

# WK02 Data Exploration

## 1 데이터 탐색

### 1.1 참고 사이트:

- RStudio Cheat Sheets (<https://rstudio.com/resources/cheatsheets/> (<https://rstudio.com/resources/cheatsheets/>)): 최신 치트시트
- Data Visualization Cheat Sheet (<https://github.com/rstudio/cheatsheets/raw/master/data-visualization-2.1.pdf> (<https://github.com/rstudio/cheatsheets/raw/master/data-visualization-2.1.pdf>))
- Data Transformation Cheat Sheet (<https://github.com/rstudio/cheatsheets/raw/master/data-transformation.pdf> (<https://github.com/rstudio/cheatsheets/raw/master/data-transformation.pdf>))
- ggplot2 사이트 (<https://docs.ggplot2.org/current/> (<https://docs.ggplot2.org/current/>))

### 1.2 htwtbd 자료설명

- htwtbd00.csv: 2021년 온라인으로 수집한 연예인 신체계측자료
  - n = 84. 여자 42명(배우 40, 가수 1, 개그맨 1), 남자 42명(배우 40, 가수 1, 개그맨 1)

변수	설명
name	이름
gnd	성별{F, M}. 이진 판별분석시 타겟
byr	출생년도
ht	키(cm). 회귀분석시 타겟
wt	몸무게(kg)
bd	혈액형{A,AB,B,O}
a	분야{actor, singer, comedian}

- Model Lookup (<https://topepo.github.io/caret/available-models.html> (<https://topepo.github.io/caret/available-models.html>))
- `install.packages("caret", dependencies=c("Depends", "Suggests"))`

### 1.3 패키지

```
library(tidyverse)
```

```
## -- Attaching packages ----- tidyverse 1.3.0 --
```

```
## √ ggplot2 3.3.3      √ purrr   0.3.4
## √ tibble  3.1.0      √ dplyr   1.0.5
## √ tidyr   1.1.3      √ stringr 1.4.0
## √ readr   1.4.0      √ forcats 0.5.1
```

```
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()    masks stats::lag()
```

```
library(gridExtra) # ggplot 객체를 한 페이지에 표시. grid.arrange(..., nrow, ncol)
```

```
##  
## Attaching package: 'gridExtra'
```

```
## The following object is masked from 'package:dplyr':  
##  
##      combine
```

```
library(scales) # 시각화 축조정 scale_x_xxx
```

```
##  
## Attaching package: 'scales'
```

```
## The following object is masked from 'package:purrr':  
##  
##      discard
```

```
## The following object is masked from 'package:readr':  
##  
##      col_factor
```

```
library(skimr) # 기초통계량 + 결측정보  
library(naniar) # 결측 정보
```

```
##  
## Attaching package: 'naniar'
```

```
## The following object is masked from 'package:skimr':  
##  
##      n_complete
```

```
library(corrplot)
```

```
## corrplot 0.84 loaded
```

## 1.4 읽기

```
# as.data.frame으로 안바꾸면 caret vs tidyverse 호환문제 때문에 경고 발생  
# as.data.frame해도 문자변수를 factor화 하지 않음  
DF <- as.data.frame(read_csv('D:/Github/Statics/DataMining/0321/htwtbd00.csv'))
```

```
##
## -- Column specification -----
## cols(
##   name = col_character(),
##   gnd = col_character(),
##   byr = col_double(),
##   ht = col_double(),
##   wt = col_double(),
##   bd = col_character(),
##   a = col_character(),
##   ftln = col_double()
## )
```

```
head(DF)
```

```
##   name gnd  byr  ht wt bd    a ftln
## 1 강소라  F 1990 168 NA  A actor  NA
## 2 김고은  F 1991 167 NA  B actor  NA
## 3 김민희  F 1982 170 49  A actor  240
## 4 김아중  F 1982 170 48  A actor  NA
## 5 김태리  F 1990 166 46  B actor  NA
## 6 김태희  F 1980 165 45  0 actor  NA
```

```
dim(DF)
```

```
## [1] 84  8
```

```
str(DF) # (Old) sapply(DF, class)
```

```
## 'data.frame':   84 obs. of  8 variables:
## $ name: chr  "강소라" "김고은" "김민희" "김아중" ...
## $ gnd : chr  "F" "F" "F" "F" ...
## $ byr : num  1990 1991 1982 1982 1990 ...
## $ ht : num  168 167 170 170 166 165 170 168 168 164 ...
## $ wt : num  NA NA 49 48 46 45 NA 45 48 NA ...
## $ bd : chr  "A" "B" "A" "A" ...
## $ a : chr  "actor" "actor" "actor" "actor" ...
## $ ftln: num  NA NA 240 NA NA NA NA NA NA ...
## - attr(*, "spec")=
## .. cols(
## ..   name = col_character(),
## ..   gnd = col_character(),
## ..   byr = col_double(),
## ..   ht = col_double(),
## ..   wt = col_double(),
## ..   bd = col_character(),
## ..   a = col_character(),
## ..   ftln = col_double()
## .. )
```

## 1.5 전처리

- age: 나이계산
- 이산형 변수처리
  - 문자변수(gnd, bd)를 factor화
  - {0,1}로 코딩된 가변수는 그대로 숫자형으로 사용. factor화해도 되지만 해석시 유의

```
# factor로 다 바꿀 것. lm, rpart, rf은 안해도 무방, gbm에서 오류남

DF <- mutate(DF,
              age = 2021-byr,
              gnd = factor(gnd),
              bd = factor(bd),
              a = factor(a))
sapply(DF, class)
```

```
##      name      gnd      byr      ht      wt      bd
## "character" "factor" "numeric" "numeric" "numeric" "factor"
##      a      ftln      age
## "factor" "numeric" "numeric"
```

## 1.6 기초통계량/결측파악

- `skimr::skim(data):summary()`와 결측정보. `group_by`와 연결. `pandas::describe()`와 유사
- `naniar::vis_miss(data)`: 변수별 결측비율 시각화
  - 주의: 출력물의 `Missing(%)`과 `Present(%)`는 완전 결측값 비율이 아니고, 전체 셀 중 결측의 비율임
  - `sum(complete.cases(DF))`: 완전 관측값 개수 반환
  - 원자료가 너무 크면 랜덤 추출(`sample_n`)해서 파악할 것
  - `DF %>% dplyr::sample_frac(size=0.1) %>% vis_miss()`
- `naniar::miss_var_summary(data)`: 변수별 결측비율 요약

```
skim(DF)
```

### Data summary

Name	DF
Number of rows	84
Number of columns	9
Column type frequency:	
character	1
factor	3
numeric	5
Group variables	
None	

### Variable type: character

skim_variable	n_missing	complete_rate	min	max	empty	n_unique	whitespace
name	0	1	2	3	0	84	0

### Variable type: factor

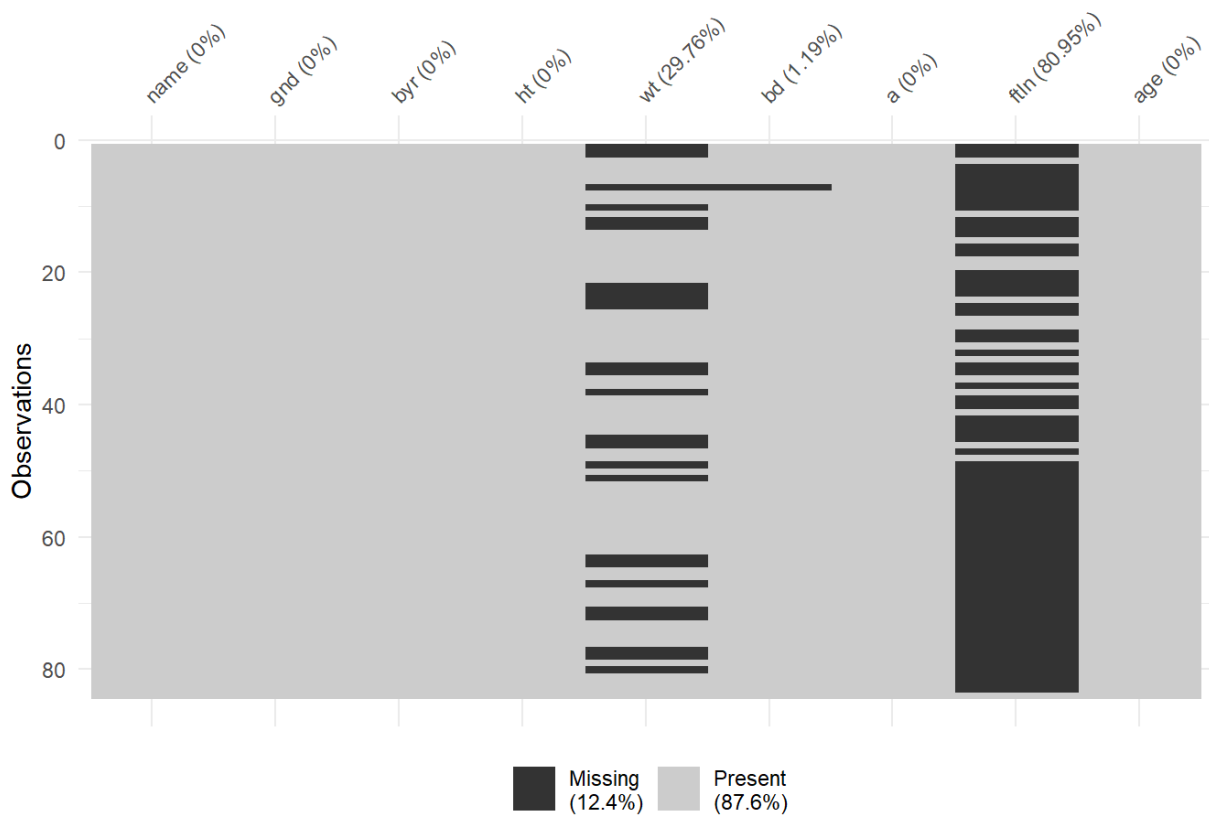
skim_variable	n_missing	complete_rate	ordered	n_unique	top_counts
gnd	0	1.00	FALSE	2	F: 42, M: 42
bd	1	0.99	FALSE	4	B: 26, A: 25, O: 21, AB: 11

skim_variable	n_missing	complete_rate	ordered	n_unique	top_counts
a	0	1.00	FALSE	3	act: 80, com: 2, sin: 2

Variable type: numeric

skim_variable	n_missing	complete_rate	mean	sd	p0	p25	p50	p75	p100	hist
byr	0	1.00	1982.65	7.40	1967	1978.00	1982	1989.00	1994	
ht	0	1.00	173.33	8.68	149	166.75	172	181.00	188	
wt	25	0.70	59.27	13.14	43	47.50	60	70.00	100	
ftln	68	0.19	245.31	23.06	215	235.00	240	246.25	310	
age	0	1.00	38.35	7.40	27	32.00	39	43.00	54	

# 변수별 결측비율, Missing=결측셀/전체셀, Present=비결측셀/전체셀  
 naniar::vis\_miss(DF)



naniar::miss\_var\_summary(DF)

```
## # A tibble: 9 x 3
##   variable n_miss pct_miss
##   <chr>     <int>    <dbl>
## 1 ftln      68      81.0
## 2 wt       25      29.8
## 3 bd        1       1.19
## 4 name      0        0
## 5 gnd       0        0
## 6 byr       0        0
## 7 ht       0        0
## 8 a        0        0
## 9 age      0        0
```

```
# 완전 관측값 비율 = 15%
# 회귀분석계통 분석방법을 그대로 적용하면 전체 자료의 15%만 사용하게 됨
sum(complete.cases(Df))/nrow(Df)*100 # prop_complete_case(Df)
```

```
## [1] 15.47619
```

## 1.7 탐색

### 1.7.1 단변량 탐색

- 연속형 변수의 탐색
  - 수치요약: 평균, 표준편차
  - 시각화: 히스토그램, density(커널분포추정), 상자그림, rug

```
summary(Df)
```

```
##      name      gnd      byr      ht      wt
## Length:84      F:42   Min.   :1967   Min.   :149.0   Min.   : 43.00
## Class :character M:42   1st Qu.:1978   1st Qu.:166.8   1st Qu.: 47.50
## Mode  :character      Median :1982   Median :172.0   Median : 60.00
##                      Mean   :1983   Mean   :173.3   Mean   : 59.27
##                      3rd Qu.:1989   3rd Qu.:181.0   3rd Qu.: 70.00
##                      Max.   :1994   Max.   :188.0   Max.   :100.00
##                      NA's    :25
##      bd      a      ftln      age
## A :25 actor :80   Min.   :215.0   Min.   :27.00
## AB :11 comedian: 2   1st Qu.:235.0   1st Qu.:32.00
## B :26 singer : 2   Median :240.0   Median :39.00
## 0 :21          Mean :245.3   Mean :38.35
## NA's: 1          3rd Qu.:246.2   3rd Qu.:43.00
##                      Max. :310.0   Max. :54.00
##                      NA's :68
```

```
# summarize_if(.tbl, .predicate:logical, .funs:list, ...)
# summarize_at(, tbl, .vars_vector, .fybs:list, ...)
summarize_if(Df, is.numeric, list(mn=mean, sd=sd), na.rm=TRUE)
```

```
##      byr_mn   ht_mn   wt_mn ftln_mn   age_mn   byr_sd   ht_sd   wt_sd
## 1 1982.655 173.3333 59.27119 245.3125 38.34524 7.398297 8.683641 13.13698
##      ftln_sd   age_sd
## 1 23.05564 7.398297
```

```
summarize_at(Df, c('ht', 'wt'), list(mn=mean, sd=sd), na.rm=TRUE)
```






```
##      ht_mn    wt_mn    ht_sd    wt_sd
## 1 173.3333 59.27119 8.683641 13.13698
```

```
DF%>% dplyr::select_if(is.numeric) %>% skim()
```

### Data summary

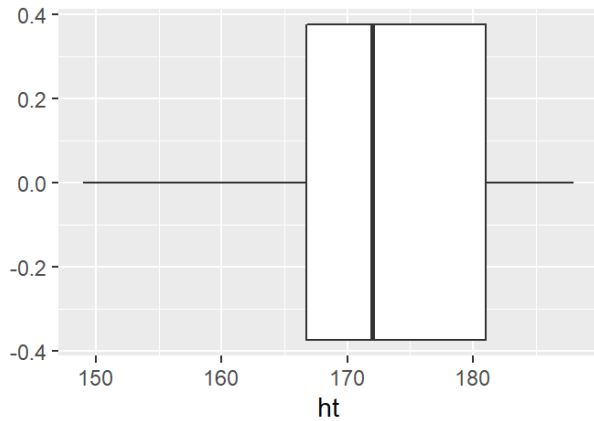
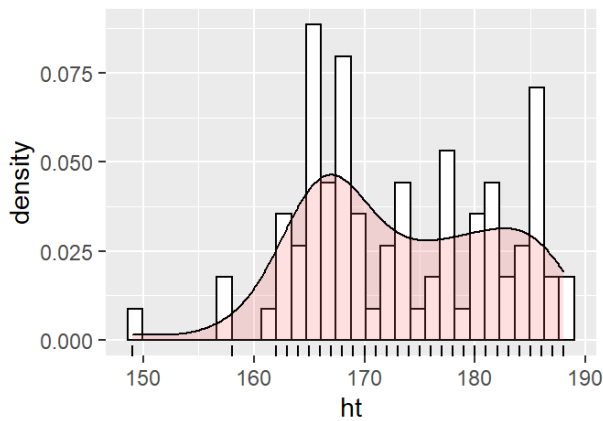
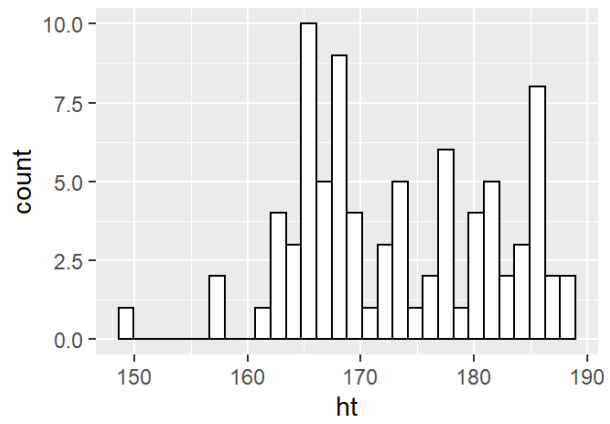
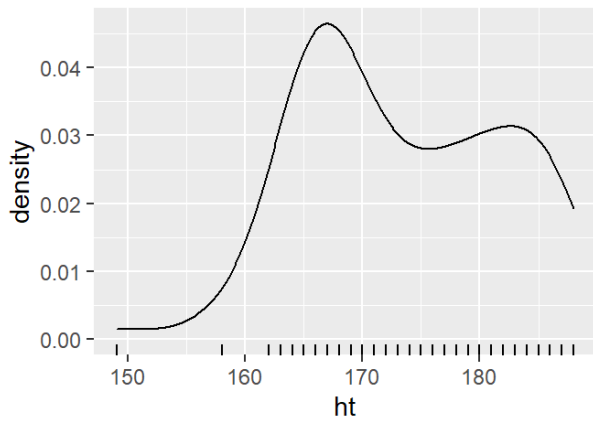
Name	Piped data
Number of rows	84
Number of columns	5
<hr/>	
Column type frequency:	
numeric	5
<hr/>	
Group variables	None

### Variable type: numeric

skim_variable	n_missing	complete_rate	mean	sd	p0	p25	p50	p75	p100	hist
byr	0	1.00	1982.65	7.40	1967	1978.00	1982	1989.00	1994	
ht	0	1.00	173.33	8.68	149	166.75	172	181.00	188	
wt	25	0.70	59.27	13.14	43	47.50	60	70.00	100	
ftln	68	0.19	245.31	23.06	215	235.00	240	246.25	310	
age	0	1.00	38.35	7.40	27	32.00	39	43.00	54	

```
g1 <- ggplot(DF, aes(x=ht)) + geom_density() + geom_rug()
g2 <- ggplot(DF, aes(x=ht)) + geom_histogram(color='black', fill='white')
g3 <- ggplot(DF, aes(x=ht)) + geom_histogram(aes(y=..density..), color='black', fill='white') + geom_density(alpha=0.2, fill='#FF6666') + geom_rug()
g4 <- ggplot(DF, aes(x=ht)) + geom_boxplot()
grid.arrange(g1, g2, g3, g4, nrow=2, ncol=2)
```

```
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```



- 이산형 변수의 탐색
  - 수치요약: 빈도, 상대빈도
  - 시각화: 막대그래프(barplot)

```
DF %>% dplyr::select_if(is.factor) %>% skim()
```

#### Data summary

Name	Piped data
Number of rows	84
Number of columns	3
Column type frequency:	
factor	3
Group variables	
None	

#### Variable type: factor

skim_variable	n_missing	complete_rate	ordered	n_unique	top_counts
gnd	0	1.00	FALSE	2	F: 42, M: 42
bd	1	0.99	FALSE	4	B: 26, A: 25, O: 21, AB: 11
a	0	1.00	FALSE	3	act: 80, com: 2, sin: 2

```
table(DF$gnd)
```



```
##
##  F  M
## 42 42
```

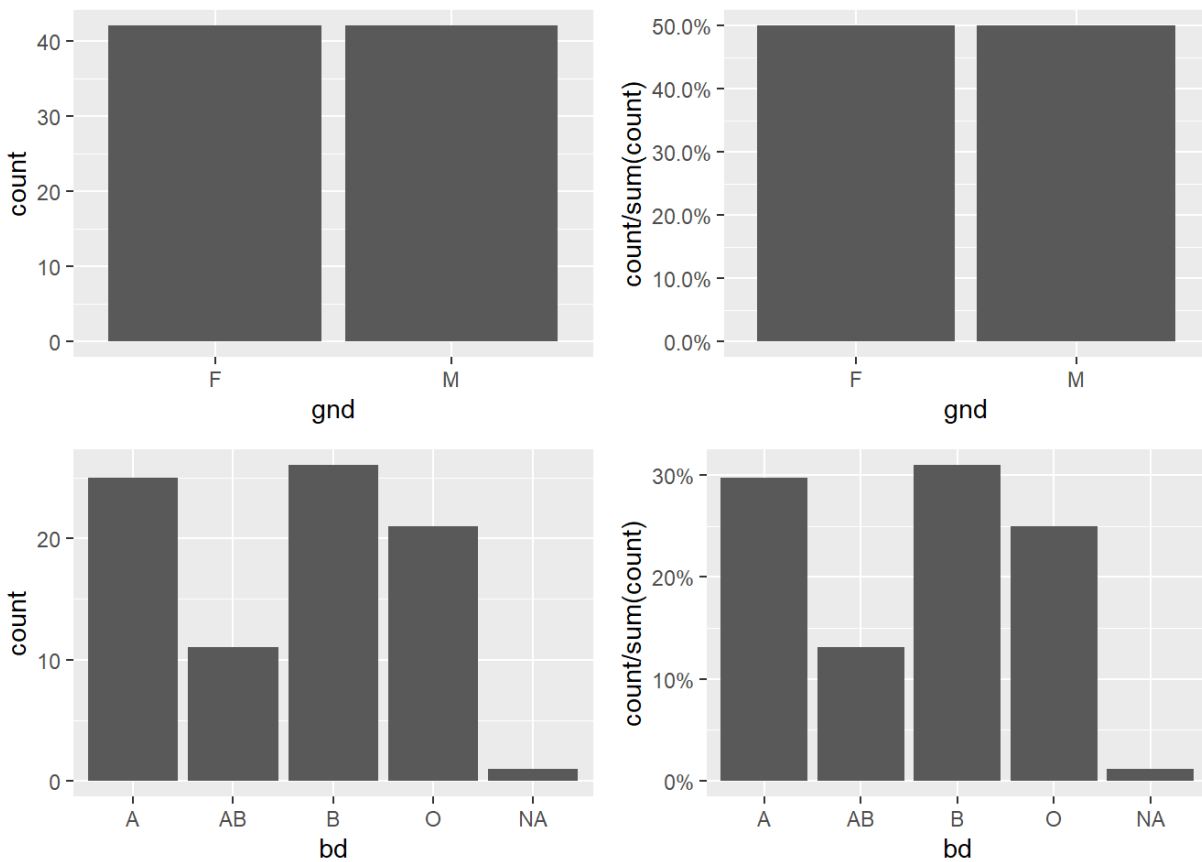
```
xtabs(~bd, data=DF)
```

```
## bd
##  A AB  B  O
## 25 11 26 21
```

```
xtabs(~a, data=DF)
```

```
## a
##   actor comedian   singer
##    80         2         2
```

```
g1 <- ggplot(DF, aes(x=gnd)) + geom_bar()
g2 <- ggplot(DF, aes(x=gnd)) + geom_bar(aes(y=..count../sum(..count..))) + scale_y_continuous(labels=percent)
g3 <- ggplot(DF, aes(x=bd)) + geom_bar()
g4 <- ggplot(DF, aes(x=bd)) + geom_bar(aes(y=..count../sum(..count..))) + scale_y_continuous(labels=percent)
grid.arrange(g1, g2, g3, g4, nrow=2, ncol=2)
```



## 1.7.2 이변량 탐색

- 연속 ~ 이산

```
DF %>% group_by(gnd) %>% dplyr::select_if(is.numeric) %>% skim()
```

Data summary

Name	Piped data
Number of rows	84
Number of columns	6
Column type frequency:	
numeric	5
Group variables	
gnd	

Variable type: numeric

skim_variable	gnd	n_missing	complete_rate	mean	sd	p0	p25	p50	p75	p100	hist
byr	F	0	1.00	1985.10	6.22	1970	1981.25	1985.5	1990.00	1994	
byr	M	0	1.00	1980.21	7.74	1967	1975.25	1979.5	1987.50	1994	
ht	F	0	1.00	166.12	4.33	149	164.25	166.5	168.00	173	
ht	M	0	1.00	180.55	5.22	168	177.00	181.0	185.00	188	
wt	F	13	0.69	47.55	3.55	43	45.00	47.0	49.00	60	
wt	M	12	0.71	70.60	7.93	55	65.50	70.0	72.75	100	
ftln	F	29	0.31	236.15	9.61	215	235.00	240.0	240.00	250	
ftln	M	39	0.07	285.00	22.91	265	272.50	280.0	295.00	310	
age	F	0	1.00	35.90	6.22	27	31.00	35.5	39.75	51	
age	M	0	1.00	40.79	7.74	27	33.50	41.5	45.75	54	

```
DF %>%
  group_by(gnd) %>%
  summarize_at(c('ht', 'wt'), list(mn=mean, sd=sd), na.rm=TRUE)
```

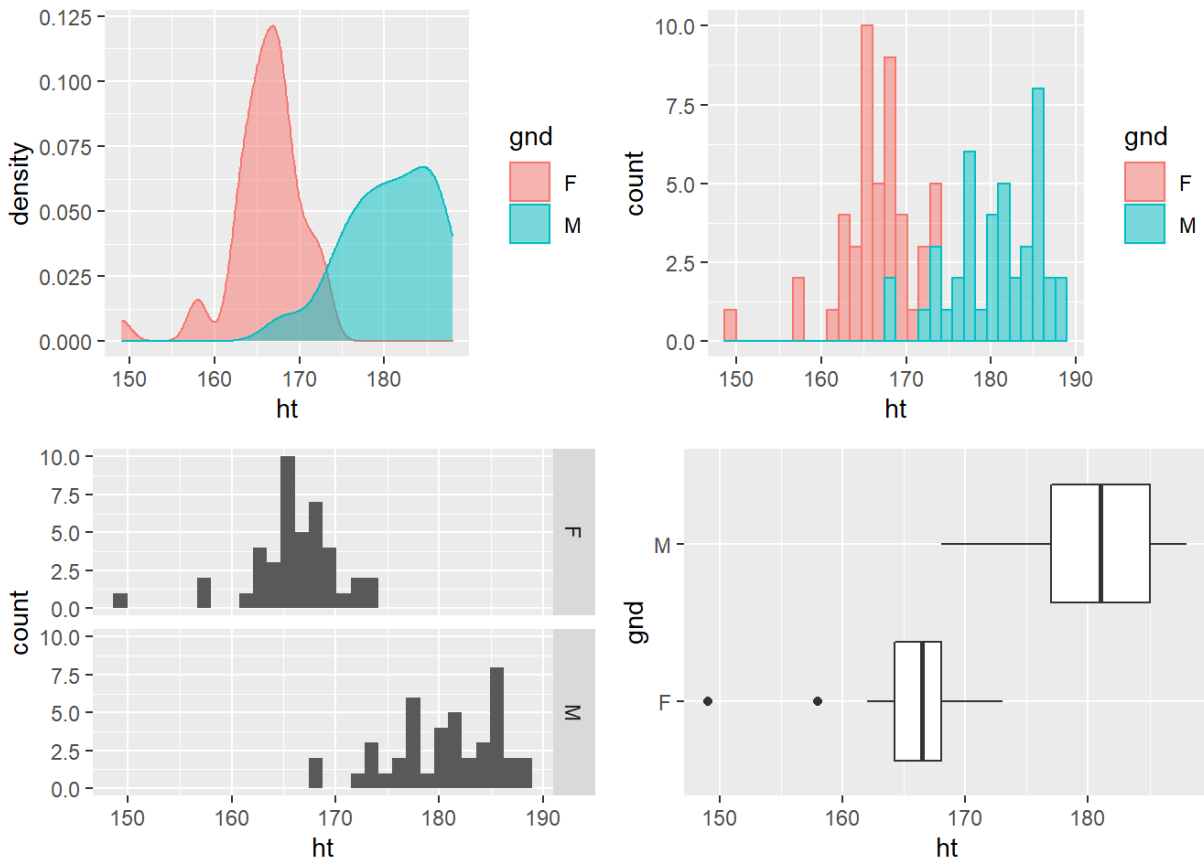
```
## # A tibble: 2 x 5
##   gnd   ht_mn wt_mn ht_sd wt_sd
##   <fct> <dbl> <dbl> <dbl> <dbl>
## 1 F      166.  47.6  4.33  3.55
## 2 M      181.  70.6  5.22  7.93
```

```
DF %>%
  group_by(gnd) %>%
  summarize_if(is.numeric, list(mn = mean, sd=sd), na.rm=TRUE)
```

```
## # A tibble: 2 x 11
##   gnd   byr_mn ht_mn wt_mn ftln_mn age_mn byr_sd ht_sd wt_sd ftln_sd age_sd
##   <fct> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl>
## 1 F      1985.  166.  47.6   236.   35.9   6.22  4.33  3.55   9.61   6.22
## 2 M      1980.  181.  70.6   285   40.8   7.74  5.22  7.93  22.9   7.74
```

```
g1 <- ggplot(DF, aes(x=ht, col=gnd, fill=gnd)) + geom_density(alpha=0.5)
g2 <- ggplot(DF, aes(x=ht, col=gnd, fill=gnd)) + geom_histogram(alpha=0.5)
g3 <- ggplot(DF, aes(x=ht)) + geom_histogram() + facet_grid(gnd~.)
g4 <- ggplot(DF, aes(x=gnd, y=ht)) + geom_boxplot() + coord_flip()
grid.arrange(g1, g2, g3, g4, nrow=2, ncol=2)
```

```
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```



```
t.test(ht~gnd, data=DF, var.equal=TRUE)
```

```
##
## Two Sample t-test
##
## data: ht by gnd
## t = -13.784, df = 82, p-value < 2.2e-16
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -16.51091 -12.34623
## sample estimates:
## mean in group F mean in group M
##      166.1190      180.5476
```

```
summary(aov(ht~bd, data=DF))
```

```
##           Df Sum Sq Mean Sq F value Pr(>F)
## bd         3      70    23.21   0.297  0.828
## Residuals  79   6178    78.20
## 1 observation deleted due to missingness
```

• 연속 ~ 연속

```
# cor(DF[,sapply(DF, is.numeric)], use='pairwise.complete.obs')
R <- cor(DF%>% select_if(is.numeric), use='pairwise.complete.obs')
R
```

```
##          byr          ht          wt          ftn          age
## byr    1.0000000 -0.2659908 -0.3483837 -0.3646281 -1.0000000
## ht     -0.2659908  1.0000000  0.8110682  0.8912221  0.2659908
## wt     -0.3483837  0.8110682  1.0000000  0.8381219  0.3483837
## ftn    -0.3646281  0.8912221  0.8381219  1.0000000  0.3646281
## age    -1.0000000  0.2659908  0.3483837  0.3646281  1.0000000
```

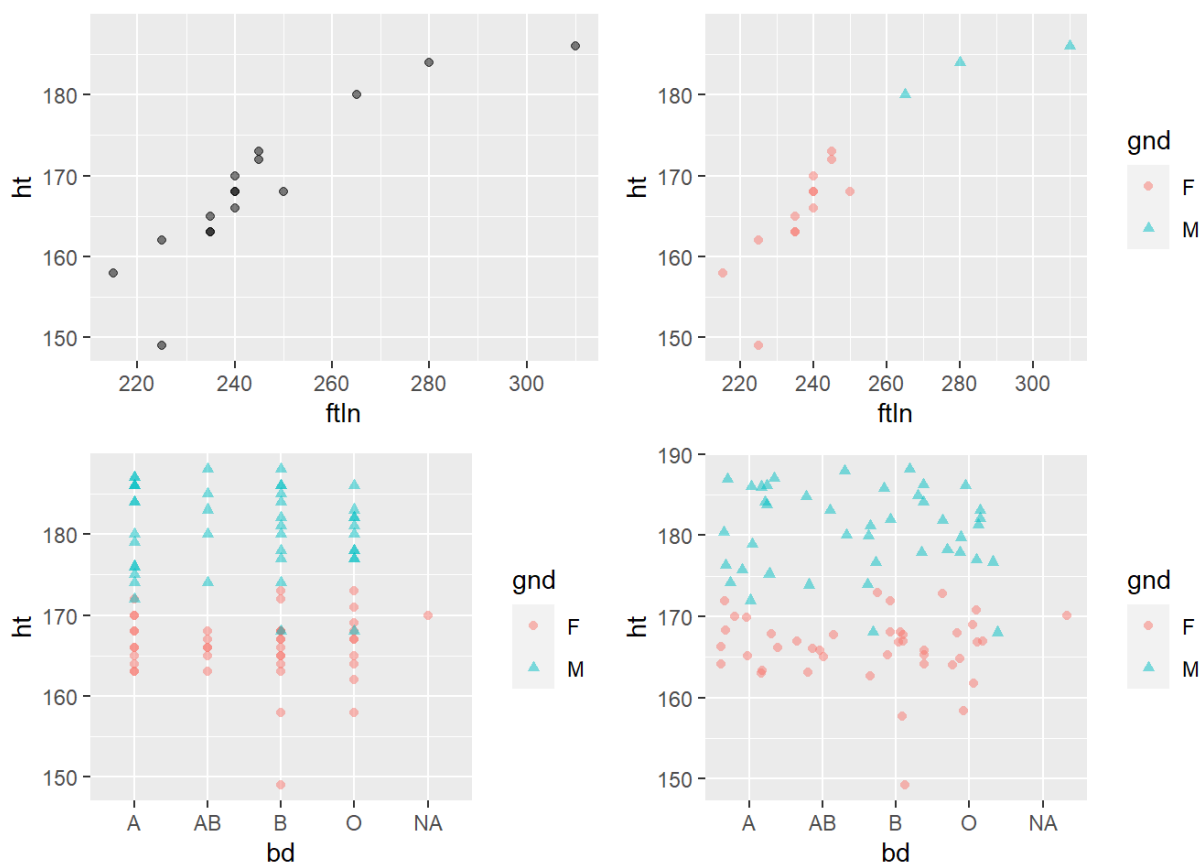
```
sort(R['ht',], decreasing=TRUE)
```

```
##          ht          ftn          wt          age          byr
## 1.0000000  0.8912221  0.8110682  0.2659908 -0.2659908
```

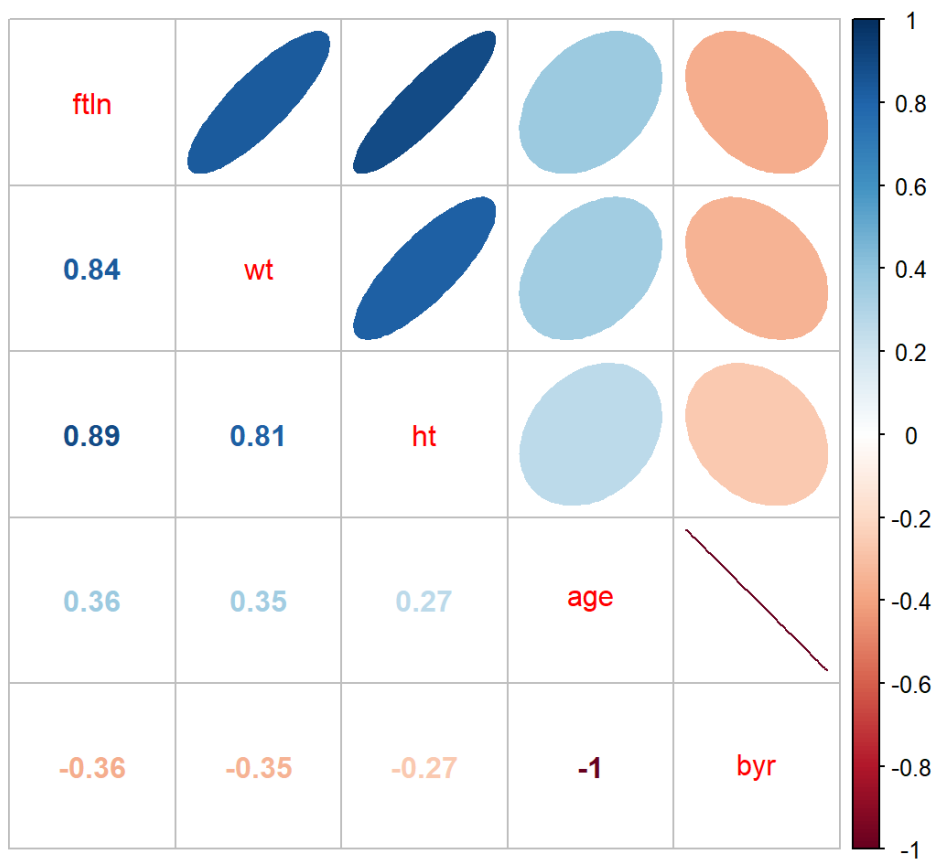
```
g1 <- ggplot(DF, aes(x=ftln, y=ht)) + geom_point(alpha=0.5)
g2 <- ggplot(DF, aes(x=ftln, y=ht, color=gnd, shape=gnd)) + geom_point(alpha=0.5)
g3 <- ggplot(DF, aes(x=bd, y=ht, color=gnd, shape=gnd)) + geom_point(alpha=0.5)
g4 <- ggplot(DF, aes(x=bd, y=ht, color=gnd, shape=gnd)) + geom_jitter(alpha=0.5)
grid.arrange(g1, g2, g3, g4, nrow=2, ncol=2)
```

```
## Warning: Removed 68 rows containing missing values (geom_point).
```

```
## Warning: Removed 68 rows containing missing values (geom_point).
```



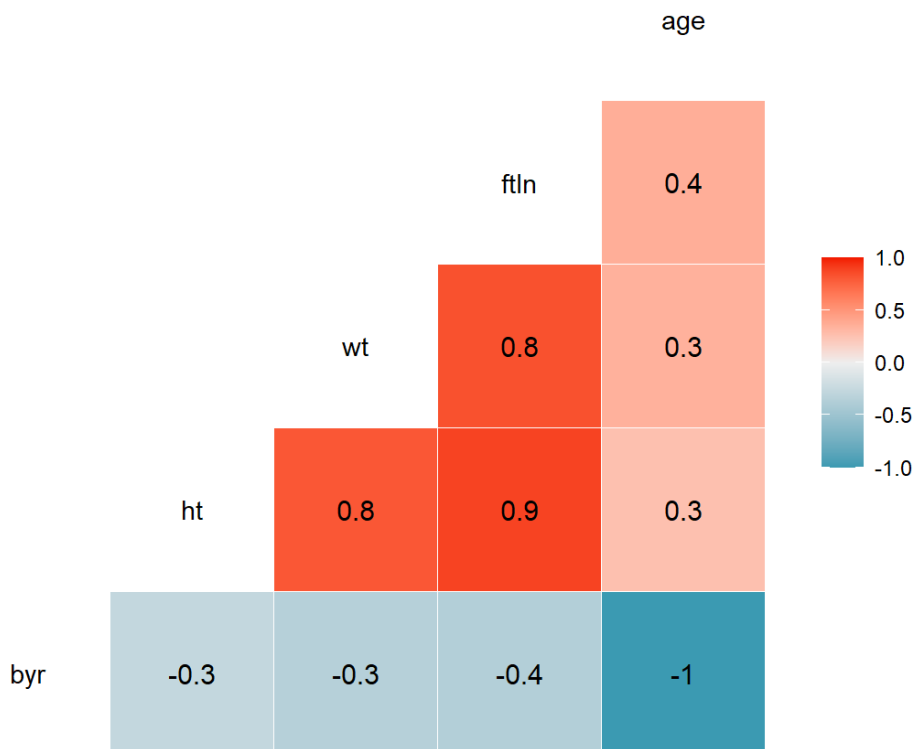
```
corrplot::corrplot.mixed(R, upper='ellipse', order='FPC')
```



```
library(GGally) # ggcorr, ggparis
```

```
## Registered S3 method overwritten by 'GGally':
##   method from
##   +.gg      ggplot2
```

```
ggcorr(DF %>% select_if(is.numeric), geom = 'tile', label=TRUE)
```



```
ggpairs(DF,
  columns=c('ht', 'ftln', 'wt'),
  lower=list(continuous=wrap('points', alpha=0.05, col='blue')),
  diag=list(continuous='barDiag')) # diag=list(continuous='densityDiag')
```

```
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```

```
## Warning in ggally_statistic(data = data, mapping = mapping, na.rm = na.rm, :
## Removed 68 rows containing missing values
```

```
## Warning in ggally_statistic(data = data, mapping = mapping, na.rm = na.rm, :
## Removed 25 rows containing missing values
```

```
## Warning: Removed 68 rows containing missing values (geom_point).
```

```
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```

```
## Warning: Removed 68 rows containing non-finite values (stat_bin).
```

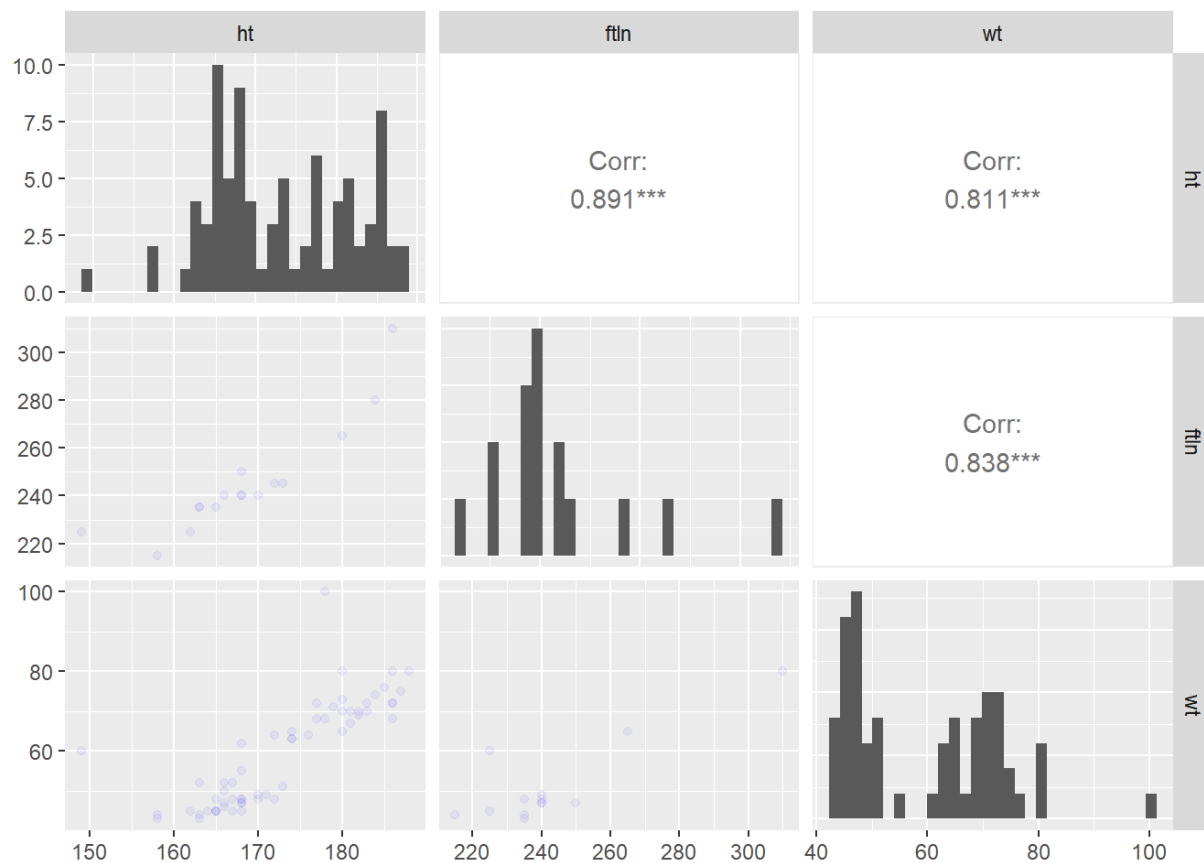
```
## Warning in ggally_statistic(data = data, mapping = mapping, na.rm = na.rm, :
## Removed 71 rows containing missing values
```

```
## Warning: Removed 25 rows containing missing values (geom_point).
```

```
## Warning: Removed 71 rows containing missing values (geom_point).
```

```
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```

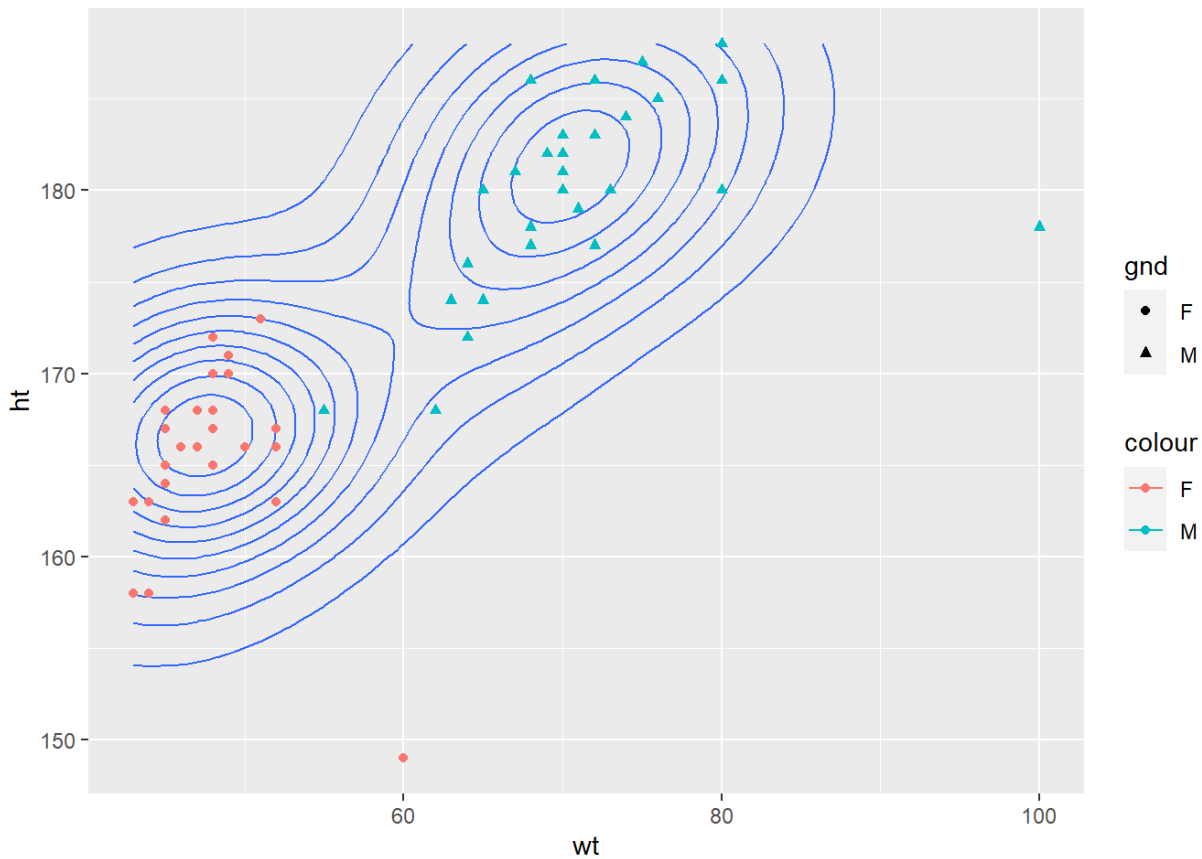
```
## Warning: Removed 25 rows containing non-finite values (stat_bin).
```



```
ggplot(DF, aes(x=wt, y=ht)) + geom_density2d() + geom_point(aes(col=gnd, shape=gnd))
```

```
## Warning: Removed 25 rows containing non-finite values (stat_density2d).
```

```
## Warning: Removed 25 rows containing missing values (geom_point).
```



- 이산 ~ 이산

```
g1 <- ggplot(DF, aes(x=bd, fill=gnd)) + geom_bar()
g2 <- ggplot(DF, aes(x=bd, fill=gnd)) + geom_bar(aes(y=..count../sum(..count..)))

# Or
tb <- table(DF$gnd, DF$bd)
tb <- xtabs(~bd+gnd, data=DF)
df <- data.frame(tb)
df
```

```
##   bd gnd Freq
## 1  A   F   11
## 2 AB   F    6
## 3  B   F   14
## 4  O   F   10
## 5  A   M   14
## 6 AB   M    5
## 7  B   M   12
## 8  O   M   11
```

```
g3 <- ggplot(df, aes(x=gnd, y=Freq)) + geom_bar(aes(fill=bd), stat='identity')

tb <- prop.table(xtabs(~gnd+bd, data=DF), 1)
tb
```

```
##   bd
## gnd      A      AB      B      O
##  F 0.2682927 0.1463415 0.3414634 0.2439024
##  M 0.3333333 0.1190476 0.2857143 0.2619048
```

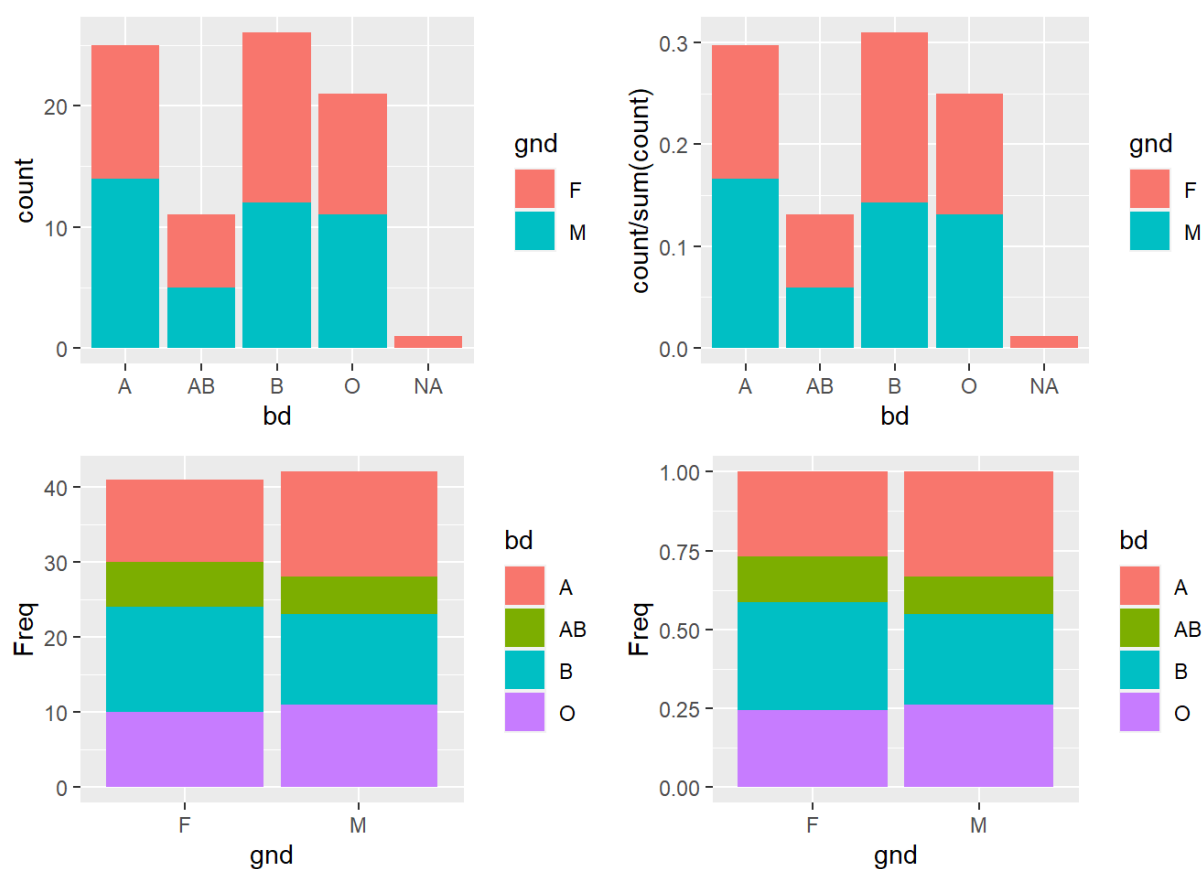


```
df <- data.frame(tb)
df
```

```
##   gnd bd      Freq
## 1  F  A 0.2682927
## 2  M  A 0.3333333
## 3  F AB 0.1463415
## 4  M AB 0.1190476
## 5  F  B 0.3414634
## 6  M  B 0.2857143
## 7  F  O 0.2439024
## 8  M  O 0.2619048
```

```
g4 <- ggplot(df, aes(x=gnd, y=Freq)) + geom_bar(aes(fill=bd), stat='identity')

grid.arrange(g1, g2, g3, g4, nrow=2, ncol=2)
```



```
chisq.test(xtabs(~gnd+bd, data=DF), correct=FALSE)
```

```
##
## Pearson's Chi-squared test
##
## data:  xtabs(~gnd + bd, data = DF)
## X-squared = 0.64042, df = 3, p-value = 0.8871
```