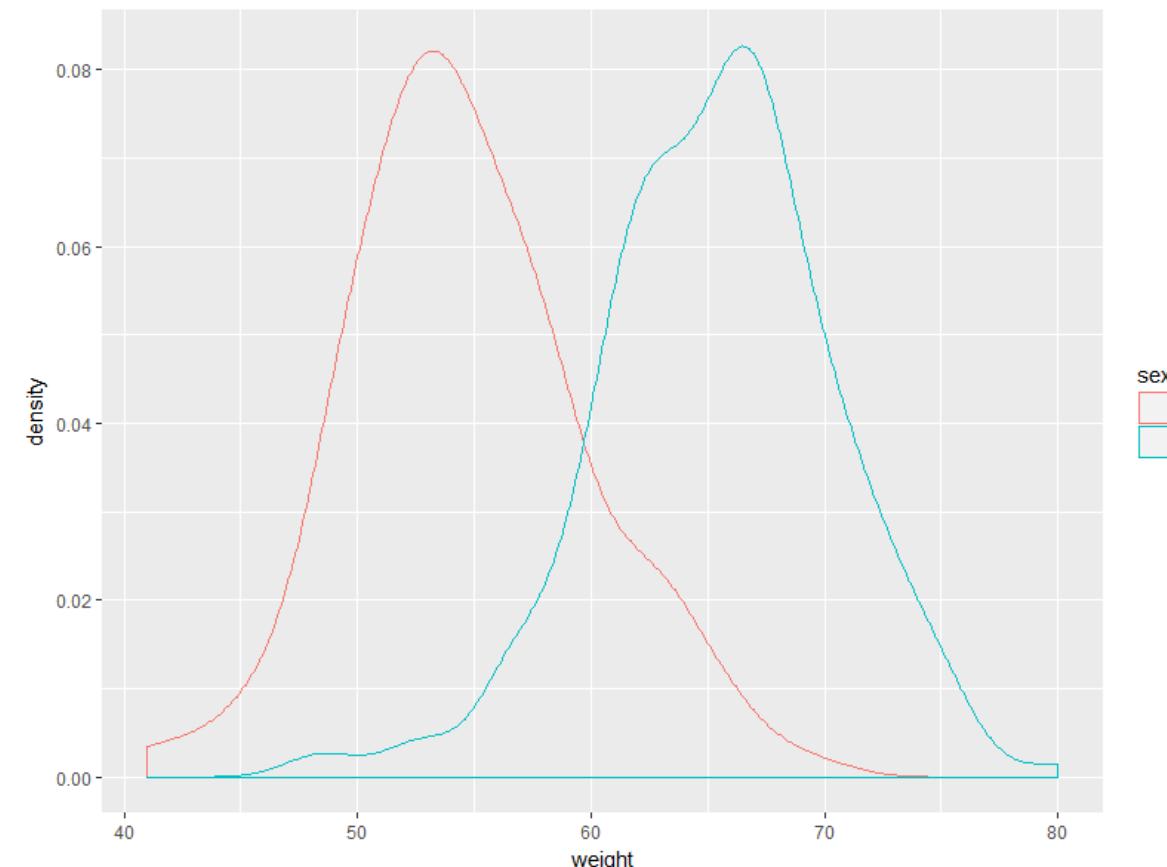
小提琴图

```
1 p + geom_violin(alpha=0.3,width=0.9)+  
2   geom_jitter(shape=21)  
3
```



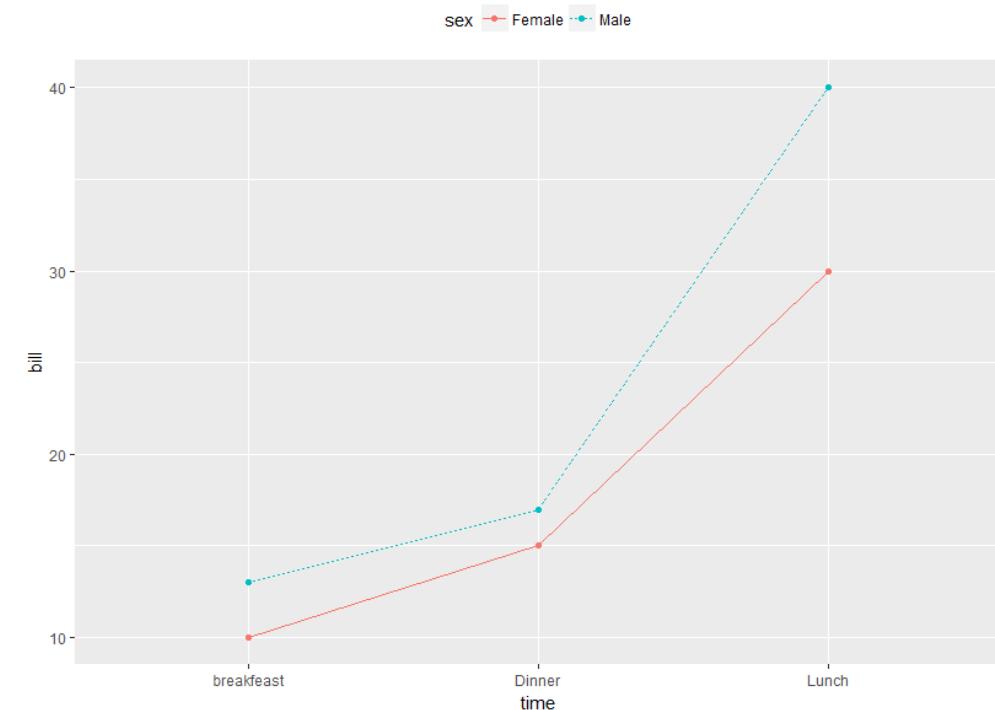
密度图

```
1 set.seed(1234)
2 df <- data.frame(
3   sex=factor(rep(c("F", "M"), each=200)),
4   weight=round(c(rnorm(200, mean=55, sd=5),
5                 rnorm(200, mean=65, sd=5)))
6 )
7 head(df)
8   sex weight
9 1 F    49
10 2 F    56
11 3 F    60
12 4 F    43
13 5 F    57
14 6 F    58
15 p <- ggplot(df, aes(x=weight, color=sex)) +
16   geom_density()
17 p
18
```



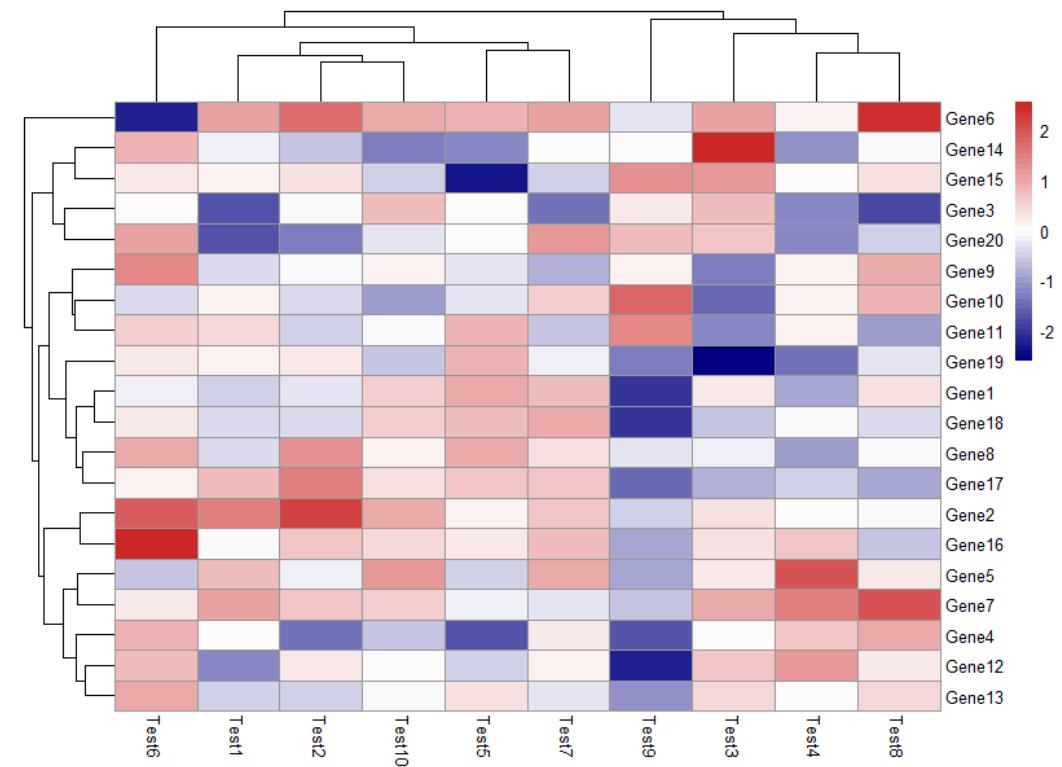
线图

```
1 df2 <- data.frame(sex = rep(c("Female", "Male"), each=3),  
2                     time=c("breakfeast", "Lunch", "Dinner"),  
3                     bill=c(10, 30, 15, 13, 40, 17) )  
4 head(df2)  
5   sex      time bill  
6 1 Female breakfeast  10  
7 2 Female      Lunch  30  
8 3 Female     Dinner  15  
9 4 Male   breakfeast  13  
10 5 Male      Lunch  40  
11 6 Male     Dinner  17  
12 P <- ggplot(df2, aes(x=time, y=bill, group=sex)) +  
13   geom_line(aes(linetype=sex, color=sex))+  
14   geom_point(aes(color=sex))+  
15   theme(legend.position="top")  
16 P  
17
```



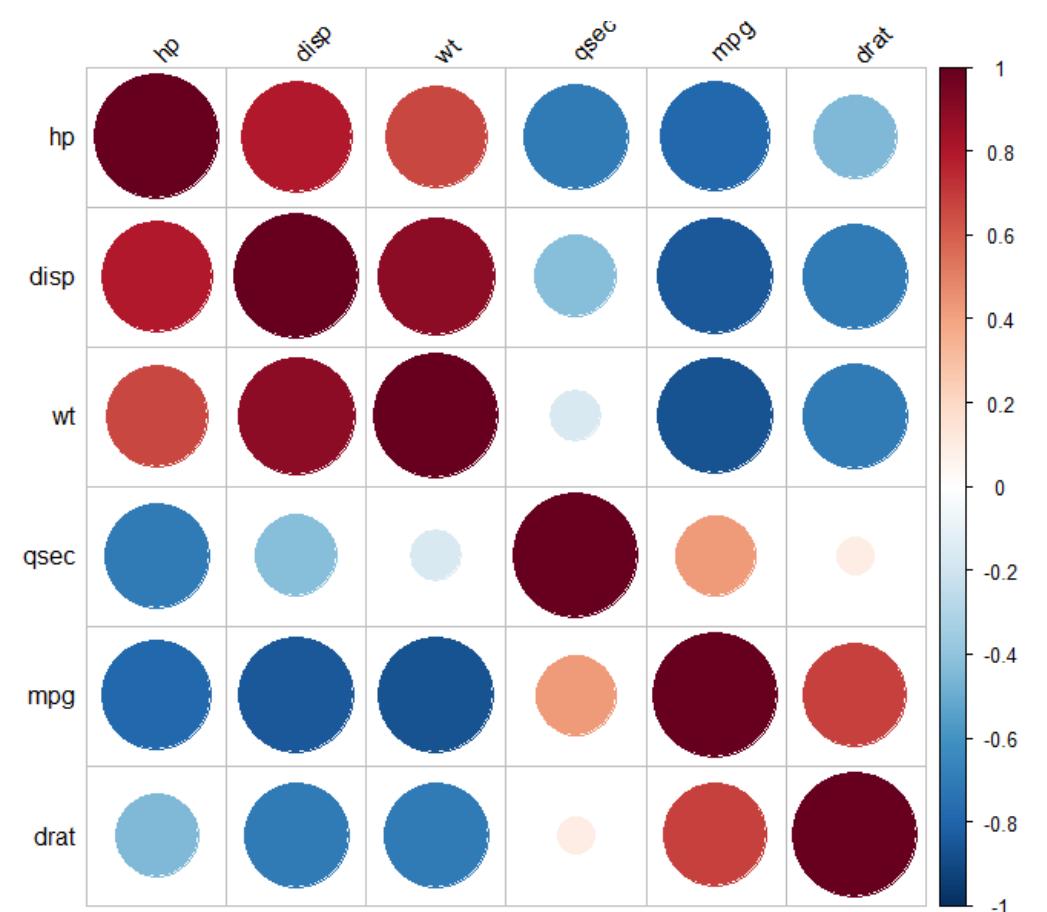
热图

```
1 library(pheatmap)
2 test = matrix(rnorm(200), 20, 10)
3 help(rnorm)
4 -----
5 random generation for the normal distribution with
6 mean equal to mean and standard deviation equal to sd.
7 rnorm(n, mean = 0, sd = 1)
8 -----
9 colnames(test) = paste("Test", 1:10, sep = "")
10 rownames(test) = paste("Gene", 1:20, sep = "")
11 pheatmap(test, color = colorRampPalette(c("navy", "white", "firebrick3"))(50))
12
```



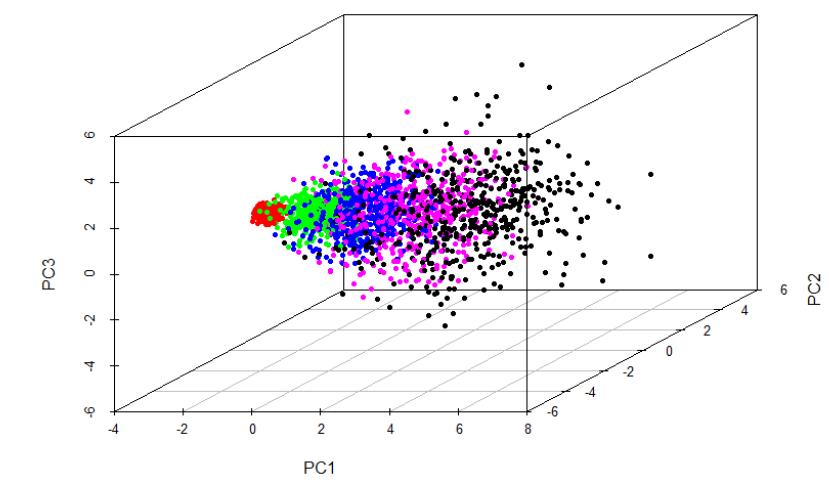
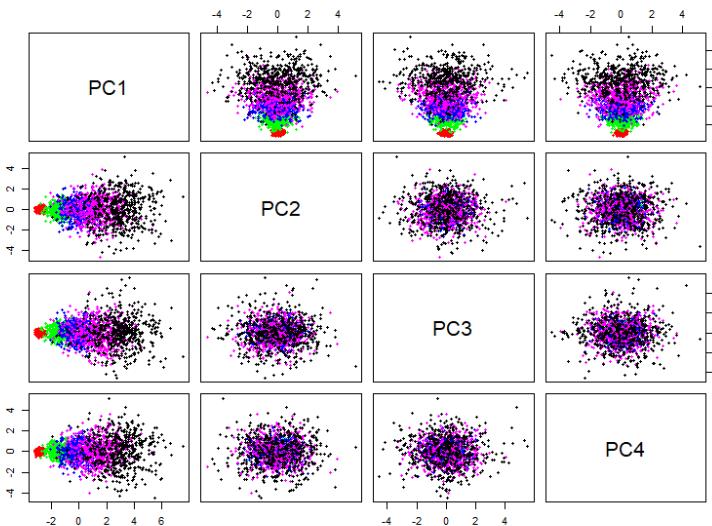
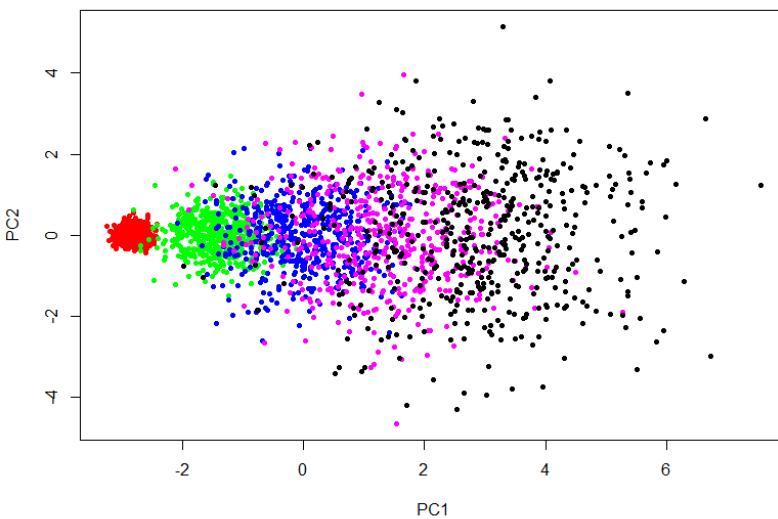
相关性图

```
1 install.packages("corrplot")
2 library(corrplot)
3 mydata <- select(mtcars, hp, disp, wt, qsec, mpg, drat)
4 source("http://www.sthda.com/upload/rquery_cormat.r")
5 rquery.cormat<-function(x, type=c('lower', 'upper', 'full', 'flatten'), [...]
6 {...}
7 }
8 rquery.cormat(mydata, type="full")
91 }
92 rquery.cormat(mydata, type="full")
93 }
```



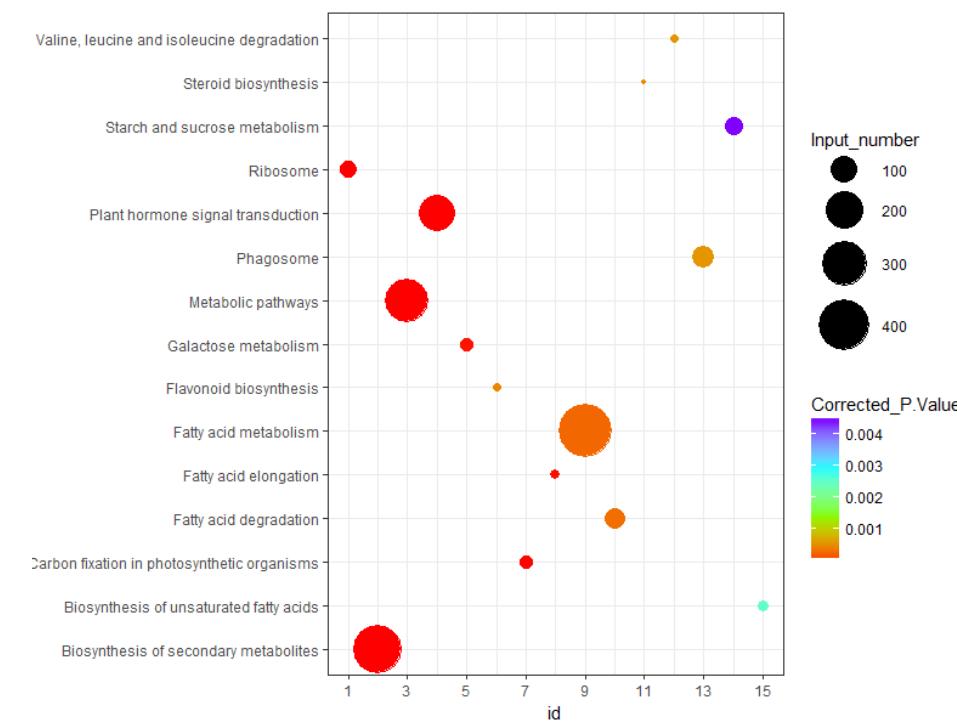
主成份分析 (PCA)

```
1 z1 <- rnorm(10000, mean=1, sd=1)
2 z2 <- rnorm(10000, mean=3, sd=3)
3 z3 <- rnorm(10000, mean=5, sd=5)
4 z4 <- rnorm(10000, mean=7, sd=7)
5 z5 <- rnorm(10000, mean=9, sd=9)
6 mydata <- matrix(c(z1, z2, z3, z4, z5), 2500, 20, byrow=T, dimnames=list(paste("R", 1:2500, sep=""), paste("C", 1:20, sep="")))
7 pca <- prcomp(mydata, scale=T)
8 summary(pca)$importance[, 1:6]
9 mycolors <- c("red", "green", "blue", "magenta", "black")
10
11 plot(pca$x[,1:2], pch=20, col=mycolors[sort(rep(1:5, 500))])
12
13 pairs(pca$x[,1:4], pch=20, col=mycolors[sort(rep(1:5, 500))])
14
15 library(scatterplot3d)
16 scatterplot3d(pca$x[,1:3], pch=20, color=mycolors[sort(rep(1:5, 500))])
17
```



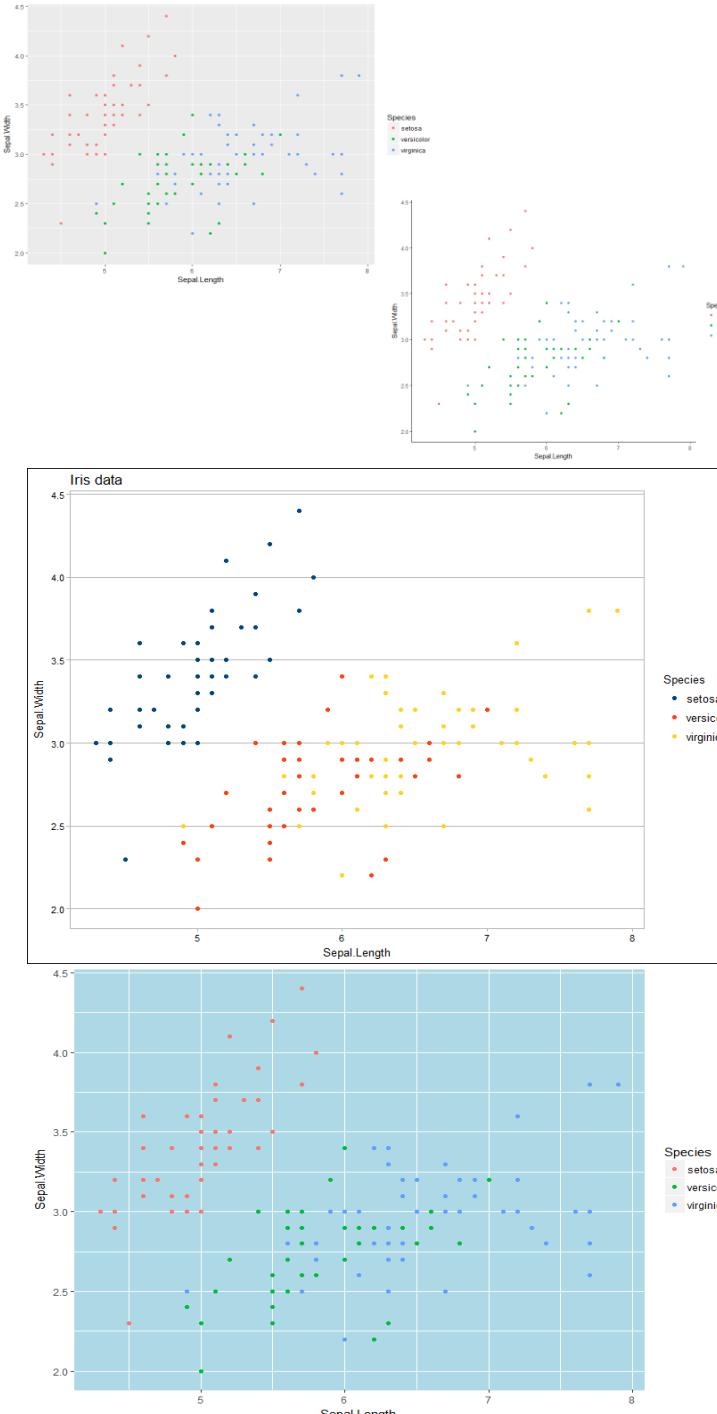
气泡图 (Bubbles)

```
1  require(ggplot2)
2  df<- read.csv("Bubbles.csv")
3  > df
4
5      Term  Input_number      P.Value id
6  1     Ribosome 0.0000000000000000797516 1
7  2 Biosynthesis of secondary metabolites 0.000000000000000149654 2
8  3           Metabolic pathways 0.000000000000394082733 3
9  4 Plant hormone signal transduction 0.00000001163614774940 4
10 5           Galactose metabolism 0.00001489712860000000 5
11 6 Flavonoid biosynthesis 0.00048351702379700000 6
12 7 Carbon fixation in photosynthetic organisms 0.00000197712300000000 7
13 8           Fatty acid elongation 0.00000852890000000000 8
14 9           Fatty acid metabolism 0.00029620699999999999 9
15 10          Fatty acid degradation 0.00033999500000000000 10
16 11          Steroid biosynthesis 0.00051699999999999999 11
17 12 Valine, leucine and isoleucine degradation 0.0005875519999999997 12
18 13           Phagosome 0.0005879299999999996 13
19 14 Starch and sucrose metabolism 0.004420000000000032 14
20 15 Biosynthesis of unsaturated fatty acids 0.00253350000000000016 15
21 > ggplot(df, aes(x = id,y=Term,label = Term)) +
22   geom_point(aes(size = Input_number, colour = P.Value)) +
23   #geom_text(hjust = 1, size = 2) +
24   scale_size(range = c(1,15)) +
25   scale_x_continuous(breaks = seq(1, 15, 2)) +
26   scale_colour_gradientn(colours=rainbow(4)) +
27   theme_bw()
28
```



美化 (themes and background)

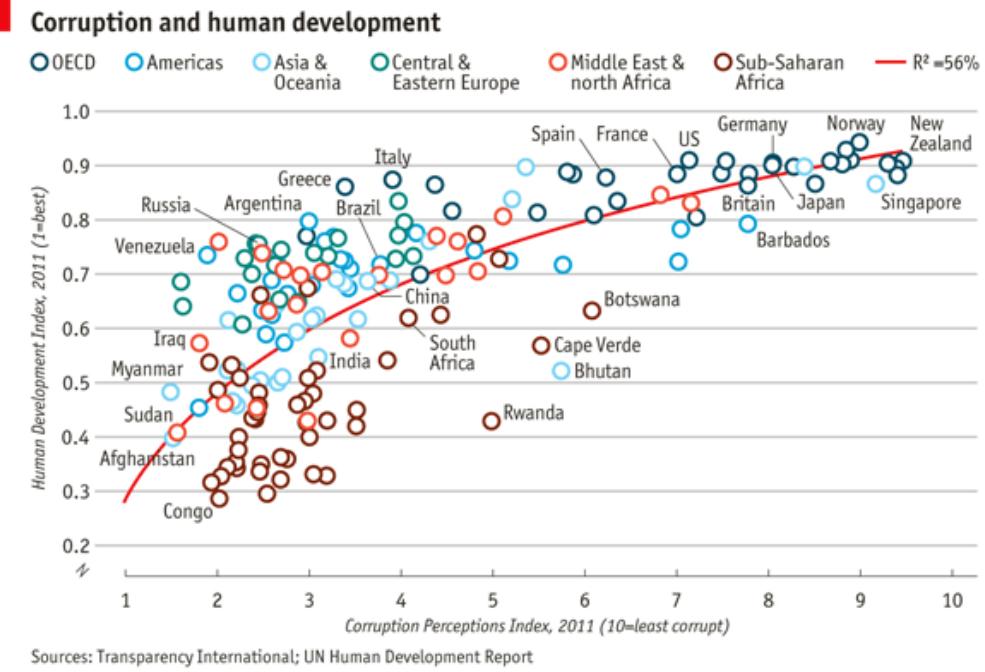
```
1 # ggplot2自带主题
2 p <- ggplot(iris, aes(Sepal.Length, Sepal.Width, colour = Species))+  
  geom_point()
3 p
4 p + theme_classic()
5
6 # 主题包
7 install.packages("ggthemes") # Install
8 library(ggthemes) # Load
9
10 p + theme_calc() + scale_colour_calc() +
11   ggtitle("Iris data")
12
13 # 定制主题
14 p + theme(
15   panel.background = element_rect(fill = "lightblue",
16                                     colour = "lightblue",
17                                     size = 0.5, linetype = "solid"),
18   panel.grid.major = element_line(size = 0.5, linetype = 'solid',
19                                     colour = "white"),
20   panel.grid.minor = element_line(size = 0.25, linetype = 'solid',
21                                     colour = "white"))
22 )
23
```



RNA-Seq (DESeq2)

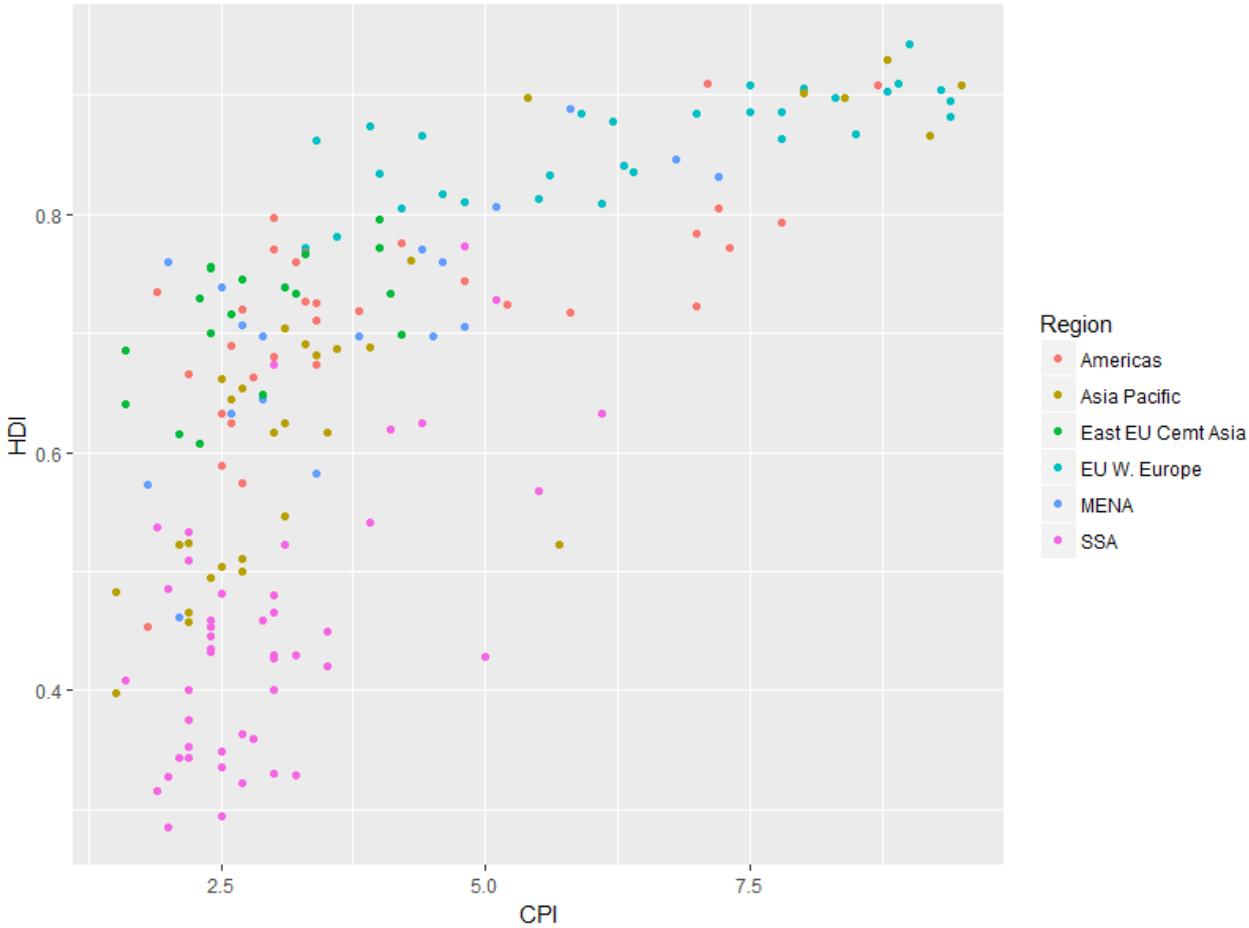
```
1 library(DESeq2);library(limma);library(pasilla);data(pasillaGenes);exprSet=counts(pasillaGenes)
2 # 表达矩阵
3 head(exprSet)
4   treated1fb treated2fb treated3fb untreated1fb untreated2fb untreated3fb untreated4fb
5 FBgn0000003      0       0       1       0       0       0       0
6 FBgn0000008     78      46      43      47      89      53      27
7 FBgn0000014      2       0       0       0       0       1       0
8 FBgn0000015      1       0       1       0       1       1       2
9 FBgn0000017    3187     1672     1859     2445     4615    2063    1711
10 FBgn0000018     369     150     176     288     383     135     174
11 # 构建dds对象
12 colData <- data.frame(row.names=colnames(exprSet), group_list=group_list)
13 > colData
14   group_list
15 treated1fb   treated
16 treated2fb   treated
17 treated3fb   treated
18 untreated1fb  untreated
19 untreated2fb  untreated
20 untreated3fb  untreated
21 untreated4fb  untreated
22 dds <- DESeqDataSetFromMatrix(countData = exprSet,colData = colData,design = ~ group_list)
23 # normalization
24 dds2 <- DESeq(dds)
25 # 提取差异分析结果
26 resultsNames(dds2)
27 res <- results(dds2, contrast=c("group_list","treated","untreated"))
28 resOrdered <- res[order(res$padj),]
29 resOrdered=as.data.frame(resOrdered)
30 head(resOrdered)
31   baseMean log2FoldChange      lfcSE      stat      pvalue      padj
32 FBgn0039155  453.2753 -3.714214 0.1600580 -23.20543 4.013291e-119 3.089431e-115
33 FBgn0029167 2165.0445 -2.082793 0.1035963 -20.10491 6.684454e-90 2.572846e-86
34 FBgn0035085 366.8279 -2.227243 0.1369744 -16.26028 1.888618e-59 4.846194e-56
35 FBgn0029896 257.9027 -2.206780 0.1586969 -13.90563 5.854593e-44 1.126716e-40
36 FBgn0034736 118.4074 -2.565002 0.1847628 -13.88268 8.067448e-44 1.242064e-40
37 FBgn0040091 610.6035 -1.430433 0.1201539 -11.90501 1.114552e-32 1.429970e-29
38
```

复杂图形修改



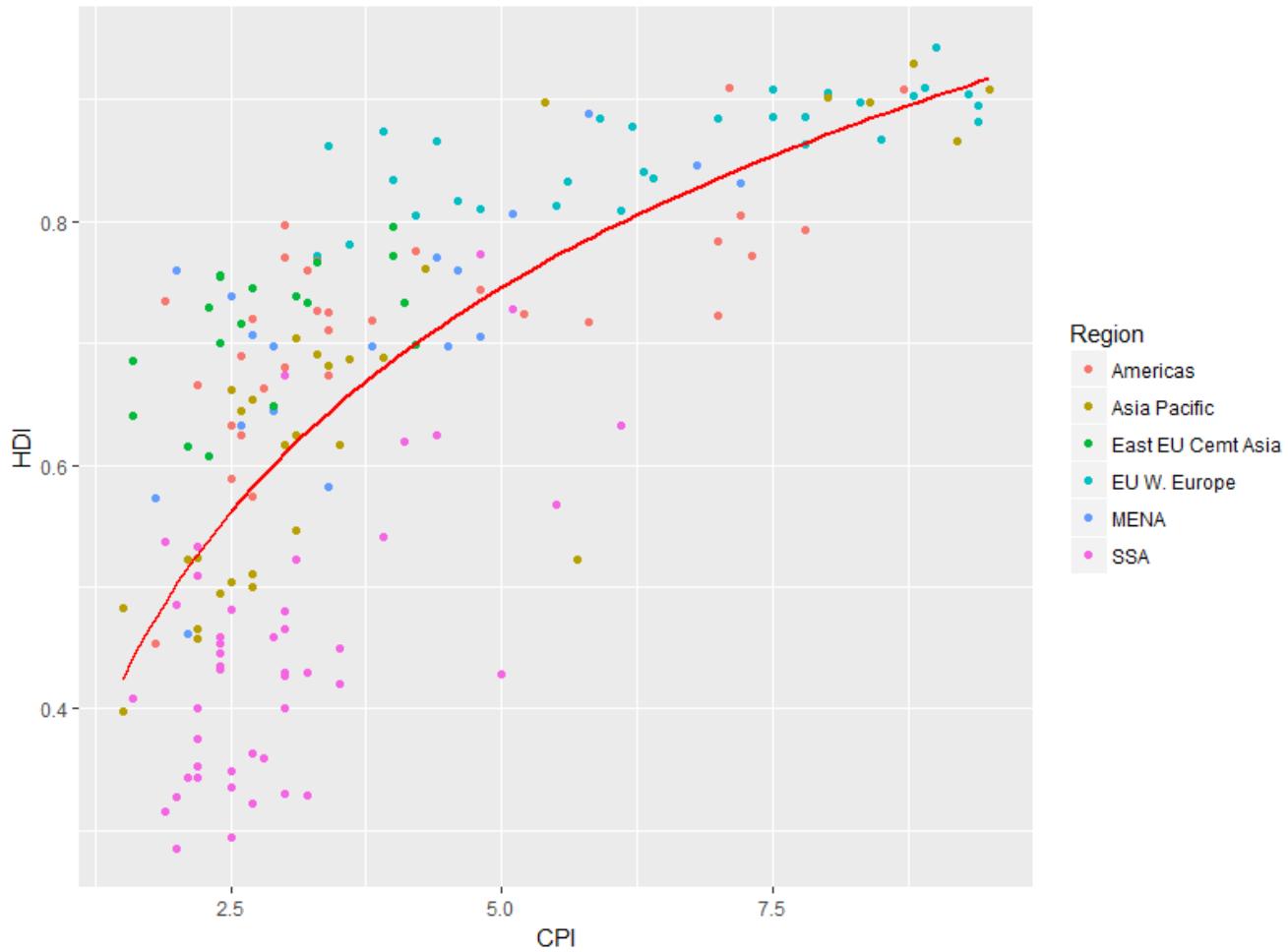
1. 基本图形

```
1 library(ggplot2)
2 dat <- read.csv("EconomistData.csv")
3 # Basic plot
4 pc1 <- ggplot(dat,aes(x = CPI, y = HDI, color = Region))+
5   geom_point()
6 pc1
7
```



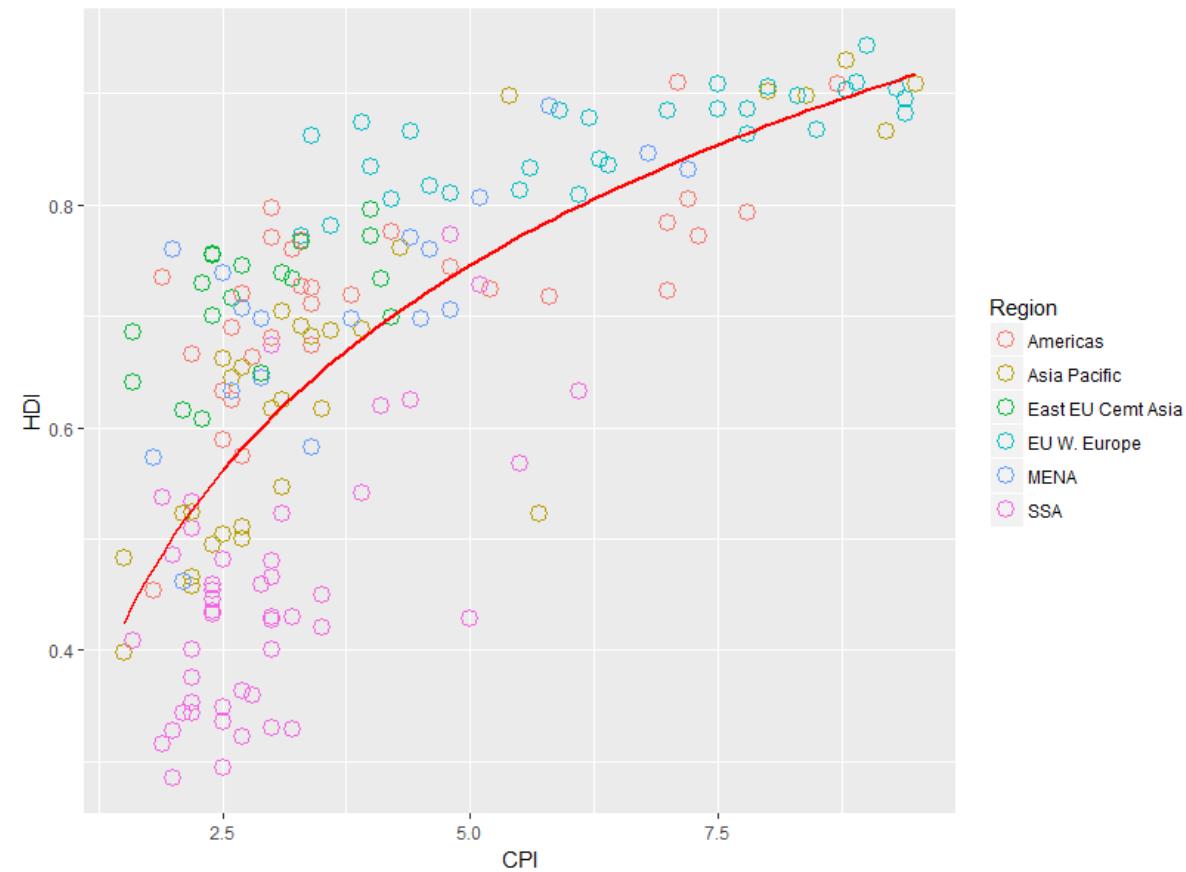
2. 添加趋势线

```
1 # Trend line
2 pc2 <- pc1 +
3   geom_smooth(aes(group = 1),
4     method = "lm",
5     formula = y ~ log(x),
6     se = FALSE,
7     color = "red")
8
9 pc2
```



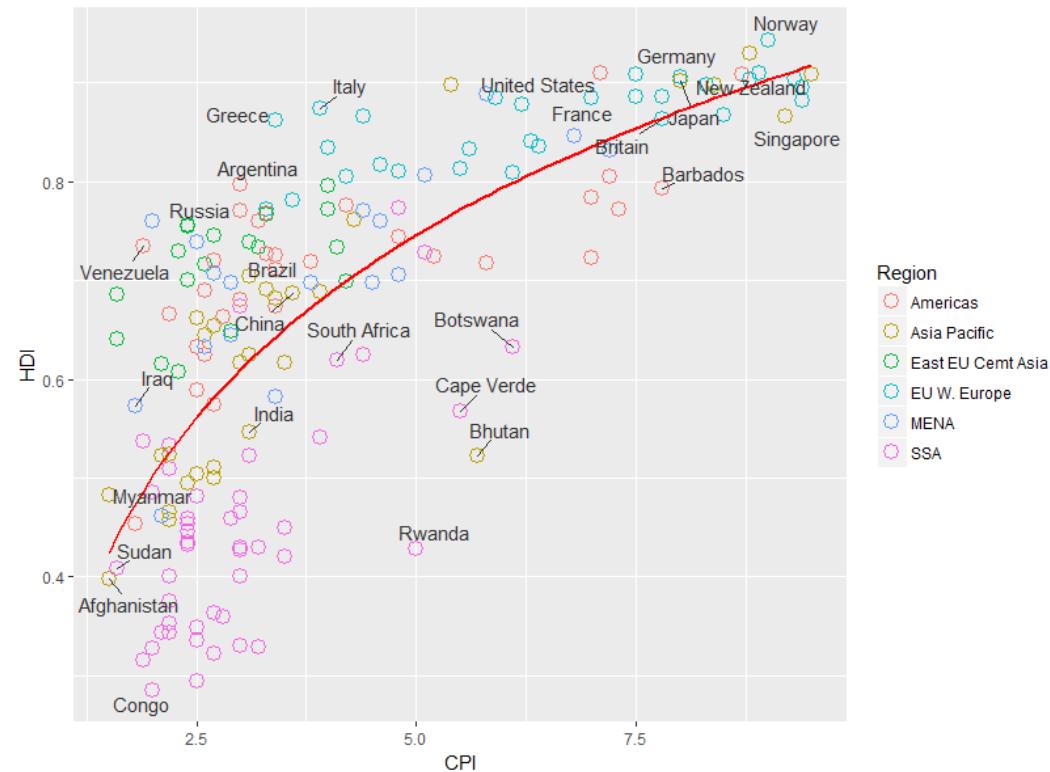
3. 圆圈

```
1  
2  
3 # Open points  
4 pc3 <- ggplot(dat,aes(x = CPI, y = HDI, color = Region))+  
  geom_point(shape = 1, size = 4) +  
  geom_smooth(aes(group = 1),  
              method = "lm",  
              formula = y ~ log(x),  
              se = FALSE,  
              color = "red")  
11 pc3  
12
```



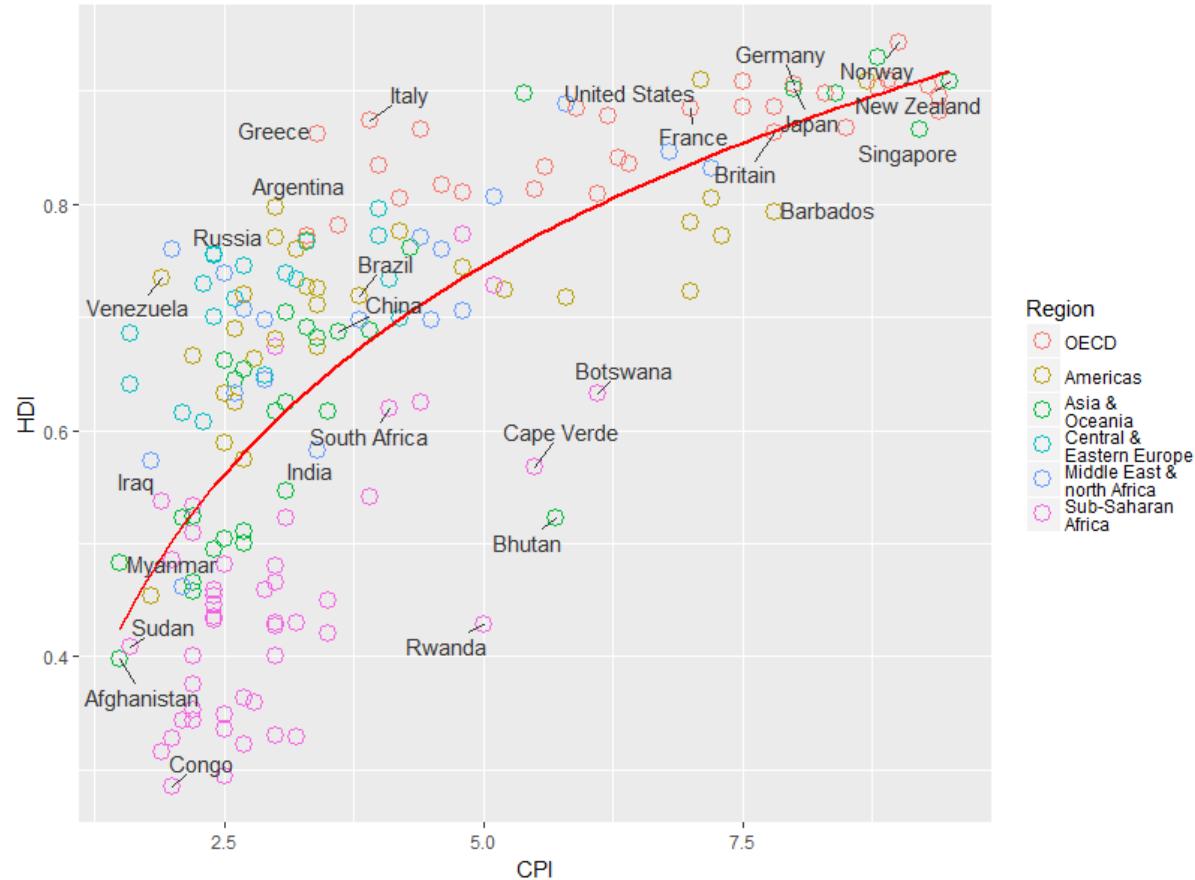
4. 标注想要的点

```
1 pointsToLabel <- c("Russia", "Venezuela", "Iraq", "Myanmar", "Sudan",
2 "Afghanistan", "Congo", "Greece", "Argentina", "Brazil",
3 "India", "Italy", "China", "South Africa", "Spain",
4 "Botswana", "Cape Verde", "Bhutan", "Rwanda",
5 "France", "United States", "Germany", "Britain", "Barbados",
6 "Norway", "Japan",
7 "New Zealand", "Singapore")
8 library("ggrepel")
9 pc4 <- pc3 + geom_text_repel(aes(label = Country),
10 color = "gray20",
11 data = subset(dat, Country %in% pointsToLabel),
12 force = 10)
13 pc4
14
```



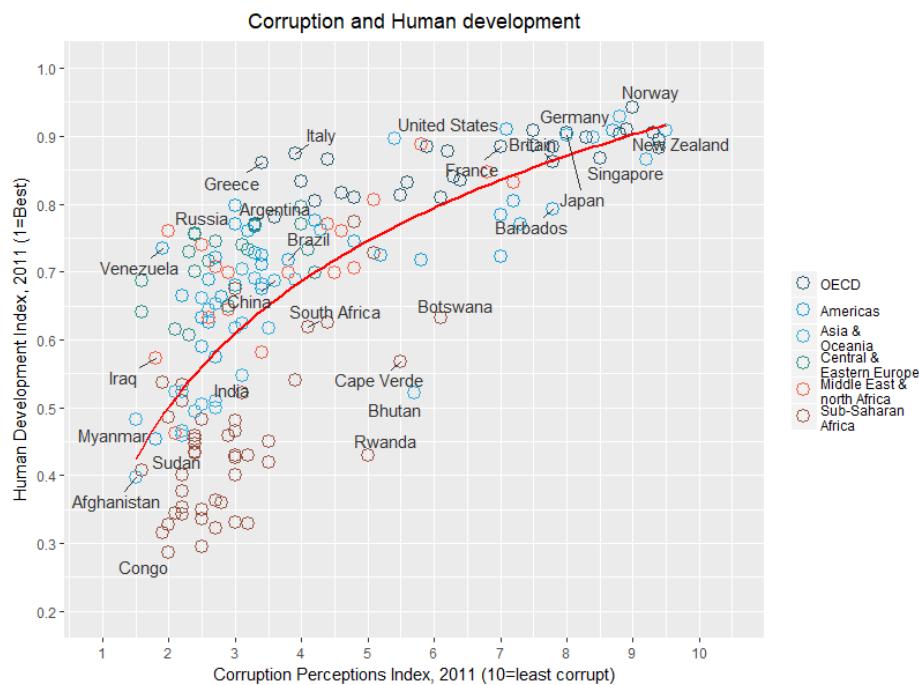
5. 修改图例值和顺序

```
1 dat$Region <- factor(dat$Region,  
2   levels = c("EU W. Europe",  
3     "Americas",  
4     "Asia Pacific",  
5     "East EU Cemt Asia",  
6     "MENA",  
7     "SSA"),  
8   labels = c("OECD",  
9     "Americas",  
10    "Asia &\nOceania",  
11    "Central &\nEastern Europe",  
12    "Middle East &\nnorth Africa",  
13    "Sub-Saharan\nAfrica"))  
14 pc4$data <- dat  
15 pc4  
16
```



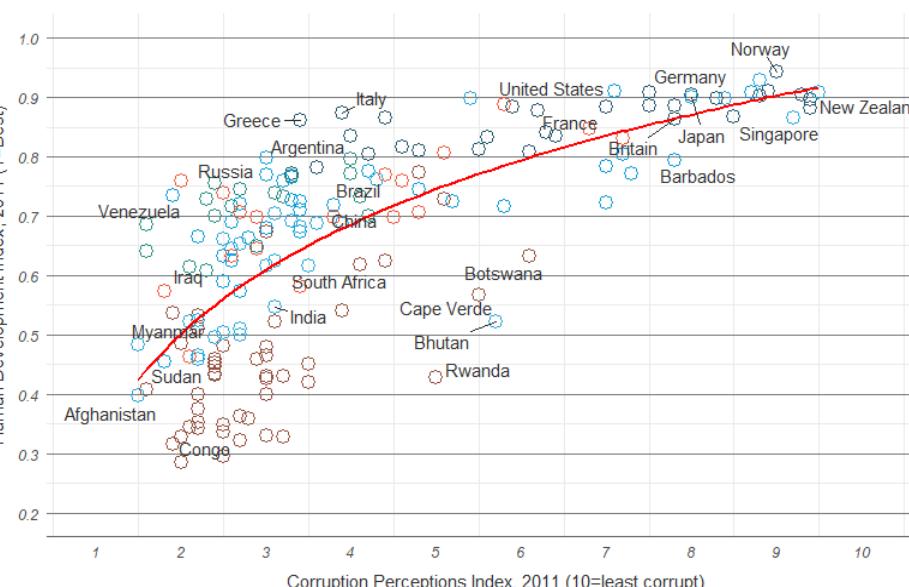
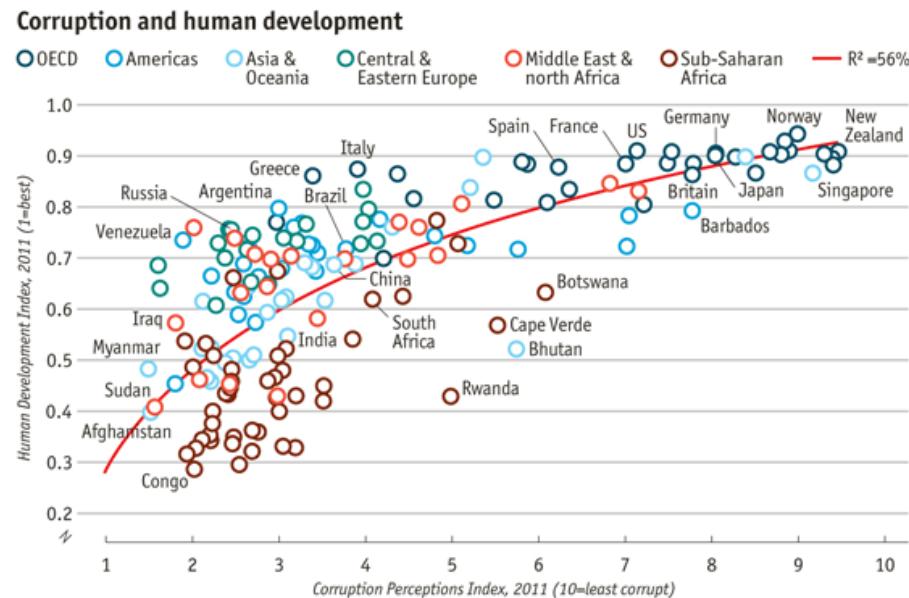
6. 利用scale来修改x, y轴, 颜色和标出title

```
1 library(grid)
2 pc5 <- pc4 +
3   scale_x_continuous(name = "Corruption Perceptions Index, 2011 (10=least corrupt)",
4     limits = c(.9, 10.5),
5     breaks = 1:10) +
6   scale_y_continuous(name = "Human Development Index, 2011 (1=Best)",
7     limits = c(0.2, 1.0),
8     breaks = seq(0.2, 1.0, by = 0.1)) +
9   scale_color_manual(name = "",
10     values = c("#24576D",
11                 "#099DD7",
12                 "#28AADC",
13                 "#248E84",
14                 "#F2583F",
15                 "#96503F")) +
16   ggtitle("Corruption and Human development") +
17   theme(plot.title = element_text(hjust = 0.5))
18 pc5
19
```



7. 微调主题

```
1 library(grid) # for the 'unit' function
2 pc6 <- pc5 +
3   theme_minimal() + # start with a minimal theme and add what we need
4   theme(text = element_text(color = "gray20"),
5         legend.position = c("top"), # position the legend in the upper left
6         legend.direction = "horizontal",
7         legend.justification = 0.1, # anchor point for legend.position.
8         legend.text = element_text(size = 11, color = "gray10"),
9         axis.text = element_text(face = "italic"),
10        axis.title.x = element_text(vjust = -1), # move title away from axis
11        axis.title.y = element_text(vjust = 2), # move away for axis
12        axis.ticks.y = element_blank(), # element_blank() is how we remove elements
13        axis.line = element_line(color = "gray40", size = 0.5),
14        axis.line.y = element_blank(),
15        panel.grid.major = element_line(color = "gray50", size = 0.5),
16        panel.grid.major.x = element_blank()
17      )
18
19 pc6
```



THANKS FOR YOUR WATCHING



<http://tiramisutes.github.io/>

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