

# plot01

## 01. 변수의 특징을 찾기 위해서 사용하는 그래프

(1) 하나의 변수에 대한 것.

히스토그램(hist) : 연속형 변수 표시

Barplot : 명목형 변수 그래프 표시

Boxplot : 데이터의 분포를 본다.

(2) 산포도, 산점도(두변수에 관한 것)

## 02. plot(base), ggplot2 패키지, rCharts 패키지

base plot : 빠르게 데이터 탐색용

ggplot2 : 보다 정교한 그래프 특징 나타내기

rCharts(d3, javascript) : Report를 위한 그래프

## 03. 고수준 함수, 저수준 함수

plot, boxplot, hist

title, lines, points(그래프 타이틀, X축 이름, Y축 이름, 그래프 색 바꾸기)

## 1-1 데이터 불러오기

```
setwd("D:/dataset/R2_ex")
```

```
DF <- read.csv("example_studentlist.csv")
```

```
attach(DF)
```

```
str(DF)
```

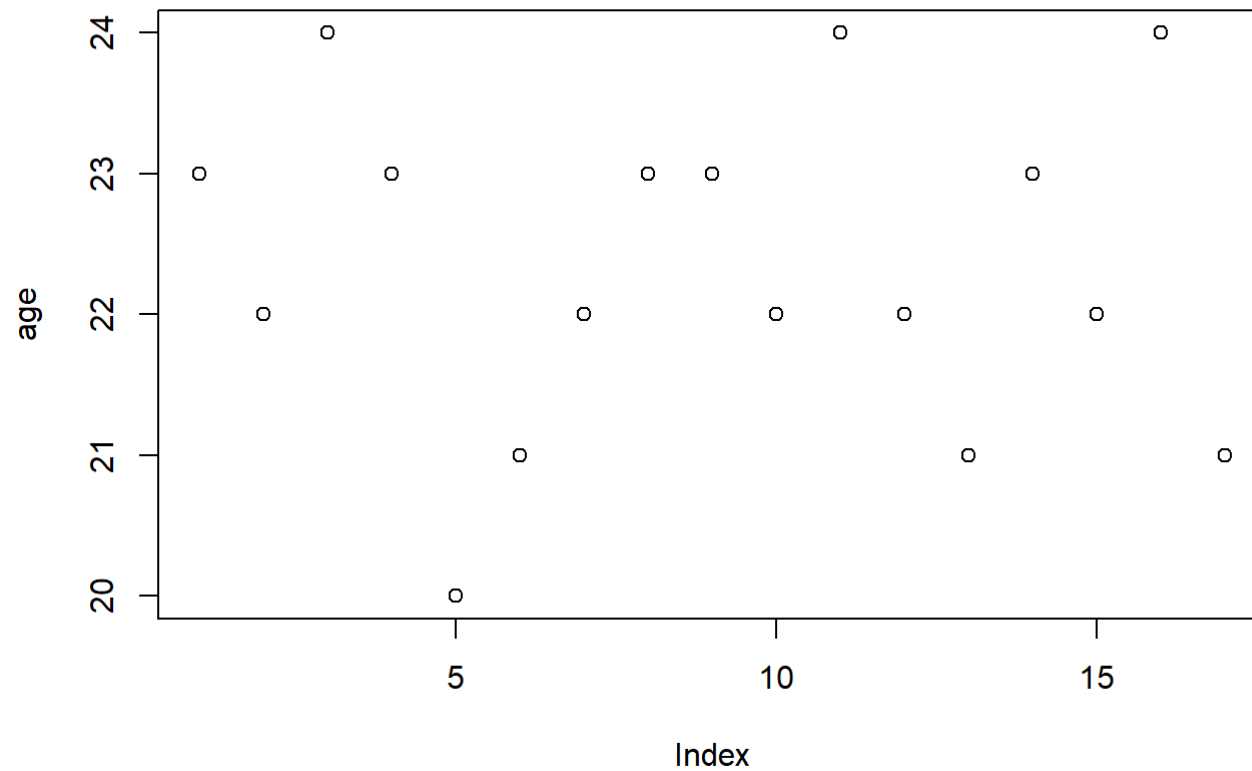
```
## 'data.frame':   17 obs. of  8 variables:
## $ name      : Factor w/ 17 levels "강수천","김길동",...: 2 12 17 6 10 7 1 14 13 9 ...
## $ sex       : Factor w/ 2 levels "남자","여자": 1 2 1 1 2 2 2 1 1 ...
## $ age       : int   23 22 24 23 20 21 22 23 23 22 ...
## $ grade     : int   3 2 4 3 1 2 1 1 3 2 ...
## $ absence   : Factor w/ 2 levels "무","유": 2 1 1 1 2 1 1 1 1 1 ...
## $ bloodtype : Factor w/ 4 levels "A","AB","B","O": 4 2 3 2 1 4 4 1 3 3 ...
## $ height    : num   165 170 175 182 168 ...
## $ weight    : num   68.2 53 80.1 85.7 49.5 52 45.3 55 64.2 61.3 ...
```

DF

```
##      name sex age grade absence bloodtype height weight
## 1 김길동 남자 23    3     유         0 165.3   68.2
## 2 이미린 여자 22    2     무        AB 170.1   53.0
## 3 홍길동 남자 24    4     무         B 175.0   80.1
## 4 김철수 남자 23    3     무        AB 182.1   85.7
## 5 손세수 여자 20    1     유         A 168.0   49.5
## 6 박미희 여자 21    2     무         0 162.0   52.0
## 7 강수천 여자 22    1     무         0 155.2   45.3
## 8 이희수 여자 23    1     무         A 176.9   55.0
## 9 이철린 남자 23    3     무         B 178.5   64.2
## 10 방희철 남자 22    2     무         B 176.1   61.3
## 11 박수호 남자 24    4     유         0 167.1   62.0
## 12 임동민 남자 22    2     무        AB 180.0   75.8
## 13 김민수 남자 21    1     무         A 162.2   55.3
## 14 이희진 여자 23    3     무         0 176.1   53.1
## 15 김미진 여자 22    2     무         B 158.2   45.2
## 16 김동수 남자 24    4     유         B 168.6   70.2
## 17 여수근 남자 21    1     무         A 169.2   62.2
```

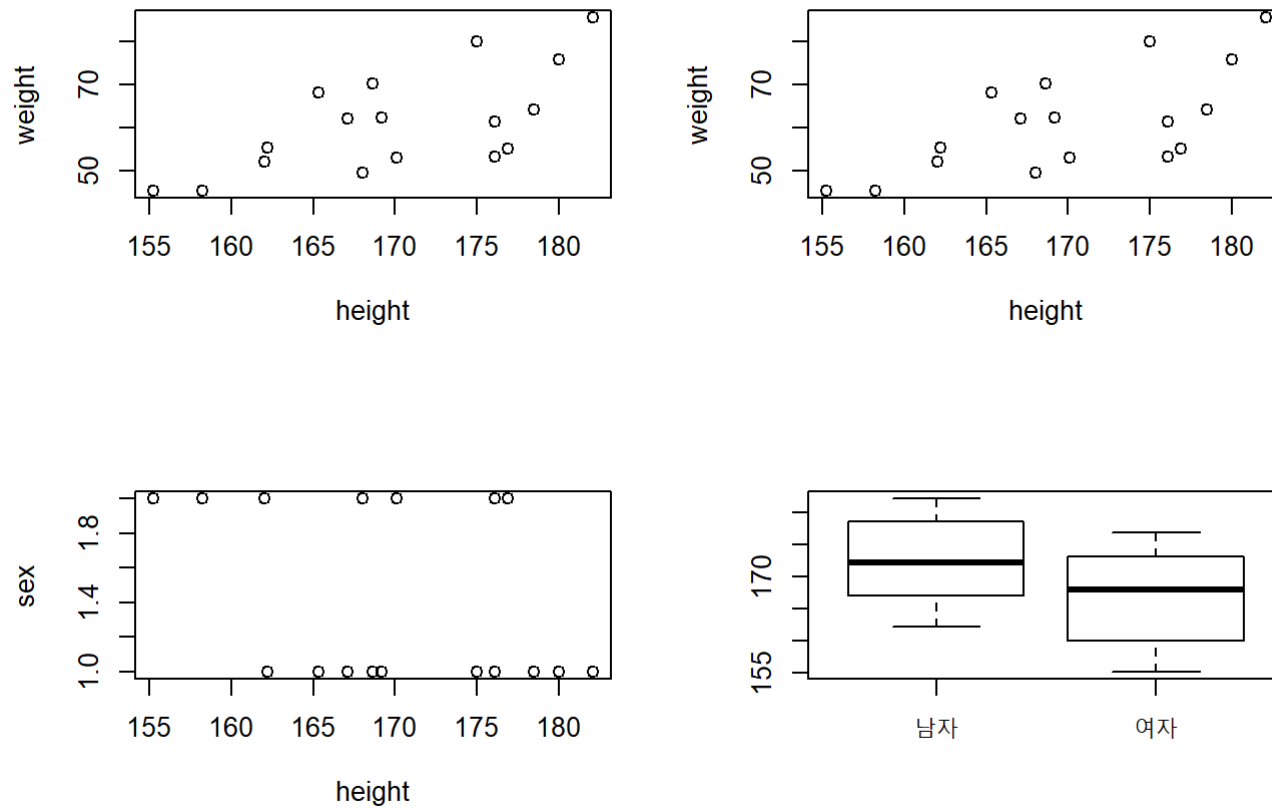
## 1-2 plot() : 값을 하나 갖는 경우,

```
plot(age)
```



1-2 plot() : 값을 두개 갖는 경우,

```
par(mfrow=c(2,2))  
plot(height, weight)  
plot(weight~height)  
plot(height, sex)  
plot(sex,height) # 성별에 따른 키의 boxplot
```



## 1-3 plot() : 변수간의 관계를 표시하기 위한 필수. scatter plot

```
names(DF)
```

```
## [1] "name"    "sex"     "age"     "grade"   "absence" "bloodtype"
## [7] "height"  "weight"
```

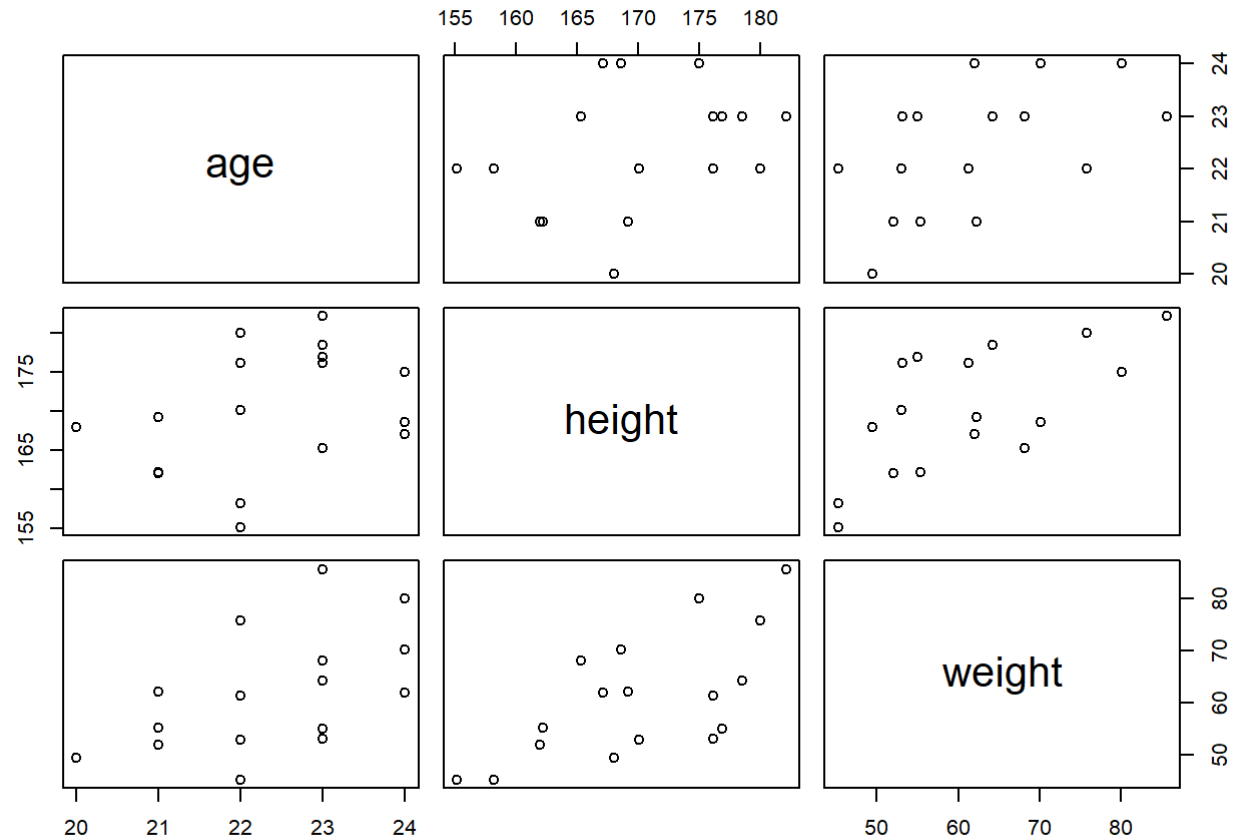
```
str(DF)
```

```
## 'data.frame': 17 obs. of 8 variables:
## $ name      : Factor w/ 17 levels "강수천","김길동",...: 2 12 17 6 10 7 1 14 13 9 ...
## $ sex       : Factor w/ 2 levels "남자","여자": 1 2 1 1 2 2 2 2 1 1 ...
## $ age       : int 23 22 24 23 20 21 22 23 23 22 ...
## $ grade     : int 3 2 4 3 1 2 1 1 3 2 ...
## $ absence   : Factor w/ 2 levels "무","유": 2 1 1 1 2 1 1 1 1 1 ...
## $ bloodtype : Factor w/ 4 levels "A","AB","B","O": 4 2 3 2 1 4 4 1 3 3 ...
## $ height    : num 165 170 175 182 168 ...
## $ weight    : num 68.2 53 80.1 85.7 49.5 52 45.3 55 64.2 61.3 ...
```

```
DF2 <- data.frame(age, height, weight)
DF2
```

```
##   age height weight
## 1  23  165.3   68.2
## 2  22  170.1   53.0
## 3  24  175.0   80.1
## 4  23  182.1   85.7
## 5  20  168.0   49.5
## 6  21  162.0   52.0
## 7  22  155.2   45.3
## 8  23  176.9   55.0
## 9  23  178.5   64.2
## 10 22  176.1   61.3
## 11 24  167.1   62.0
## 12 22  180.0   75.8
## 13 21  162.2   55.3
## 14 23  176.1   53.1
## 15 22  158.2   45.2
## 16 24  168.6   70.2
## 17 21  169.2   62.2
```

```
plot(DF2)
```



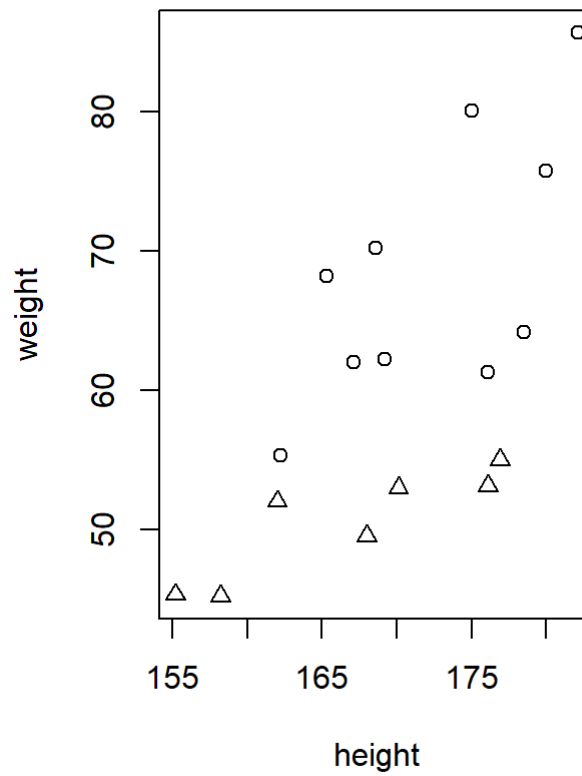
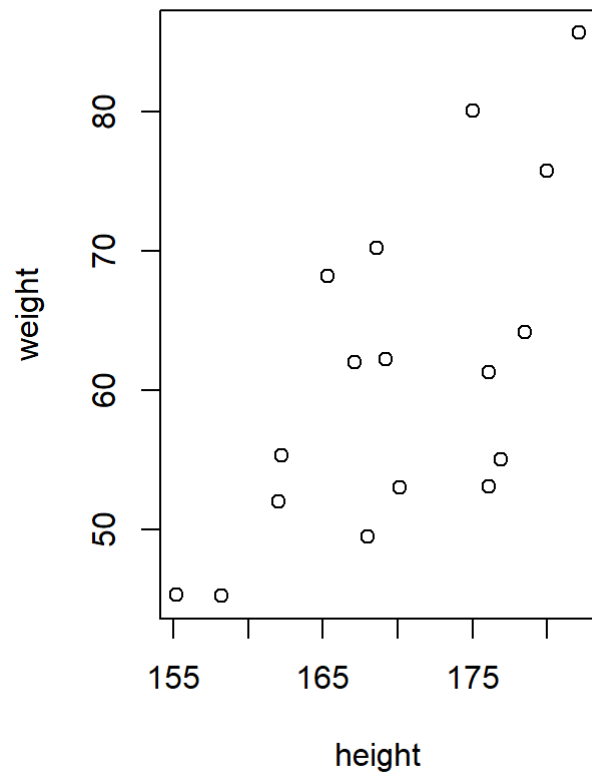
## 1-4 plot() : pch 속성

점의 형태를 이용해서 남자와 여자를 구분하여 표시

```
dat <- data.frame(sex, height, weight)
dat
```

```
##      sex height weight
## 1  남자  165.3   68.2
## 2  여자  170.1   53.0
## 3  남자  175.0   80.1
## 4  남자  182.1   85.7
## 5  여자  168.0   49.5
## 6  여자  162.0   52.0
## 7  여자  155.2   45.3
## 8  여자  176.9   55.0
## 9  남자  178.5   64.2
## 10 남자  176.1   61.3
## 11 남자  167.1   62.0
## 12 남자  180.0   75.8
## 13 남자  162.2   55.3
## 14 여자  176.1   53.1
## 15 여자  158.2   45.2
## 16 남자  168.6   70.2
## 17 남자  169.2   62.2
```

```
par(mfrow=c(1,2))
plot(weight~height)
plot(weight~height, pch=as.integer(sex))
```



## pch

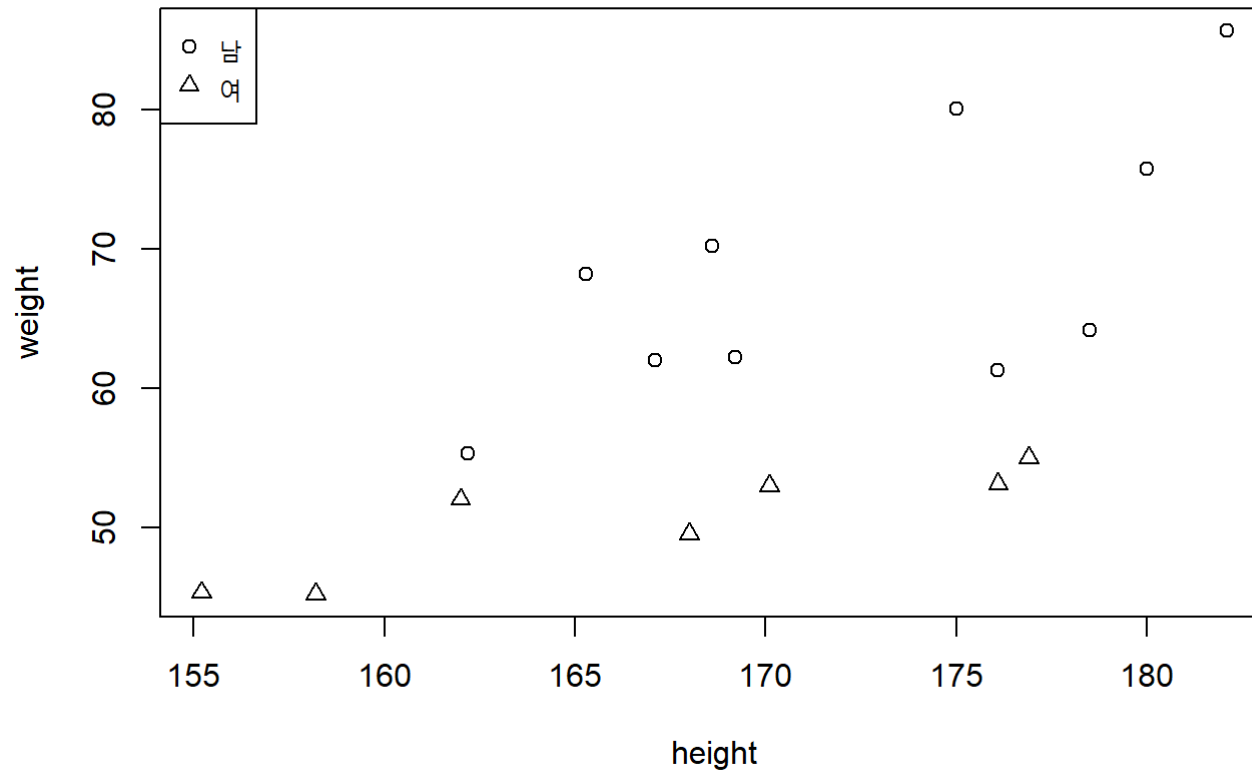
총 18개까지 가능하다.

dat



```
##      sex height weight
## 1  남자  165.3   68.2
## 2  여자  170.1   53.0
## 3  남자  175.0   80.1
## 4  남자  182.1   85.7
## 5  여자  168.0   49.5
## 6  여자  162.0   52.0
## 7  여자  155.2   45.3
## 8  여자  176.9   55.0
## 9  남자  178.5   64.2
## 10 남자  176.1   61.3
## 11 남자  167.1   62.0
## 12 남자  180.0   75.8
## 13 남자  162.2   55.3
## 14 여자  176.1   53.1
## 15 여자  158.2   45.2
## 16 남자  168.6   70.2
## 17 남자  169.2   62.2
```

```
plot(weight~height, pch=as.integer(sex))
legend("topleft", c("남", "여"), pch=dat$sex) # 별표
```

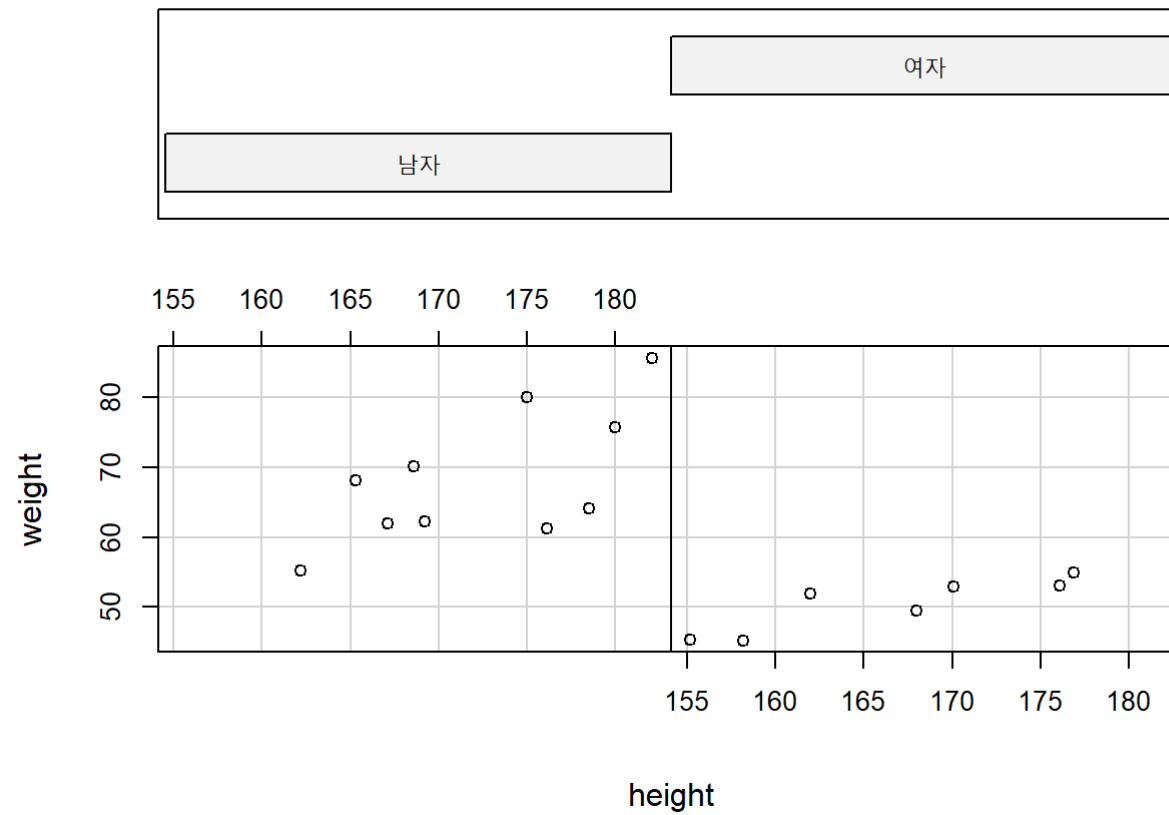


## 1-5 coplot() 함수

- (가) 명목형 변수(범주형)에 대한 그래프
- (나) Levels별 그래프를 보는 것이 목적이다.
- (다) 남녀 성별에 따른 키와 몸무게의 상관관계

```
coplot(weight~height | sex)
```

Given : sex



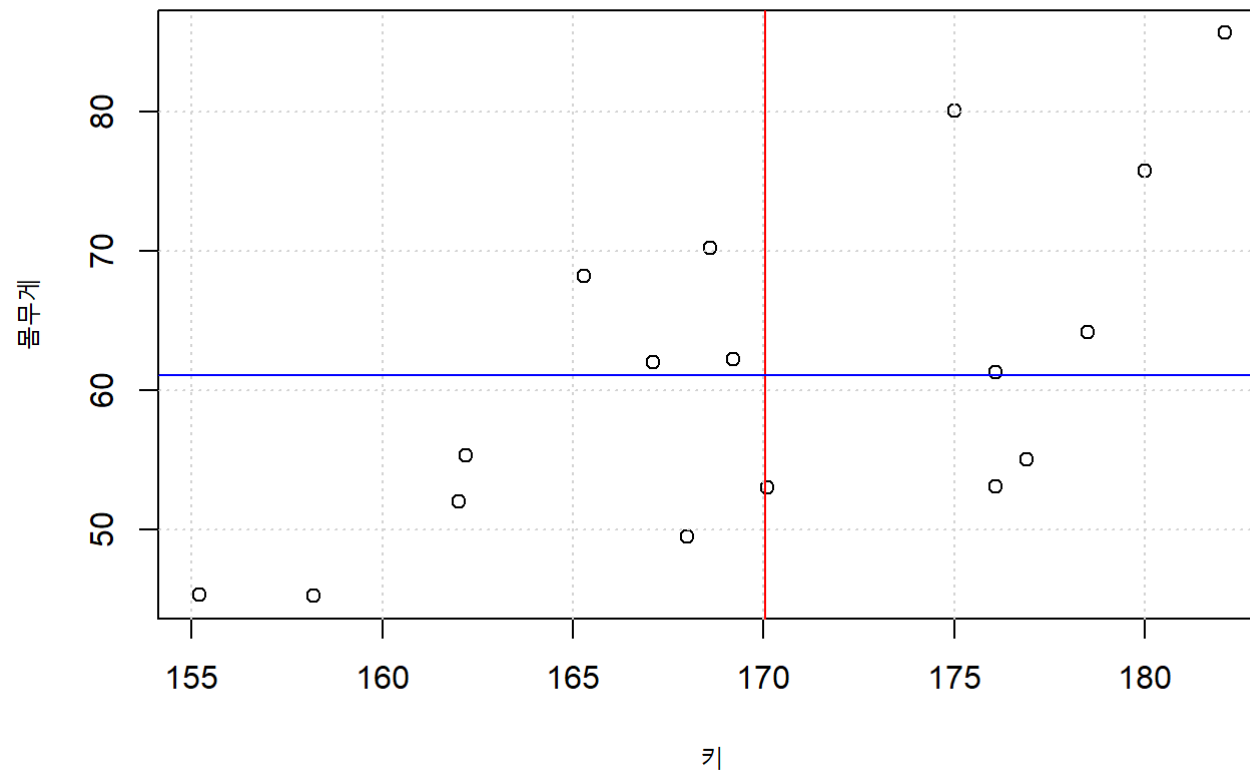
## 1-6 저수준 함수 사용

`title()` : 그래프의 타이틀 함수  
`grid()` : 격자 추가 함수  
`abline()` : 수직선, 수평선 추가

```
plot(weight~height, ann=FALSE)
title(main="A와 B의 몸무게와 키의 상관관계")
title(xlab="키", ylab="몸무게")
grid() # 격자 추가
```

```
hMean <- mean(height)
wMean <- mean(weight)
abline(v=hMean, col="red")
abline(h=wMean, col="blue")
```

### A와 B의 몸무게와 키의 상관관계



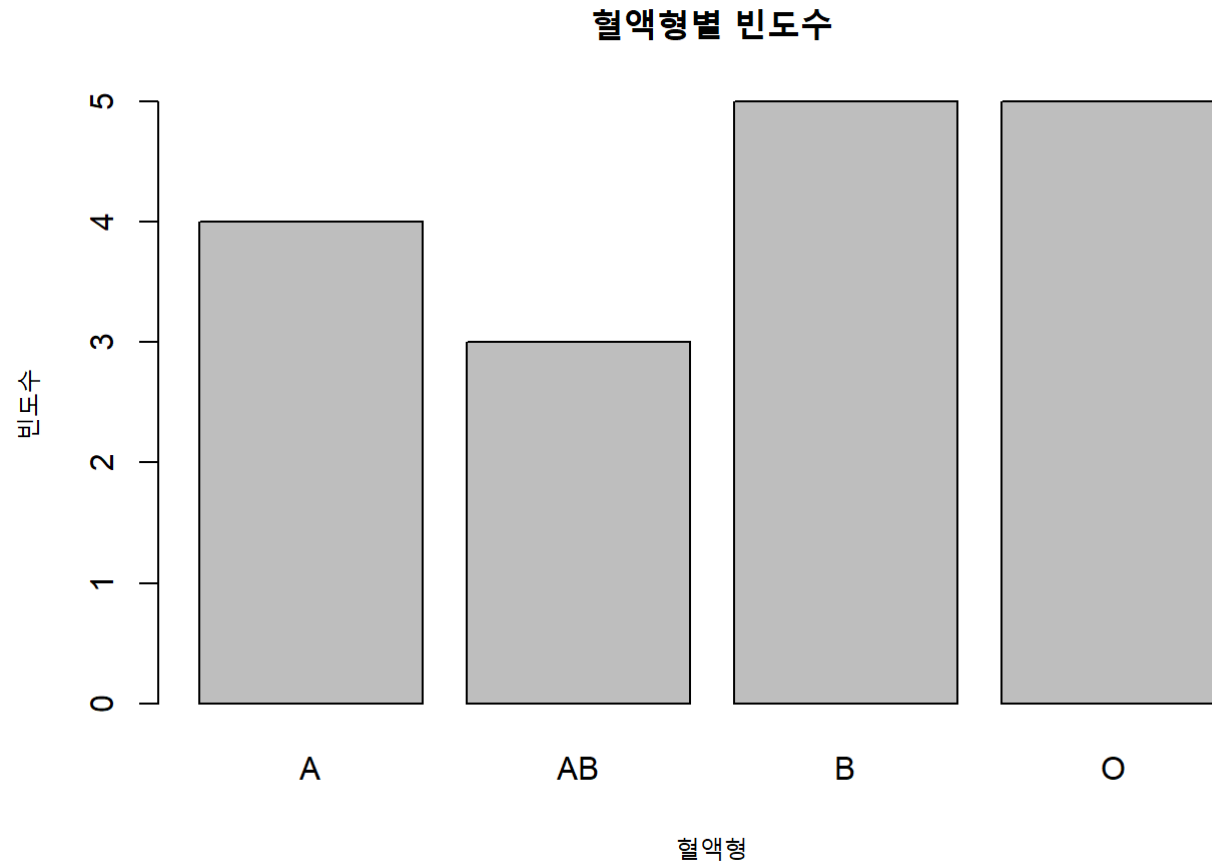
# 1-7 barplot(막대 그래프)

(가) 범주형(명목형) 변수 데이터 특징 살펴 보기  
(나) 먼저 빈도수를 확인해야 한다. 이를 위해 table() 함수 이용  
(다) 사용법 : barplot(변수객체)

```
tblBlood <- table(DF$bloodtype)
tblBlood
```

```
##
##  A AB  B  0
##  4  3  5  5
```

```
barplot(tblBlood)
title(main="혈액형별 빈도수")
title(xlab="혈액형", ylab="빈도수" )
```



1-7 barplot(막대 그래프)

barplot 혈액형별 키의 평균

```
[사용법]
tapply(X, INDEX, FUN = NULL, ...,)
X : 객체
INDEX : 하나 또는 그 이상의 범주형 리스트 (factor)
...
FUN : 적용할 함수
```

```
?tapply
```

```
## starting httpd help server ... done
```

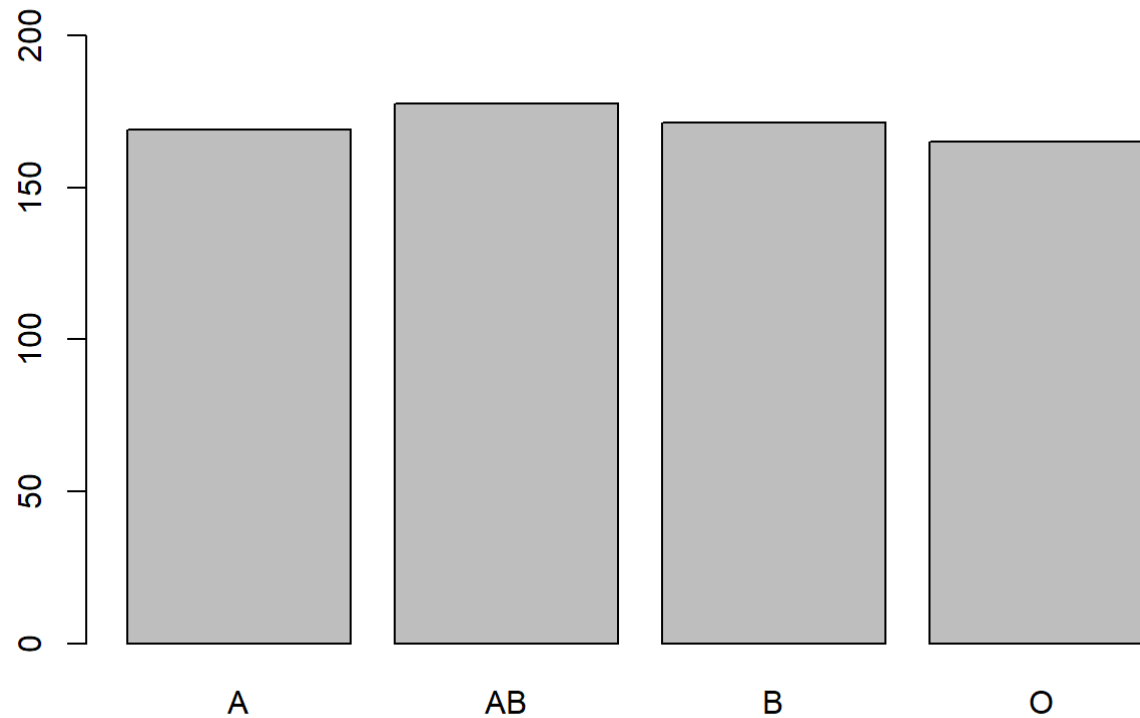
```
class(bloodtype)
```

```
## [1] "factor"
```

```
Height = tapply(height, bloodtype, mean )
Height
```

```
##      A      AB      B      O
## 169.075 177.400 171.280 165.140
```

```
barplot(Height, ylim=c(0,200))
```



## 1-7 barplot(막대 그래프)

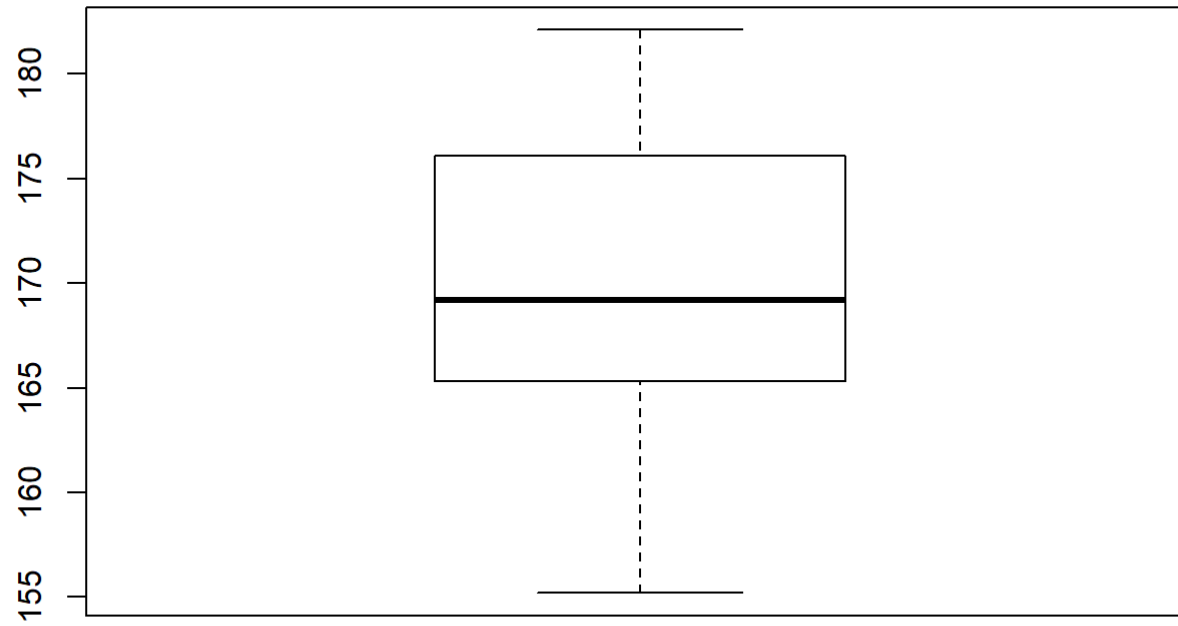
### boxplot 혈액형별(levels) 키에 대한 분포 그려보기

`boxplot(VariableA)` => 변수 하나의 경우, 분포

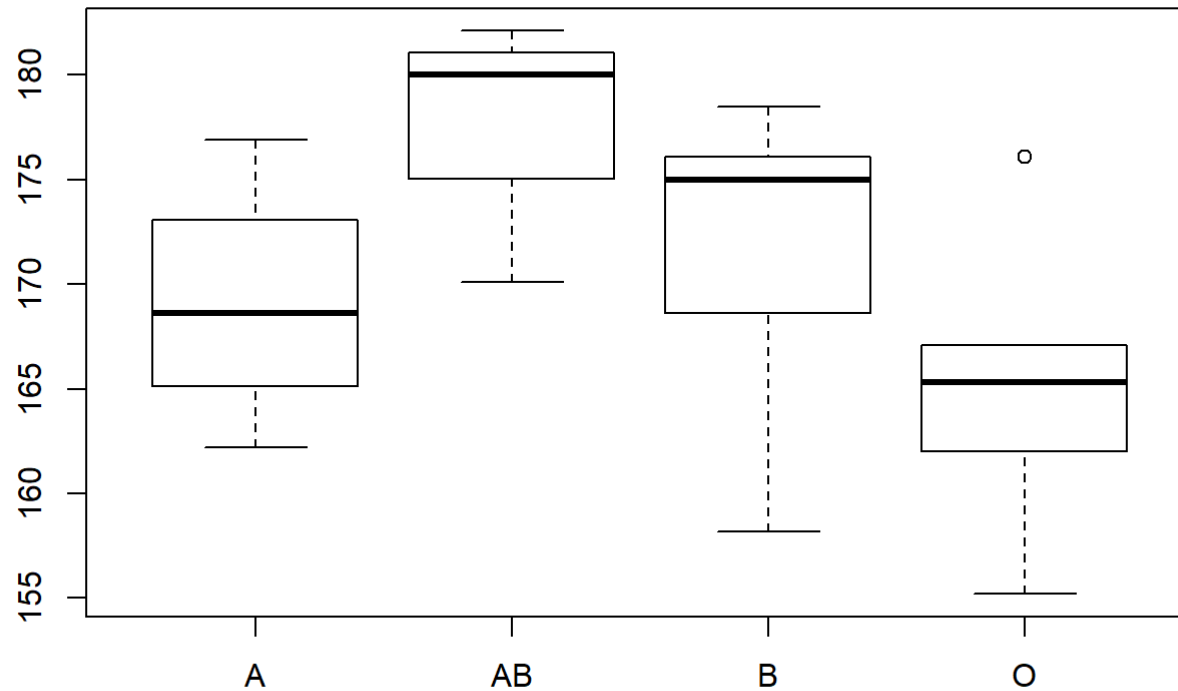
`boxplot(A ~ B)` => B(독립변수, 설명변수), A(종속변수)의 분포 (B는 범주형 변수이어야 함.)

`boxplot(height)`





```
boxplot(height~bloodtype)
```



## 1-7 hist() : 히스토그램 살펴보기

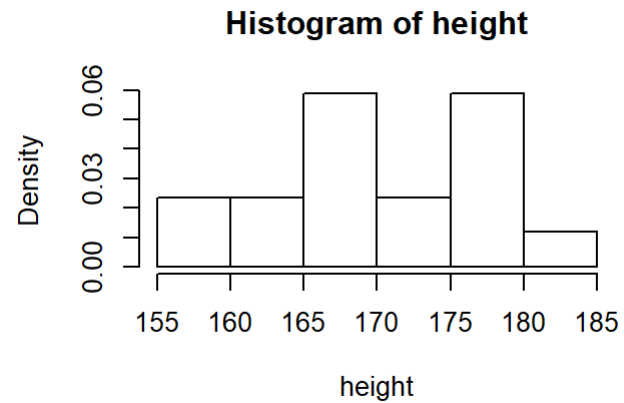
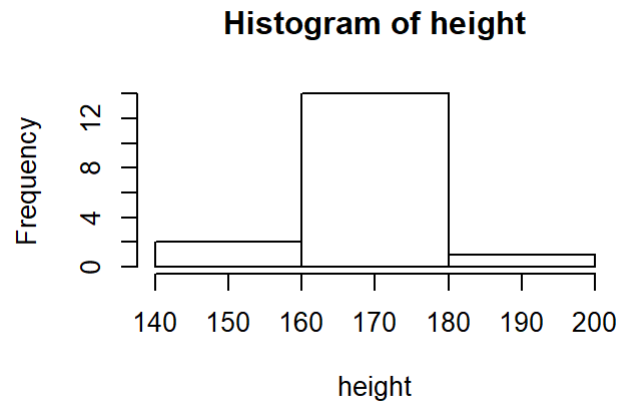
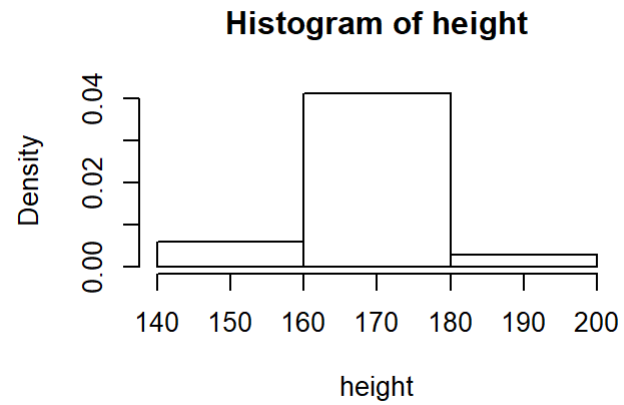
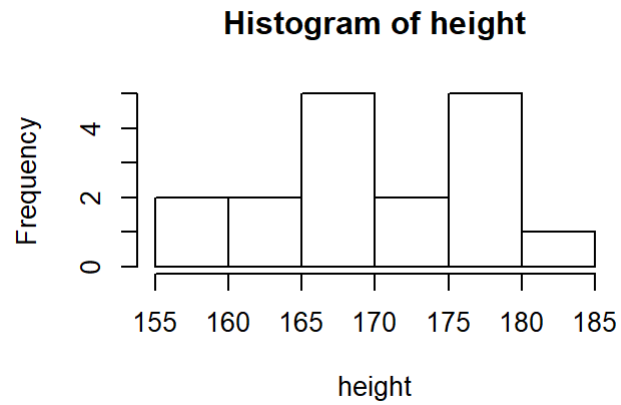
### hist

```
hist(하나의 변수) # 연속형변수  
prob : 확률(상대도수)  
breaks : 나누기
```

```
height
```

```
## [1] 165.3 170.1 175.0 182.1 168.0 162.0 155.2 176.9 178.5 176.1 167.1  
## [12] 180.0 162.2 176.1 158.2 168.6 169.2
```

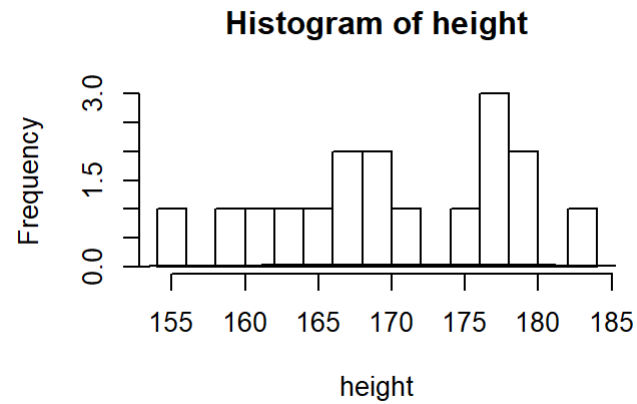
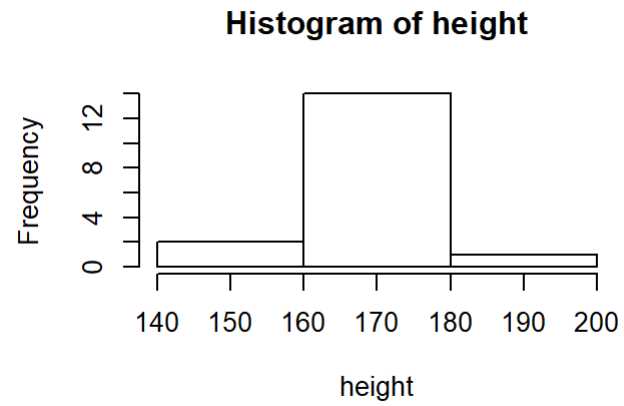
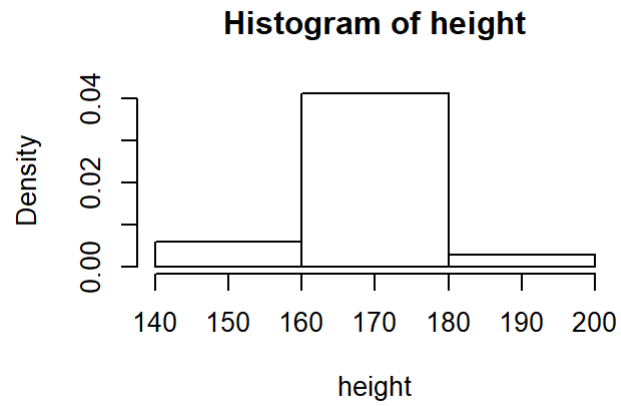
```
par(mfrow=c(2,2)) # 2행 2열  
hist(height) # 하나의 변수  
hist(height, breaks=1, prob=T)  
hist(height, breaks=1)  
hist(height, prob=T)
```



density(밀도) vs frequency(빈도)

```
par(mfrow=c(2,2))
hist(height, breaks=1, prob=T)
hist(height, breaks=1)

hist(height, breaks=10, prob=T)
hist(height, breaks=10)
lines(density(height))
```



```
length(height)
```

```
## [1] 17
```

```
print(2/34)
```

```
## [1] 0.05882353
```

```
print(14/34)
```

```
## [1] 0.4117647
```

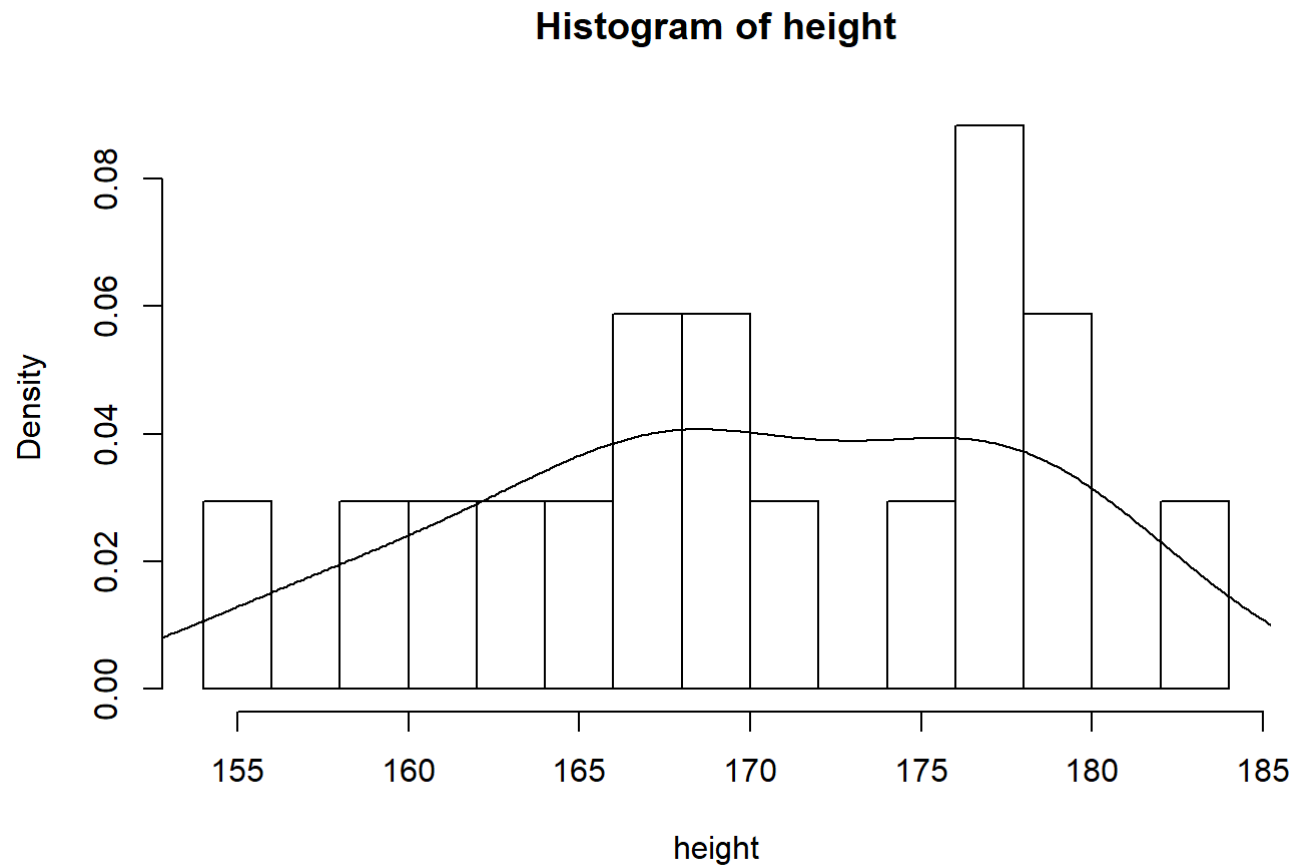
```
print(1/34)
```

```
## [1] 0.02941176
```

## 1-7 hist() : 히스토 그램 살펴보기

### 저수준 함수- 추세선 보기

```
par(mfrow=c(1,1))  
hist(height, breaks=10, prob=T)  
lines(density(height))
```



## 1-7 hist() : 히스토 그램 살펴보기

(가) seq() 함수  
seq(시작, 끝, by=증가)

(나) hist(객체, breaks=직접지정)

```
range(height)
```

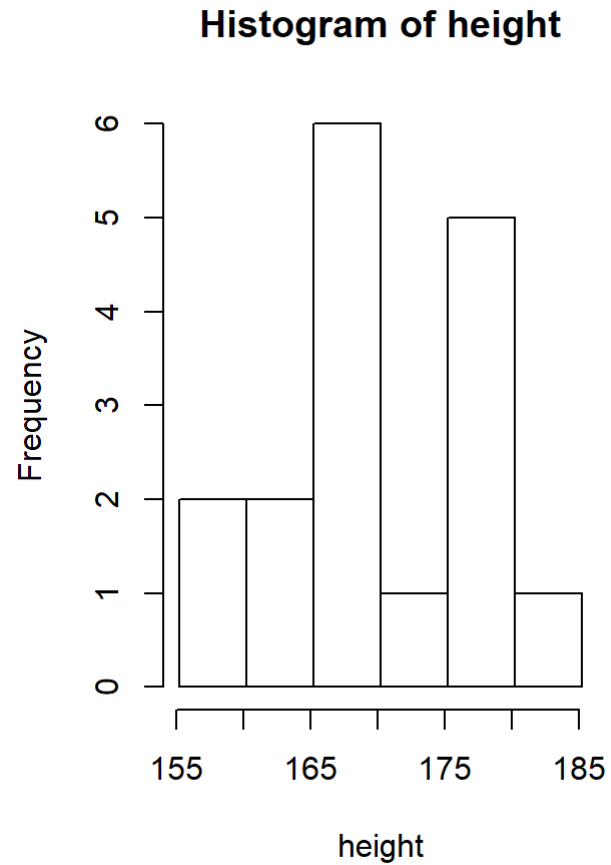
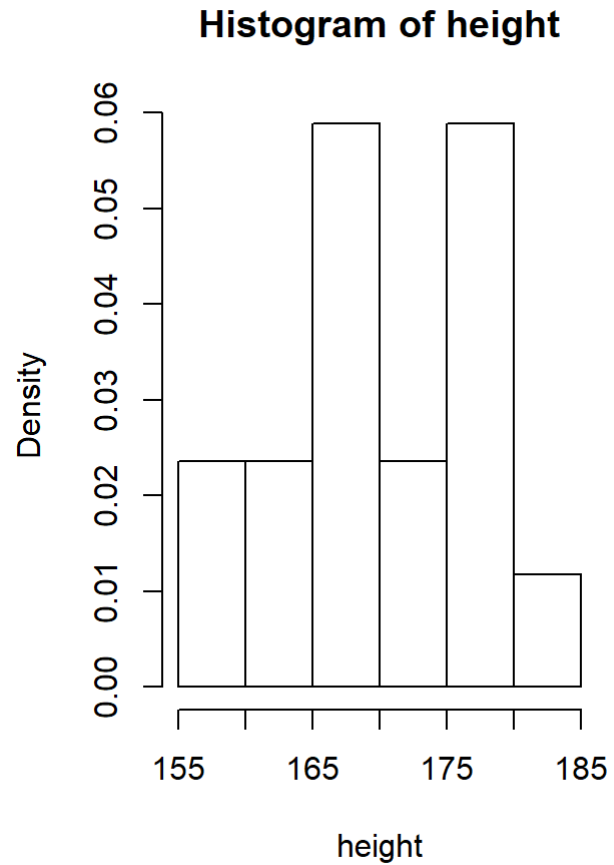
```
## [1] 155.2 182.1
```

```
BPoint = seq(min(height), max(height)+5, by=5)  
BPoint
```

```
## [1] 155.2 160.2 165.2 170.2 175.2 180.2 185.2
```

```
par(mfrow=c(1,2))  
hist(height, breaks=5, prob=T)  
hist(height, breaks=BPoint)
```



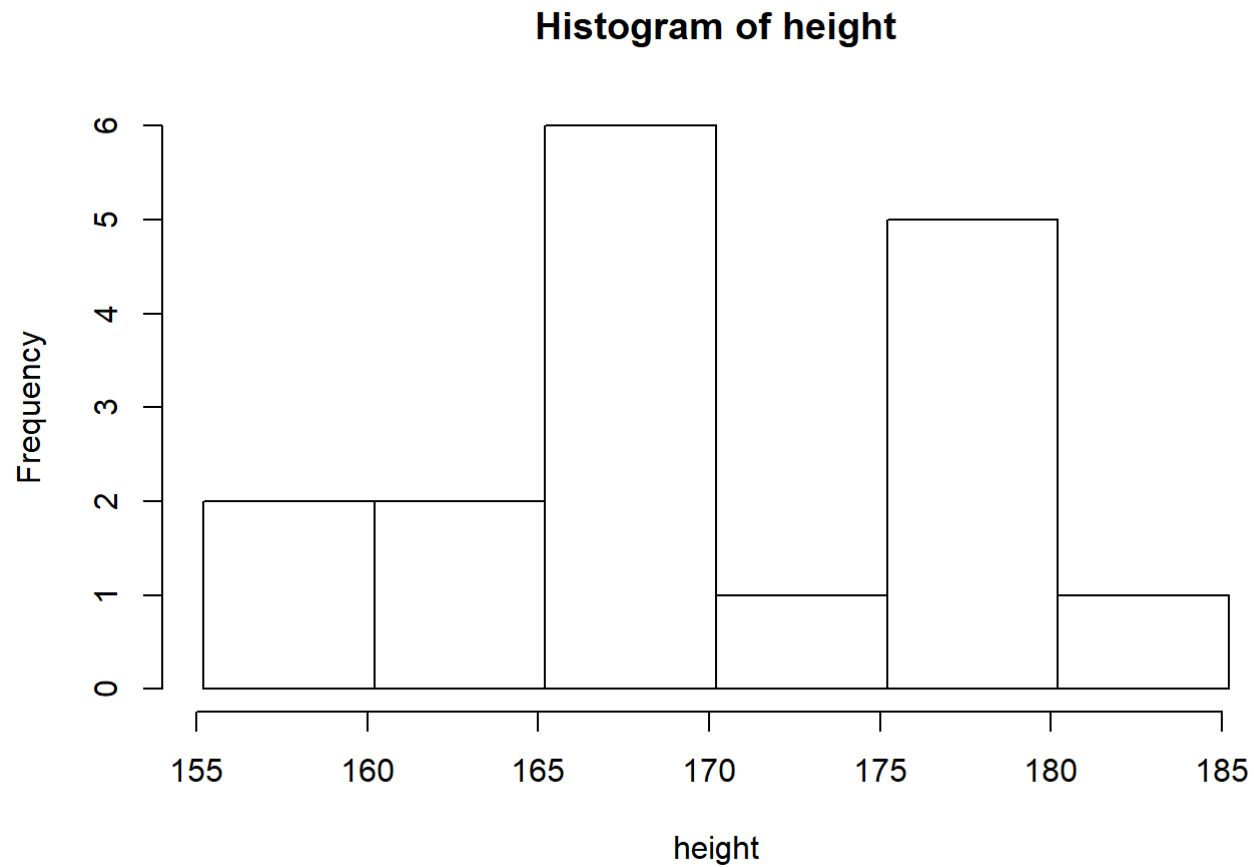


## breaks 포인트를 5로 나누기

```
BPoint = seq(min(height), max(height)+5, by=5)
BPoint
```

```
## [1] 155.2 160.2 165.2 170.2 175.2 180.2 185.2
```

```
hist(height, breaks=BPoint)
```

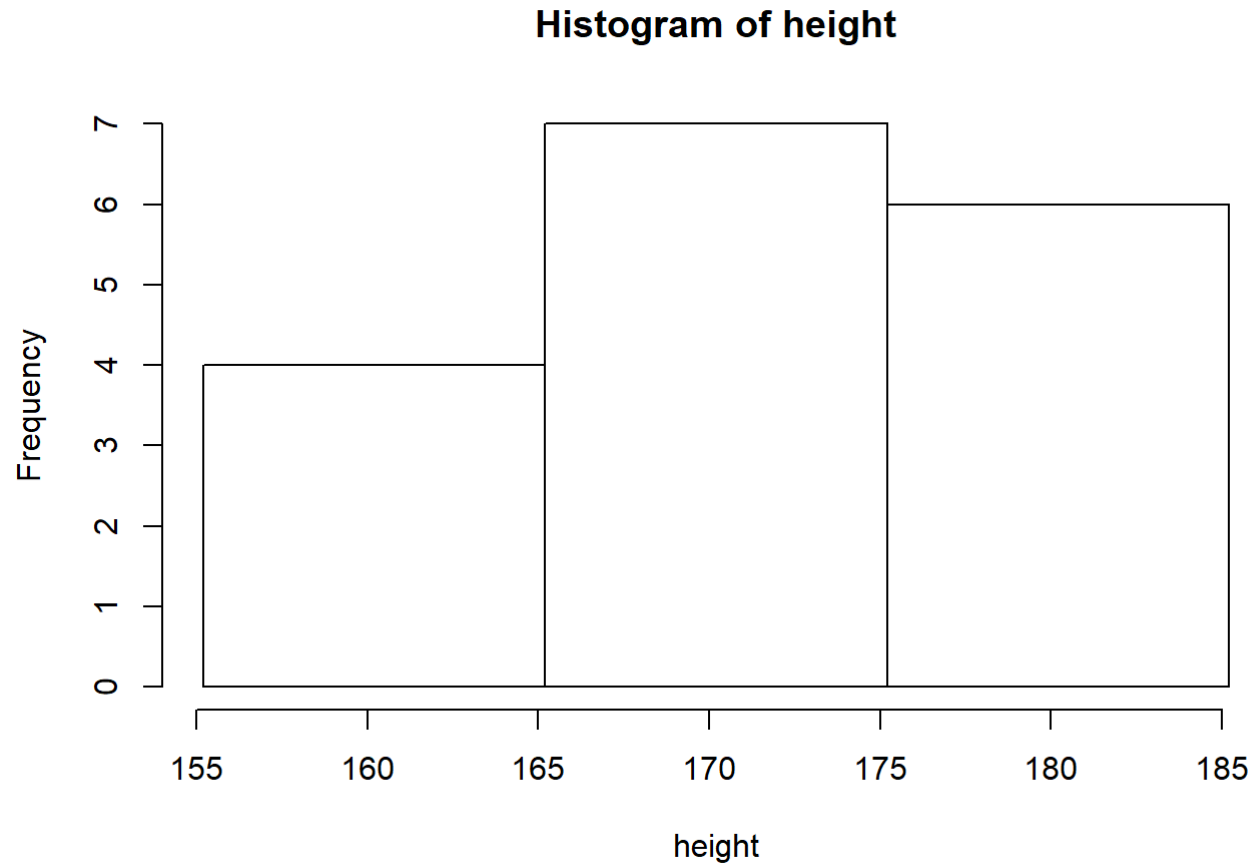


## breaks 포인트를 10로 나누기

```
BPoint = seq(min(height), max(height)+10, by=10)  
BPoint
```

```
## [1] 155.2 165.2 175.2 185.2
```

```
hist(height, breaks=BPoint)
```



2,3 열의 그래프를 그린다.

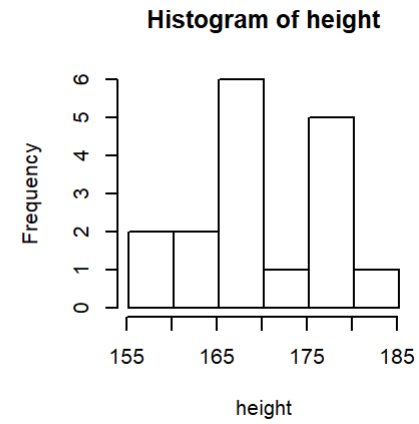
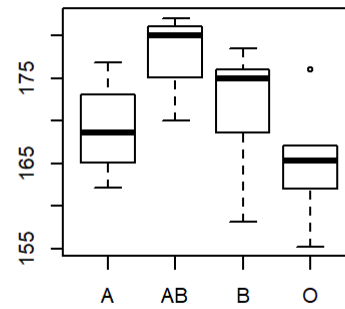
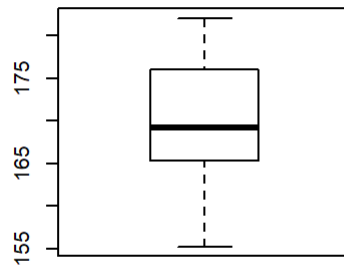
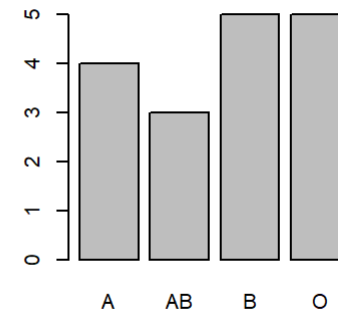
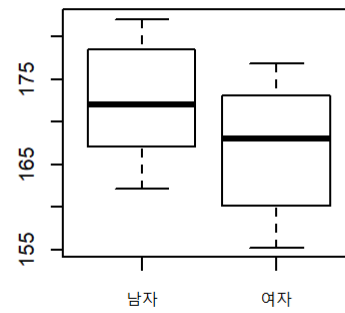
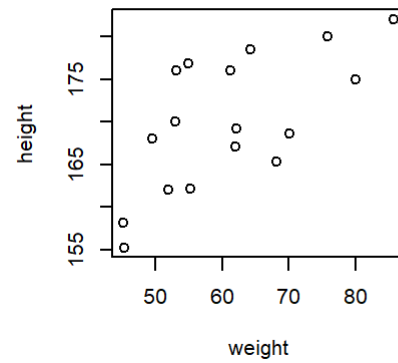
1행 1열 height, weight 에 대한 산점도 1행 2열 성별, weight에 대한 Boxplot 그리기 1행 3열에는 혈액형별 빈도수 구하기 barplot 2행 1열 boxplo

```
par(mfrow=c(2,3))

plot(weight, height)
plot(sex, height)
barplot(table(bloodtype))
boxplot(height)
boxplot(height~bloodtype)
BPoint = seq(min(height), max(height)+5, by=5)
BPoint
```

```
## [1] 155.2 160.2 165.2 170.2 175.2 180.2 185.2
```

```
hist(height, breaks=BPoint)
```



한 그래프에 두개의 그래프 그려보기

```
TS1 <- c(round(runif(30)*100))  
TS2 <- c(round(runif(30)*100))  
TS3 <- c(round(runif(30)*100))  
TS1
```

```
## [1]  8 47  1 50 13 10  3 34 66 26 26 38 24 31 14 35 88  4 95  8 30 64 66  
## [24] 48 46 51 59 57 75 51
```

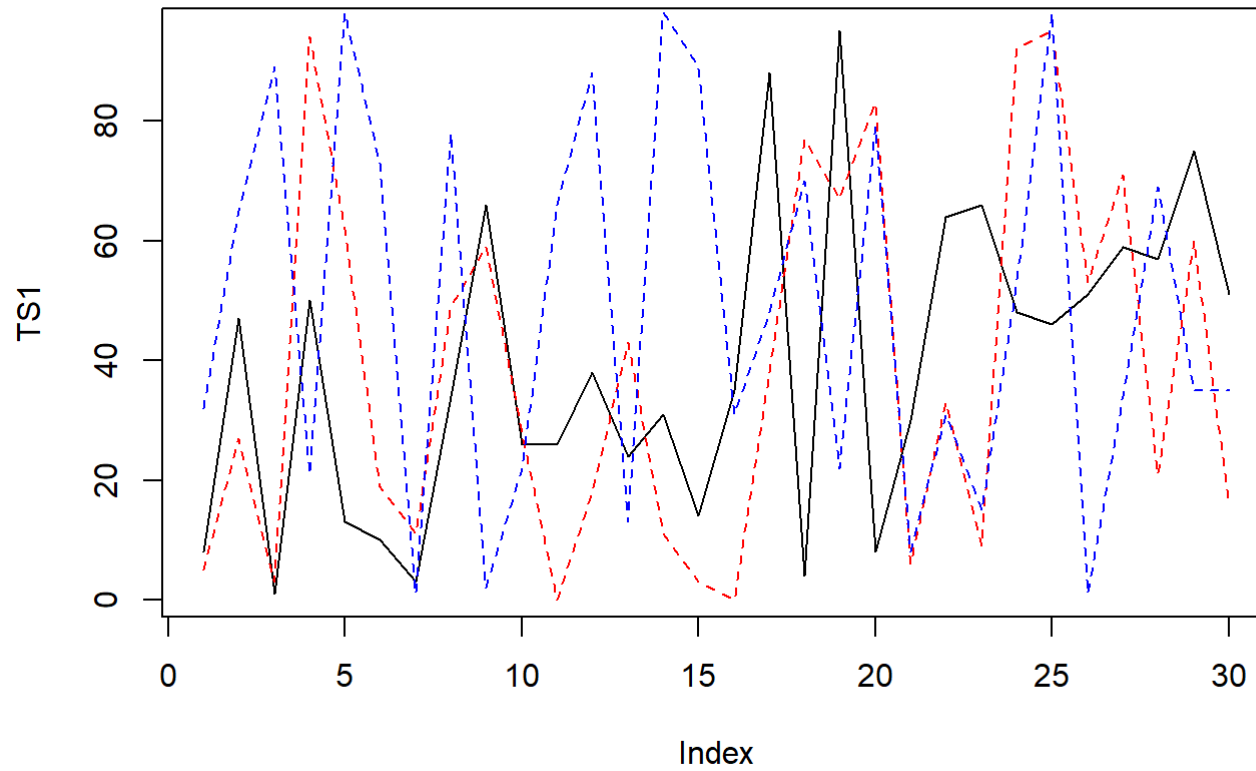
```
TS2
```

```
## [1]  5 27  3 94 62 19 11 49 59 28  0 18 43 11  3  0 38 77 67 83  6 33  9  
## [24] 92 95 53 71 21 60 16
```

```
TS3
```

```
## [1] 32 65 89 21 98 73  1 78  2 22 66 88 13 98 89 31 48 70 22 79  8 31 15  
## [24] 53 98  1 34 69 35 35
```

```
plot(TS1, type="l")  
lines(TS2, col="red", lty="dashed")  
lines(TS3, col="blue", lty="dashed")
```

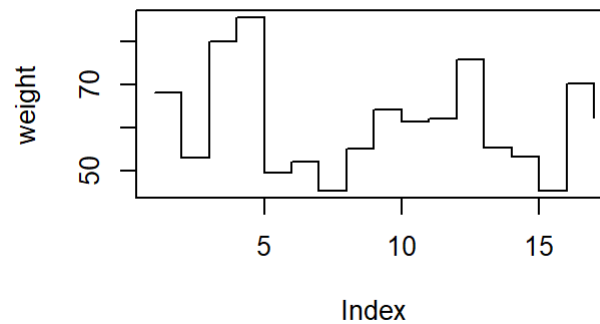
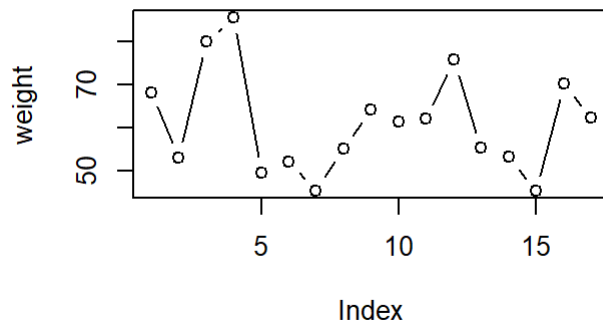
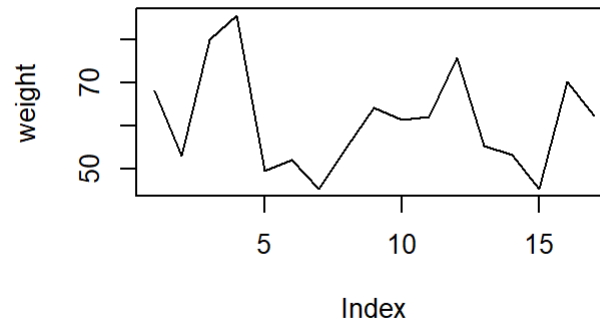
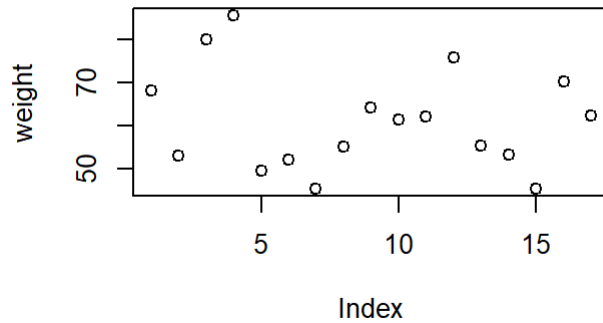


```
runif(30)*100
```

```
## [1] 93.655458 90.014450 84.940736 11.199889 62.069550 16.474411 51.390581  
## [8] 24.615840 39.604869 15.786344 98.682077 52.043595 44.352451 83.154918  
## [15] 75.744201 31.107438 88.280991 5.842717 30.018740 19.615069 64.291585  
## [22] 27.565219 83.704453 3.334375 27.892400 80.308224 14.779039 40.538484  
## [29] 76.571974 90.488745
```



```
#help(plot)  
par(mfrow=c(2,2))  
plot(weight, type="p")  
plot(weight, type="l")  
plot(weight, type="b")  
plot(weight, type="s")
```



## ggplot2

```
#install.packages("ggplot2")
#install.packages("ggthemes")
library(ggplot2)
library(ggthemes)
```

## ggplot2

```
dataset : diamond
```

```
dim(diamonds)
```

```
## [1] 53940    10
```

```
str(diamonds)
```

```
## Classes 'tbl_df', 'tbl' and 'data.frame':   53940 obs. of  10 variables:
## $ carat   : num  0.23 0.21 0.23 0.29 0.31 0.24 0.24 0.26 0.22 0.23 ...
## $ cut     : Ord.factor w/ 5 levels "Fair"<"Good"<..: 5 4 2 4 2 3 3 1 3 ...
## $ color   : Ord.factor w/ 7 levels "D"<"E"<"F"<"G"<..: 2 2 2 6 7 7 6 5 2 5 ...
## $ clarity: Ord.factor w/ 8 levels "I1"<"SI2"<"SI1"<..: 2 3 5 4 2 6 7 3 4 5 ...
## $ depth   : num  61.5 59.8 56.9 62.4 63.3 62.8 62.3 61.9 65.1 59.4 ...
## $ table   : num  55 61 65 58 58 57 57 55 61 61 ...
## $ price   : int  326 326 327 334 335 336 336 337 337 338 ...
## $ x       : num  3.95 3.89 4.05 4.2 4.34 3.94 3.95 4.07 3.87 4 ...
## $ y       : num  3.98 3.84 4.07 4.23 4.35 3.96 3.98 4.11 3.78 4.05 ...
## $ z       : num  2.43 2.31 2.31 2.63 2.75 2.48 2.47 2.53 2.49 2.39 ...
```

```
head(diamonds)
```

```
## # A tibble: 6 x 10
##   carat cut      color clarity depth table price     x     y     z
##   <dbl> <ord>    <ord> <ord>    <dbl> <dbl> <int> <dbl> <dbl> <dbl>
## 1 0.23 Ideal    E      SI2     61.5   55   326  3.95  3.98  2.43
## 2 0.21 Premium  E      SI1     59.8   61   326  3.89  3.84  2.31
## 3 0.23 Good     E      VS1     56.9   65   327  4.05  4.07  2.31
## 4 0.290 Premium I      VS2     62.4   58   334  4.2   4.23  2.63
## 5 0.31 Good     J      SI2     63.3   58   335  4.34  4.35  2.75
## 6 0.24 Very Good J      VVS2     62.8   57   336  3.94  3.96  2.48
```

```
summary(diamonds)
```

```
##      carat      cut      color      clarity
## Min.   :0.2000 Fair      : 1610 D: 6775 SI1    :13065
## 1st Qu.:0.4000 Good      : 4906 E: 9797 VS2    :12258
## Median :0.7000 Very Good:12082 F: 9542 SI2    : 9194
## Mean   :0.7979 Premium  :13791 G:11292 VS1    : 8171
## 3rd Qu.:1.0400 Ideal     :21551 H: 8304 VVS2   : 5066
## Max.   :5.0100                I: 5422 VVS1   : 3655
##                J: 2808 (Other): 2531
##
##      depth      table      price      x
## Min.   :43.00 Min.   :43.00 Min.   : 326 Min.   : 0.000
## 1st Qu.:61.00 1st Qu.:56.00 1st Qu.: 950 1st Qu.: 4.710
## Median :61.80 Median :57.00 Median : 2401 Median : 5.700
## Mean   :61.75 Mean   :57.46 Mean   : 3933 Mean   : 5.731
## 3rd Qu.:62.50 3rd Qu.:59.00 3rd Qu.: 5324 3rd Qu.: 6.540
## Max.   :79.00 Max.   :95.00 Max.   :18823 Max.   :10.740
##
##      y      z
## Min.   : 0.000 Min.   : 0.000
## 1st Qu.: 4.720 1st Qu.: 2.910
## Median : 5.710 Median : 3.530
## Mean   : 5.735 Mean   : 3.539
## 3rd Qu.: 6.540 3rd Qu.: 4.040
## Max.   :58.900 Max.   :31.800
##
```

```
names(diamonds)
```

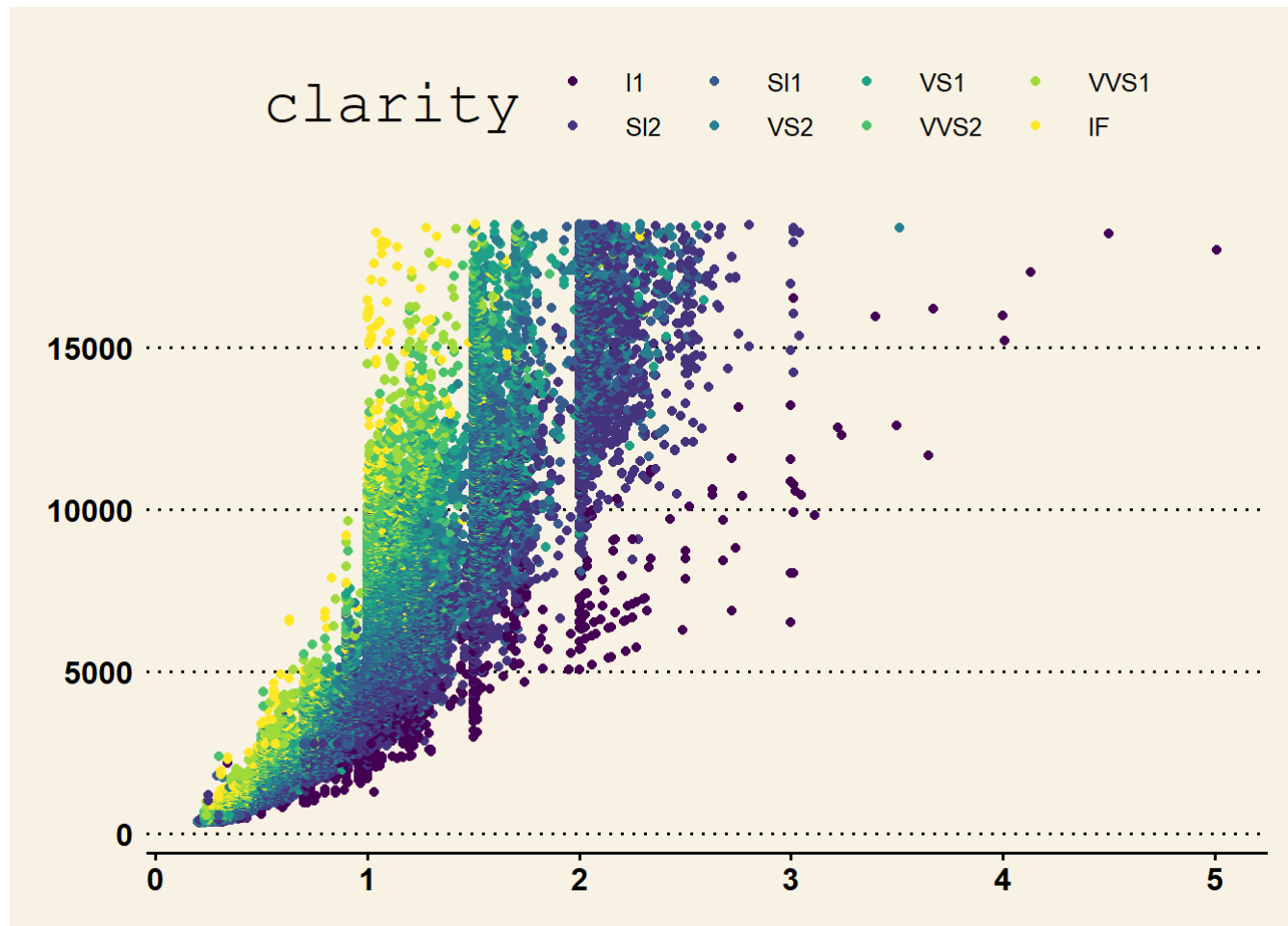
```
## [1] "carat" "cut" "color" "clarity" "depth" "table" "price"  
## [8] "x" "y" "z"
```

```
ggplot()  
geom_point()  
theme_wsj
```

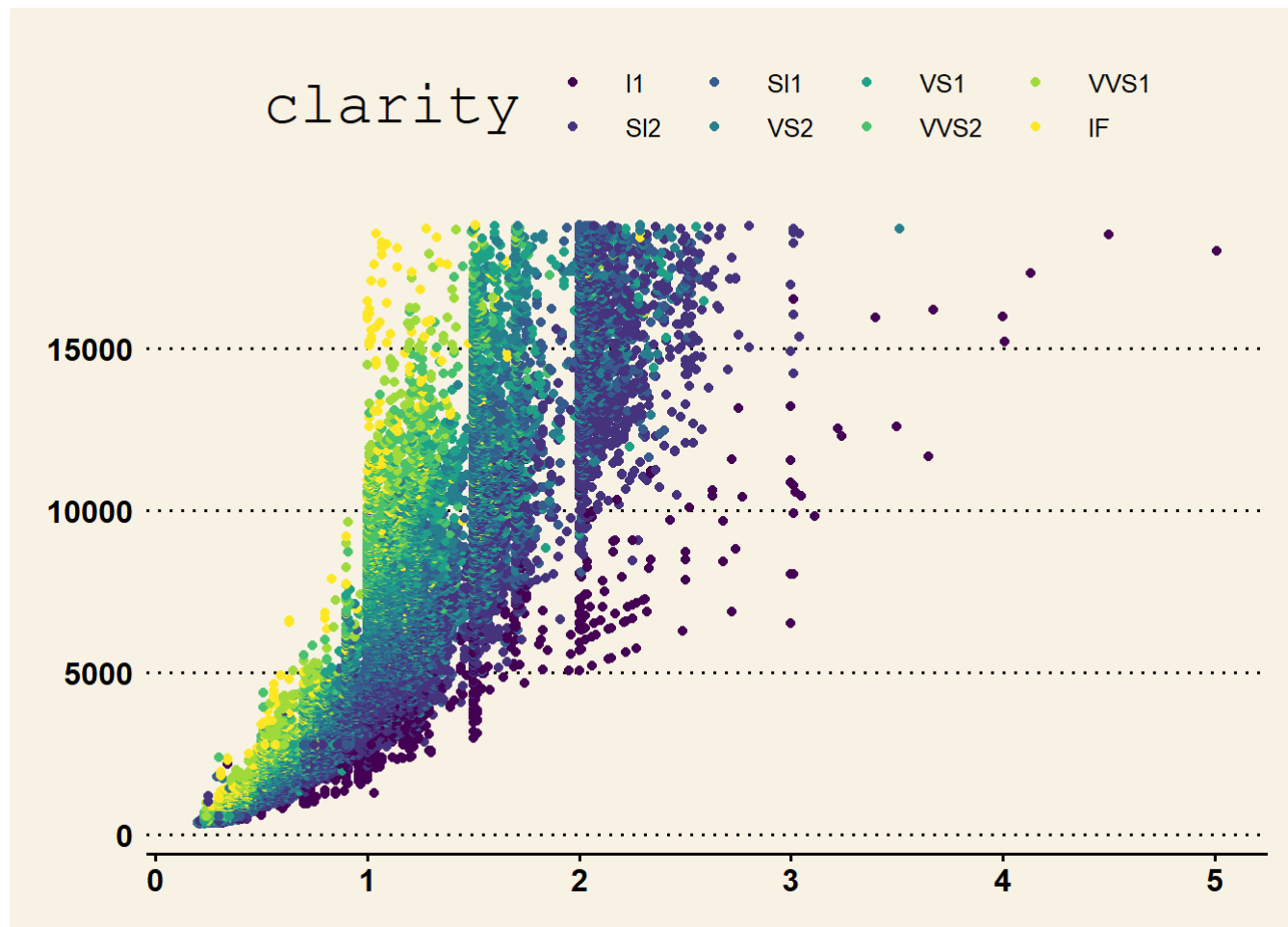
```
#help(diamonds)  
names(diamonds)
```

```
## [1] "carat" "cut" "color" "clarity" "depth" "table" "price"  
## [8] "x" "y" "z"
```

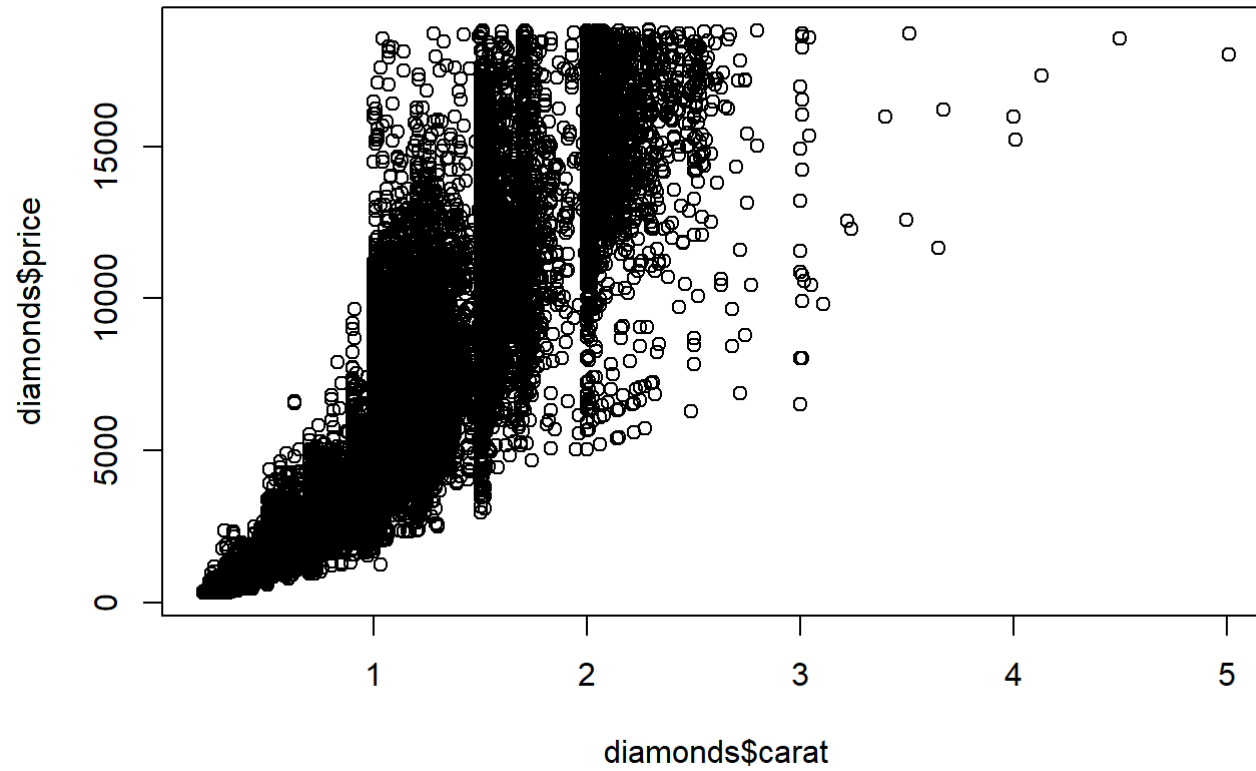
```
ggplot(data=diamonds,  
       aes(x=carat, y=price, colour=clarity)) +  
  geom_point() +  
  theme_wsj()
```



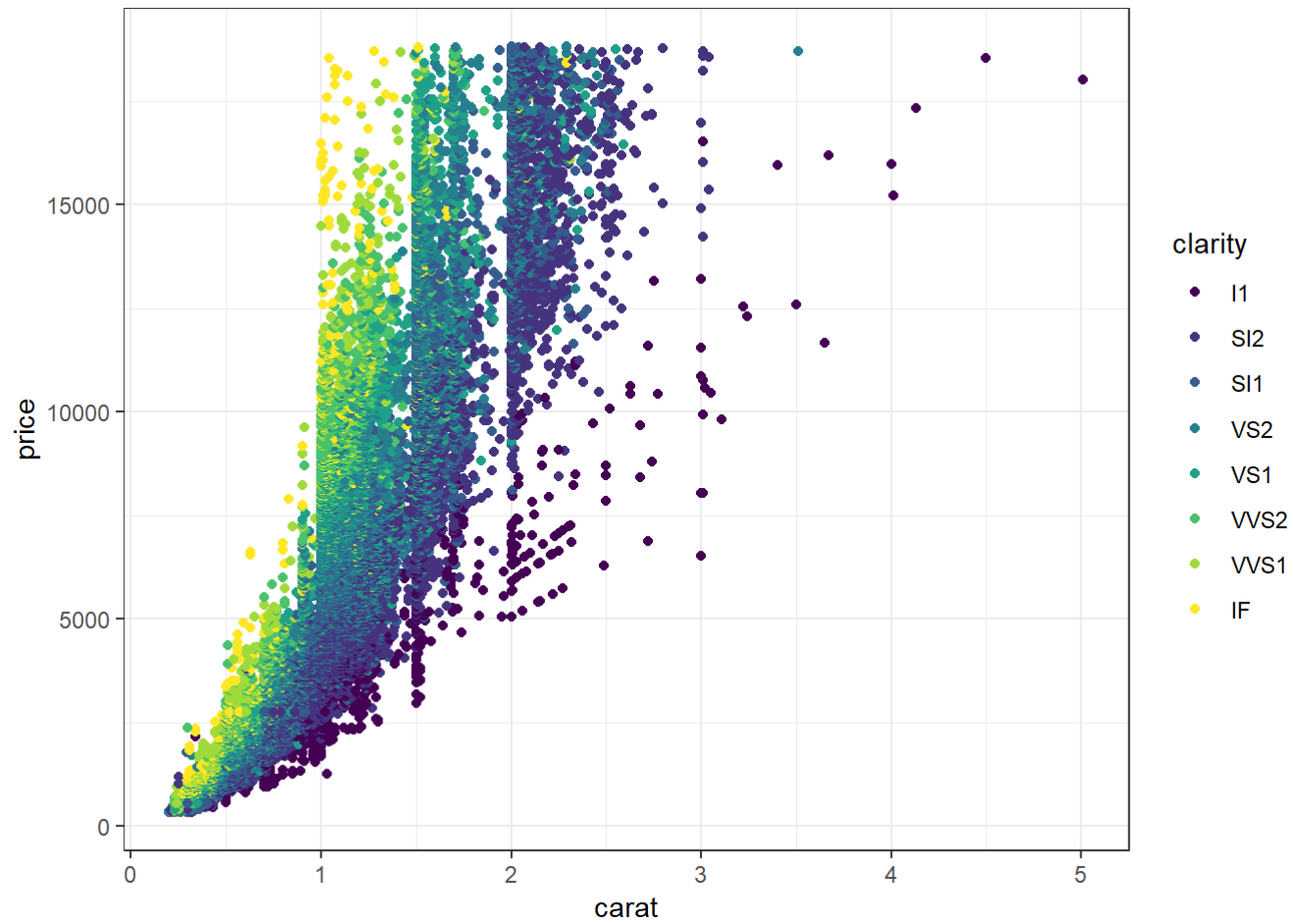
```
g1 <- ggplot(data=diamonds,  
             aes(x=carat, y=price, colour=clarity))  
g2 <- geom_point()  
g3 <- theme_wsj()  
g1 + g2 + g3
```



```
plot(diamonds$carat, diamonds$price)
```



```
g1 + g2 + theme_bw()
```



DF



```
##      name sex age grade absence bloodtype height weight
## 1 김길동 남자 23   3     유         0 165.3  68.2
## 2 이미린 여자 22   2     무         AB 170.1  53.0
## 3 홍길동 남자 24   4     무         B 175.0  80.1
## 4 김철수 남자 23   3     무         AB 182.1  85.7
## 5 손세수 여자 20   1     유         A 168.0  49.5
## 6 박미희 여자 21   2     무         0 162.0  52.0
## 7 강수친 여자 22   1     무         0 155.2  45.3
## 8 이희수 여자 23   1     무         A 176.9  55.0
## 9 이철린 남자 23   3     무         B 178.5  64.2
## 10 방희철 남자 22   2     무         B 176.1  61.3
## 11 박수호 남자 24   4     유         0 167.1  62.0
## 12 임동민 남자 22   2     무         AB 180.0  75.8
## 13 김민수 남자 21   1     무         A 162.2  55.3
## 14 이희진 여자 23   3     무         0 176.1  53.1
## 15 김미진 여자 22   2     무         B 158.2  45.2
## 16 김동수 남자 24   4     유         B 168.6  70.2
## 17 여수근 남자 21   1     무         A 169.2  62.2
```

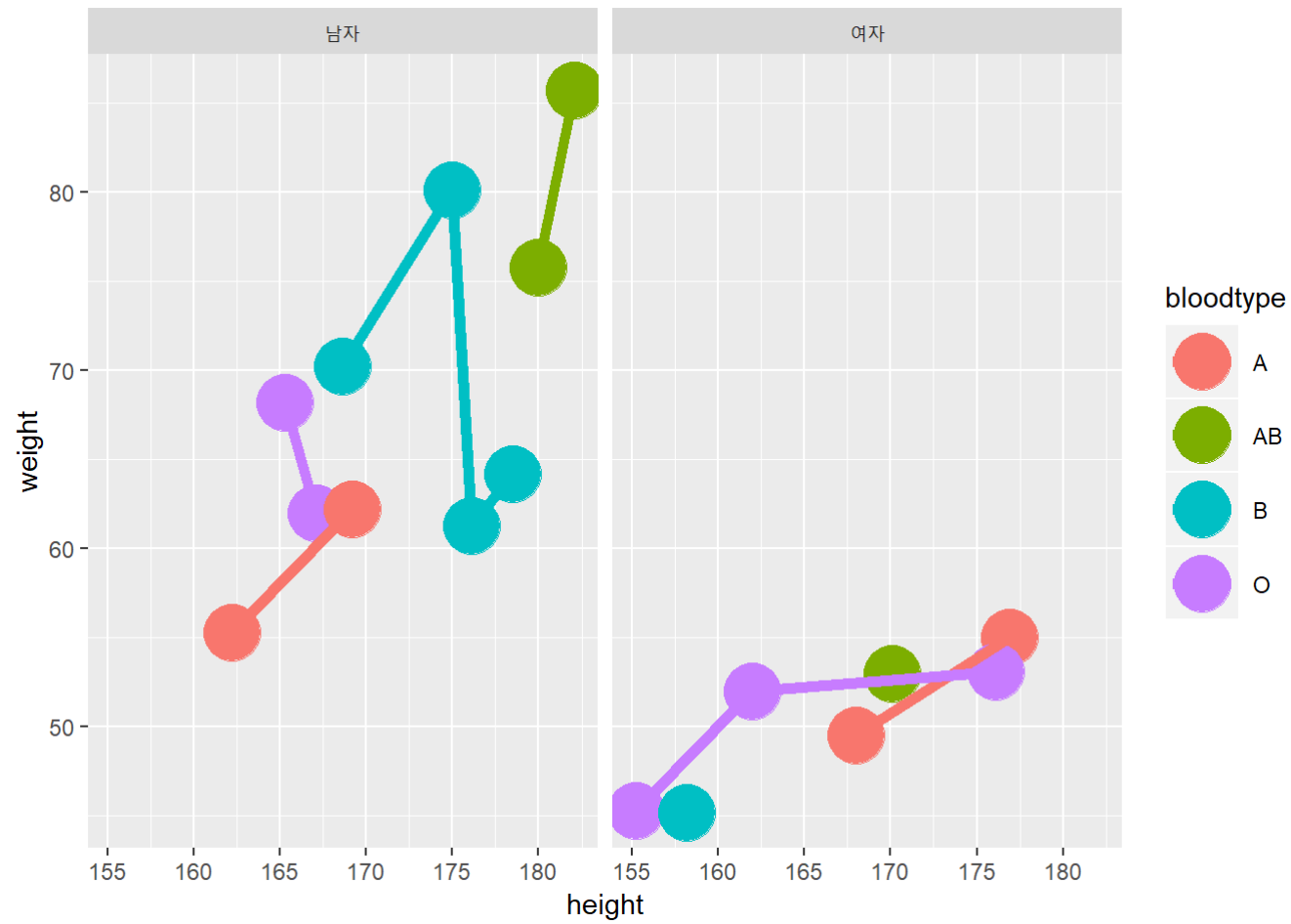
```
g1 <- ggplot(DF)
```

## ggplot2, point, line, size

x축 weight, y축 height, 색구분 : bloodtype

- (1) 점으로 해보기
  - (2) 선으로 그려보기
  - (3) 점의 사이즈를 키워보기
  - (4) 성별로 나눠서 그래프를 보기
- facet\_grid() : facet\_grid(.~sex)

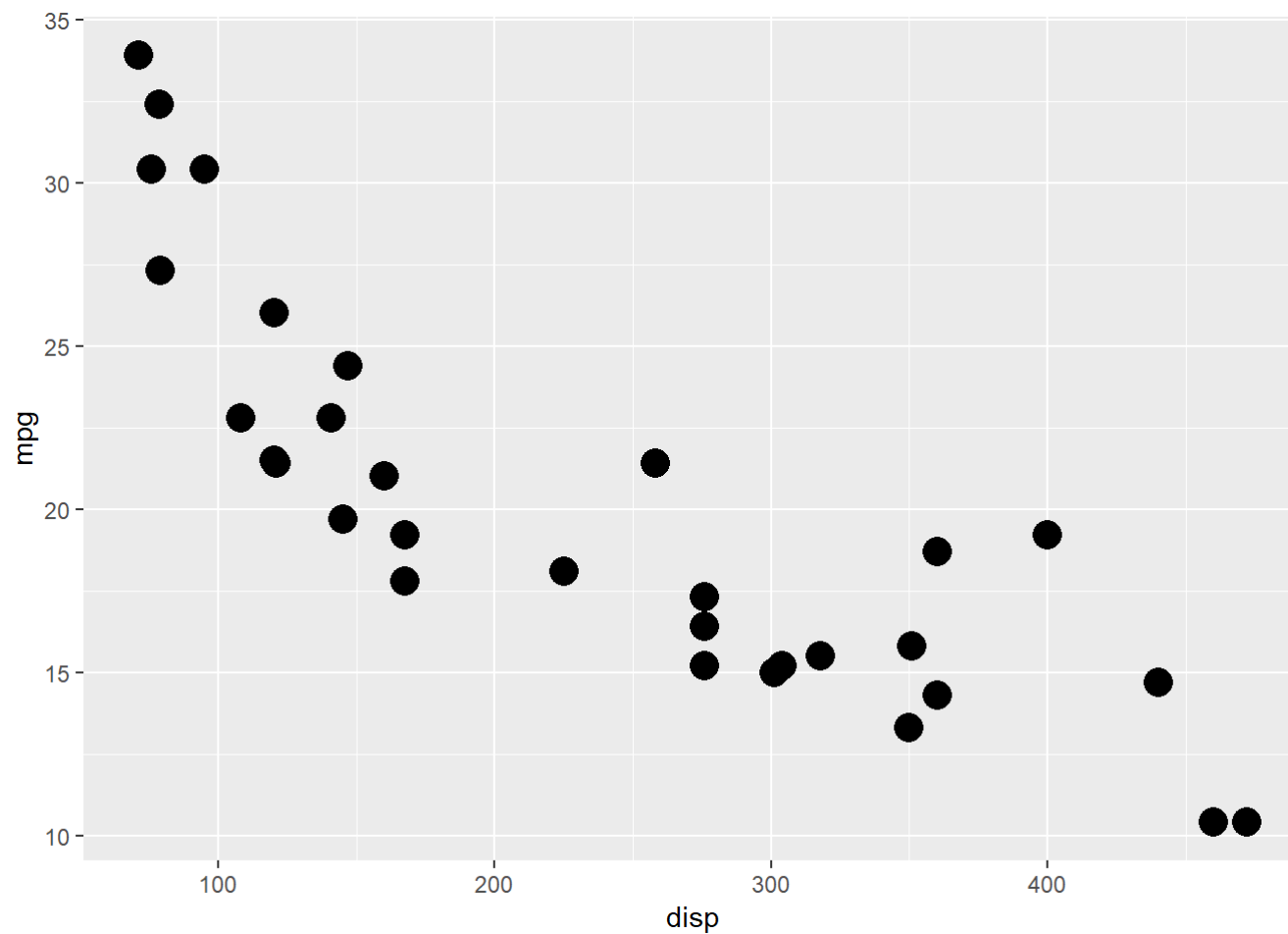
```
g1 <- ggplot(DF, aes(x=height, y=weight, colour=bloodtype))
g2_p <- geom_point(size=10)
g2_l <- geom_line(size=2)
g3 <- facet_grid(.~sex)
g1 + g2_p + g2_l + g3
```



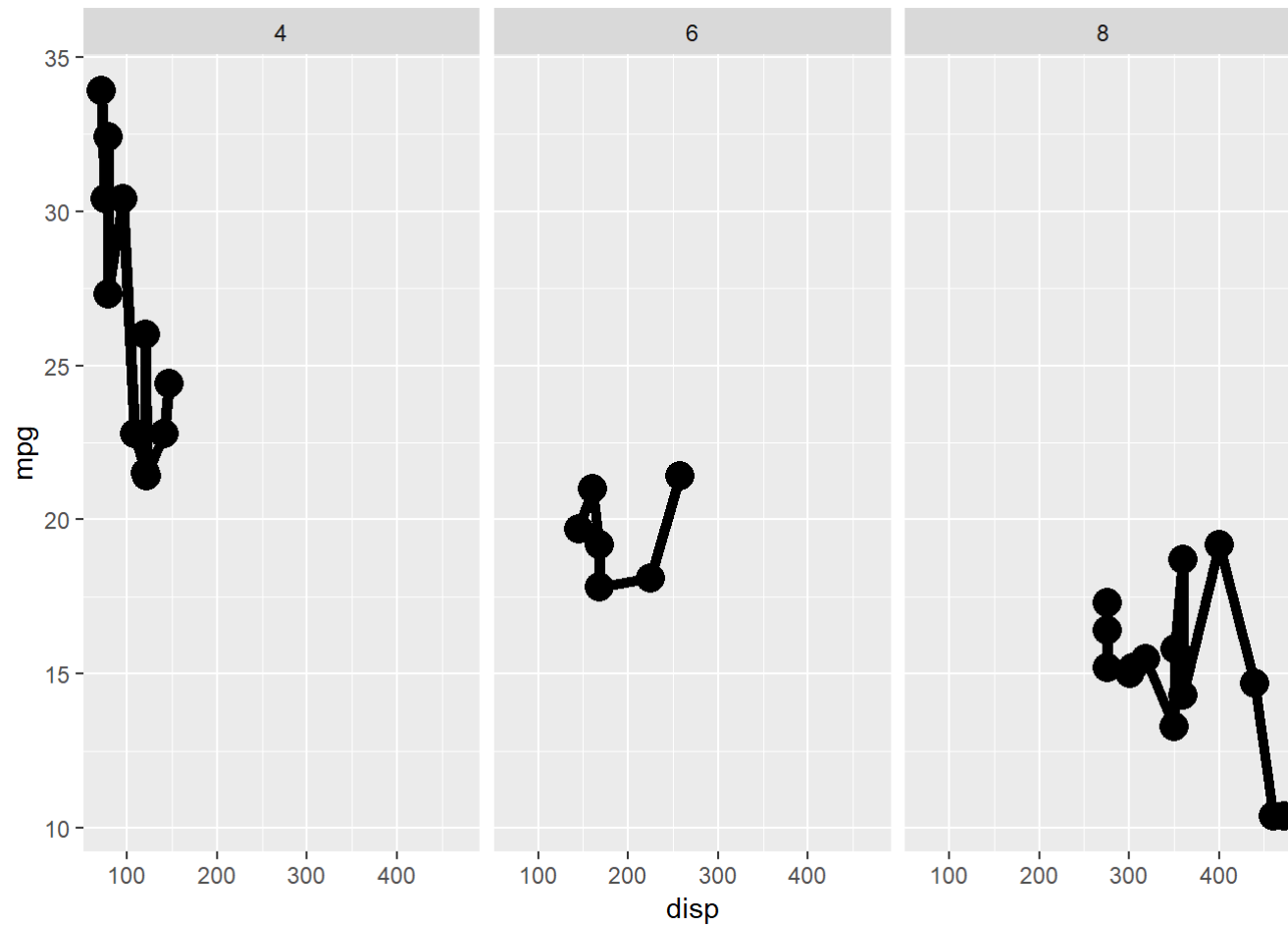
mtcars

##	mpg	cyl	disp	hp	drat	wt	qsec	vs	am	gear	carb
## Mazda RX4	21.0	6	160.0	110	3.90	2.620	16.46	0	1	4	4
## Mazda RX4 Wag	21.0	6	160.0	110	3.90	2.875	17.02	0	1	4	4
## Datsun 710	22.8	4	108.0	93	3.85	2.320	18.61	1	1	4	1
## Hornet 4 Drive	21.4	6	258.0	110	3.08	3.215	19.44	1	0	3	1
## Hornet Sportabout	18.7	8	360.0	175	3.15	3.440	17.02	0	0	3	2
## Valiant	18.1	6	225.0	105	2.76	3.460	20.22	1	0	3	1
## Duster 360	14.3	8	360.0	245	3.21	3.570	15.84	0	0	3	4
## Merc 240D	24.4	4	146.7	62	3.69	3.190	20.00	1	0	4	2
## Merc 230	22.8	4	140.8	95	3.92	3.150	22.90	1	0	4	2
## Merc 280	19.2	6	167.6	123	3.92	3.440	18.30	1	0	4	4
## Merc 280C	17.8	6	167.6	123	3.92	3.440	18.90	1	0	4	4
## Merc 450SE	16.4	8	275.8	180	3.07	4.070	17.40	0	0	3	3
## Merc 450SL	17.3	8	275.8	180	3.07	3.730	17.60	0	0	3	3
## Merc 450SLC	15.2	8	275.8	180	3.07	3.780	18.00	0	0	3	3
## Cadillac Fleetwood	10.4	8	472.0	205	2.93	5.250	17.98	0	0	3	4
## Lincoln Continental	10.4	8	460.0	215	3.00	5.424	17.82	0	0	3	4
## Chrysler Imperial	14.7	8	440.0	230	3.23	5.345	17.42	0	0	3	4
## Fiat 128	32.4	4	78.7	66	4.08	2.200	19.47	1	1	4	1
## Honda Civic	30.4	4	75.7	52	4.93	1.615	18.52	1	1	4	2
## Toyota Corolla	33.9	4	71.1	65	4.22	1.835	19.90	1	1	4	1
## Toyota Corona	21.5	4	120.1	97	3.70	2.465	20.01	1	0	3	1
## Dodge Challenger	15.5	8	318.0	150	2.76	3.520	16.87	0	0	3	2
## AMC Javelin	15.2	8	304.0	150	3.15	3.435	17.30	0	0	3	2
## Camaro Z28	13.3	8	350.0	245	3.73	3.840	15.41	0	0	3	4
## Pontiac Firebird	19.2	8	400.0	175	3.08	3.845	17.05	0	0	3	2
## Fiat X1-9	27.3	4	79.0	66	4.08	1.935	18.90	1	1	4	1
## Porsche 914-2	26.0	4	120.3	91	4.43	2.140	16.70	0	1	5	2
## Lotus Europa	30.4	4	95.1	113	3.77	1.513	16.90	1	1	5	2
## Ford Pantera L	15.8	8	351.0	264	4.22	3.170	14.50	0	1	5	4
## Ferrari Dino	19.7	6	145.0	175	3.62	2.770	15.50	0	1	5	6
## Maserati Bora	15.0	8	301.0	335	3.54	3.570	14.60	0	1	5	8
## Volvo 142E	21.4	4	121.0	109	4.11	2.780	18.60	1	1	4	2

```
g1 <- ggplot(mtcars, aes(displacement, mpg))  
g2_p <- geom_point(size=5)  
#g2_l <- geom_line(size=2)  
g3 <- facet_grid(.~cyl)  
#g  
g1 + g2_p
```



```
g1 + g2_p + g2_l + g3
```



barplot를 ggplot로

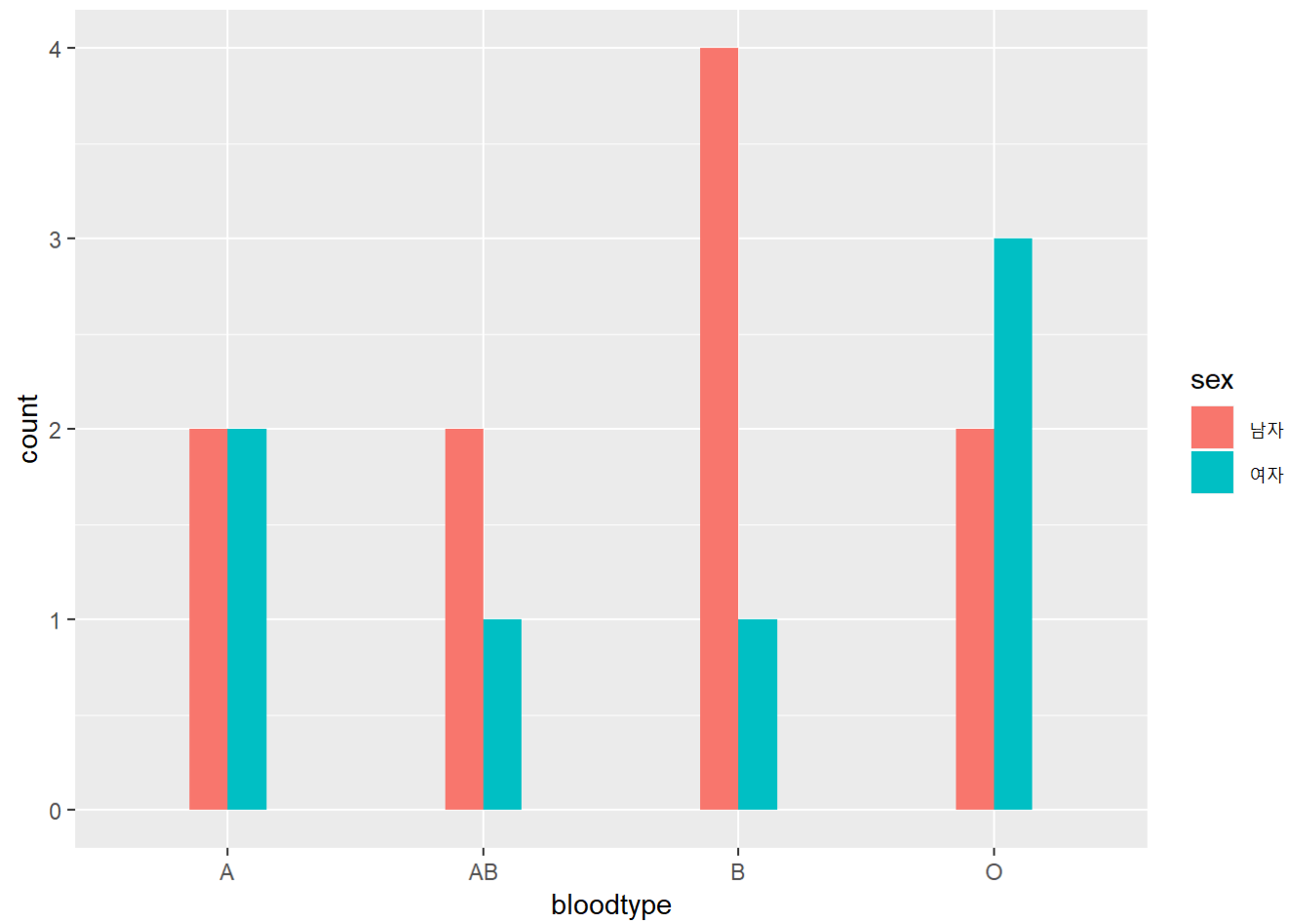
DF

##		name	sex	age	grade	absence	bloodtype	height	weight
## 1		김길동	남자	23	3	유	0	165.3	68.2
## 2		이미린	여자	22	2	무	AB	170.1	53.0
## 3		홍길동	남자	24	4	무	B	175.0	80.1
## 4		김철수	남자	23	3	무	AB	182.1	85.7
## 5		손세수	여자	20	1	유	A	168.0	49.5
## 6		박미희	여자	21	2	무	0	162.0	52.0
## 7		강수진	여자	22	1	무	0	155.2	45.3
## 8		이희수	여자	23	1	무	A	176.9	55.0
## 9		이철린	남자	23	3	무	B	178.5	64.2
## 10		방희철	남자	22	2	무	B	176.1	61.3
## 11		박수호	남자	24	4	유	0	167.1	62.0
## 12		임동민	남자	22	2	무	AB	180.0	75.8
## 13		김민수	남자	21	1	무	A	162.2	55.3
## 14		이희진	여자	23	3	무	0	176.1	53.1
## 15		김미진	여자	22	2	무	B	158.2	45.2
## 16		김동수	남자	24	4	유	B	168.6	70.2
## 17		여수근	남자	21	1	무	A	169.2	62.2

```

g1 <- ggplot(DF, aes(x=bloodtype, fill=sex))
#g2_p <- geom_point(size=5)
#g2_l <- geom_line(size=2)
g2_b <- geom_bar(position = "dodge", width=0.3)
#g3 <- facet_grid(.~cyl)
#g
g1 + g2_b

```



```
table(bloodtype,sex)
```

```
##      sex
## bloodtype 남자 여자
##      A      2    2
##      AB      2    1
##      B       4    1
##      O       2    3
```