

cda 第一次作业

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- 第一题

```
library(dplyr)
description <- read.csv('GDS5037/description.csv')
##calculate the frequency and percentage of each category and plot pie chart
s <- description %>%group_by(ID)%>%summarise(Counts=n())
#if"%>%" errors,maybe the reason is the function can't be found
library(ggplot2)
type <- s$ID
nums <- s$Counts
df <- data.frame(type = type, nums = nums)

#2.2 pie_chart
p <- ggplot(data = df, mapping = aes(x = 'Content', y = nums, fill = type)) + geom_bar(stat = "identity")

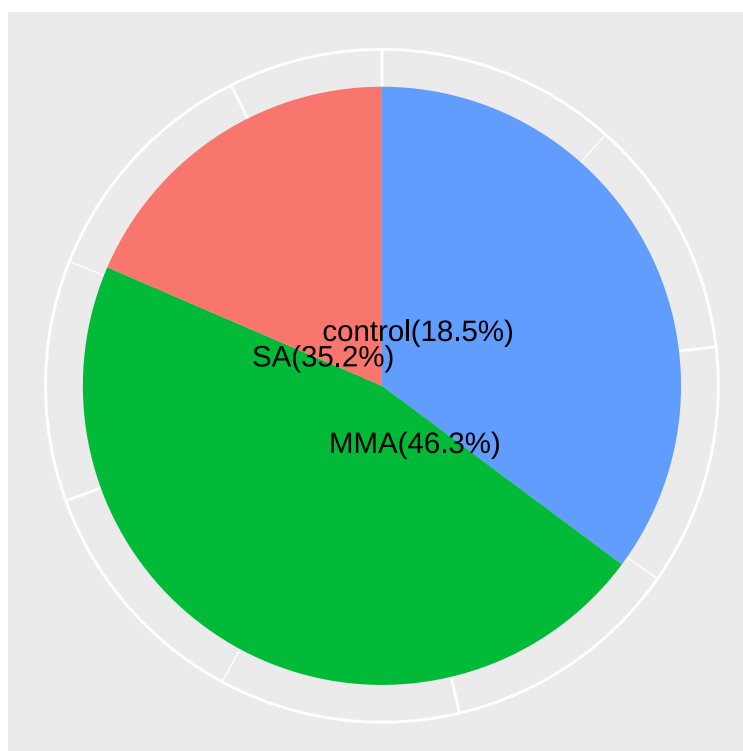
label_value <- paste('(', round(df$nums/sum(df$nums) * 100, 1), '%)', sep = '')
label_value

## [1] "(18.5%)" "(46.3%)" "(35.2%)"

label <- paste(df$type, label_value, sep = '')
label

## [1] "control(18.5%)" "MMA(46.3%)" "SA(35.2%)"
```

```
p+coord_polar(theta = 'y') + labs(x = '', y = '', title = '') + theme(axis.text = element.
```



#2.3 barchart about gender

```
male_sample <- 'GSM1068462,GSM1068463,GSM1068465,GSM1068466,GSM1068467,
GSM1068469,GSM1068470,GSM1068471,GSM1068475,GSM1068480,GSM1068484,
GSM1068485,GSM1068489,GSM1068497,GSM1068501,GSM1068504,GSM1068509,
GSM1068511,GSM1068515,GSM1068516,GSM1068519,GSM1068523,GSM1068525,
GSM1068526,GSM1068529,GSM1068530,GSM1068534,GSM1068536,GSM1068541,
GSM1068553,GSM1068554,GSM1068558,GSM1068559,GSM1068564'

female_sample = 'GSM1068458,GSM1068459,GSM1068460,GSM1068461,GSM1068464,
GSM1068468,GSM1068472,GSM1068473,GSM1068474,GSM1068476,GSM1068477,
GSM1068478,GSM1068479,GSM1068481,GSM1068482,GSM1068483,GSM1068486,
GSM1068487,GSM1068488,GSM1068490,GSM1068491,GSM1068492,GSM1068493,
GSM1068494,GSM1068495,GSM1068496,GSM1068498,GSM1068499,GSM1068500,
GSM1068502,GSM1068503,GSM1068505,GSM1068506,GSM1068507,GSM1068508,
GSM1068510,GSM1068512,GSM1068513,GSM1068514,GSM1068517,GSM1068518,
GSM1068520,GSM1068521,GSM1068522,GSM1068524,GSM1068527,GSM1068528,
GSM1068531,GSM1068532,GSM1068533,GSM1068535,GSM1068537,GSM1068538,
GSM1068539,GSM1068540,GSM1068542,GSM1068543,GSM1068544,GSM1068545,
```

```

GSM1068546,GSM1068547,GSM1068548,GSM1068549,GSM1068550,GSM1068551,
GSM1068552,GSM1068555,GSM1068556,GSM1068557,GSM1068560,GSM1068561,
GSM1068562,GSM1068563,GSM1068565'
male_sample <- strsplit(male_sample,',')[[1]]
male_sample <- gsub("\n","",male_sample)
female_sample <- strsplit(female_sample,',')[[1]]
female_sample <- gsub("\n","",female_sample)
countmale <- length(male_sample)
countfemale <- length(female_sample)
gender <- c(rep('male',countmale),rep('female',countfemale))
ID_REF <- c(male_sample,female_sample)
gendata <- data.frame(ID_REF,gender)
genderdata <- merge(description,gendata,by = 'ID_REF')
su <- genderdata %>%group_by(ID,gender)%>%summarise(Counts=n())
genderdata

```

```

##          ID_REF
## 1  GSM1068458
## 2  GSM1068459
## 3  GSM1068460
## 4  GSM1068461
## 5  GSM1068462
## 6  GSM1068463
## 7  GSM1068464
## 8  GSM1068465
## 9  GSM1068466
## 10 GSM1068467
## 11 GSM1068468
## 12 GSM1068469
## 13 GSM1068470
## 14 GSM1068471
## 15 GSM1068472
## 16 GSM1068473
## 17 GSM1068474
## 18 GSM1068475
## 19 GSM1068476
## 20 GSM1068477

```

21 GSM1068478
22 GSM1068479
23 GSM1068480
24 GSM1068481
25 GSM1068482
26 GSM1068483
27 GSM1068484
28 GSM1068485
29 GSM1068486
30 GSM1068487
31 GSM1068488
32 GSM1068489
33 GSM1068490
34 GSM1068491
35 GSM1068492
36 GSM1068493
37 GSM1068494
38 GSM1068495
39 GSM1068496
40 GSM1068497
41 GSM1068498
42 GSM1068499
43 GSM1068500
44 GSM1068501
45 GSM1068502
46 GSM1068503
47 GSM1068504
48 GSM1068505
49 GSM1068506
50 GSM1068507
51 GSM1068508
52 GSM1068509
53 GSM1068510
54 GSM1068511
55 GSM1068512
56 GSM1068513
57 GSM1068514

58 GSM1068515
59 GSM1068516
60 GSM1068517
61 GSM1068518
62 GSM1068519
63 GSM1068520
64 GSM1068521
65 GSM1068522
66 GSM1068523
67 GSM1068524
68 GSM1068525
69 GSM1068526
70 GSM1068527
71 GSM1068528
72 GSM1068529
73 GSM1068530
74 GSM1068531
75 GSM1068532
76 GSM1068533
77 GSM1068534
78 GSM1068535
79 GSM1068536
80 GSM1068537
81 GSM1068538
82 GSM1068539
83 GSM1068540
84 GSM1068541
85 GSM1068542
86 GSM1068543
87 GSM1068544
88 GSM1068545
89 GSM1068546
90 GSM1068547
91 GSM1068548
92 GSM1068549
93 GSM1068550
94 GSM1068551

```

## 95 GSM1068552
## 96 GSM1068553
## 97 GSM1068554
## 98 GSM1068555
## 99 GSM1068556
## 100 GSM1068557
## 101 GSM1068558
## 102 GSM1068559
## 103 GSM1068560
## 104 GSM1068561
## 105 GSM1068562
## 106 GSM1068563
## 107 GSM1068564
## 108 GSM1068565

```

```
##
```

```

## 1  #GSM1068458 = Value for GSM1068458: Bronchial Epithelial Cells control biological 1
## 2  #GSM1068459 = Value for GSM1068459: Bronchial Epithelial Cells control biological 1
## 3  #GSM1068460 = Value for GSM1068460: Bronchial Epithelial Cells control biological 1
## 4  #GSM1068461 = Value for GSM1068461: Bronchial Epithelial Cells control biological 1
## 5  #GSM1068462 = Value for GSM1068462: Bronchial Epithelial Cells control biological 1
## 6  #GSM1068463 = Value for GSM1068463: Bronchial Epithelial Cells control biological 1
## 7  #GSM1068464 = Value for GSM1068464: Bronchial Epithelial Cells control biological 1
## 8  #GSM1068465 = Value for GSM1068465: Bronchial Epithelial Cells control biological 1
## 9  #GSM1068466 = Value for GSM1068466: Bronchial Epithelial Cells control biological 1
## 10 #GSM1068467 = Value for GSM1068467: Bronchial Epithelial Cells control biological 1
## 11 #GSM1068468 = Value for GSM1068468: Bronchial Epithelial Cells control biological 1
## 12 #GSM1068469 = Value for GSM1068469: Bronchial Epithelial Cells control biological 1
## 13 #GSM1068470 = Value for GSM1068470: Bronchial Epithelial Cells control biological 1
## 14 #GSM1068471 = Value for GSM1068471: Bronchial Epithelial Cells control biological 1
## 15 #GSM1068472 = Value for GSM1068472: Bronchial Epithelial Cells control biological 1
## 16 #GSM1068473 = Value for GSM1068473: Bronchial Epithelial Cells control biological 2
## 17 #GSM1068474 = Value for GSM1068474: Bronchial Epithelial Cells control biological 2
## 18 #GSM1068475 = Value for GSM1068475: Bronchial Epithelial Cells control biological 2
## 19 #GSM1068476 = Value for GSM1068476: Bronchial Epithelial Cells control biological 2
## 20 #GSM1068477 = Value for GSM1068477: Bronchial Epithelial Cells control biological 2
## 21      #GSM1068478 = Value for GSM1068478: Bronchial Epithelial Cells MMA biological 0
## 22      #GSM1068479 = Value for GSM1068479: Bronchial Epithelial Cells MMA biological 1

```

23 #GSM1068480 = Value for GSM1068480: Bronchial Epithelial Cells MMA biological 1
24 #GSM1068481 = Value for GSM1068481: Bronchial Epithelial Cells MMA biological 1
25 #GSM1068482 = Value for GSM1068482: Bronchial Epithelial Cells MMA biological 1
26 #GSM1068483 = Value for GSM1068483: Bronchial Epithelial Cells MMA biological 1
27 #GSM1068484 = Value for GSM1068484: Bronchial Epithelial Cells MMA biological 1
28 #GSM1068485 = Value for GSM1068485: Bronchial Epithelial Cells MMA biological 1
29 #GSM1068486 = Value for GSM1068486: Bronchial Epithelial Cells MMA biological 1
30 #GSM1068487 = Value for GSM1068487: Bronchial Epithelial Cells MMA biological 2
31 #GSM1068488 = Value for GSM1068488: Bronchial Epithelial Cells MMA biological 2
32 #GSM1068489 = Value for GSM1068489: Bronchial Epithelial Cells MMA biological 2
33 #GSM1068490 = Value for GSM1068490: Bronchial Epithelial Cells MMA biological 2
34 #GSM1068491 = Value for GSM1068491: Bronchial Epithelial Cells MMA biological 2
35 #GSM1068492 = Value for GSM1068492: Bronchial Epithelial Cells MMA biological 2
36 #GSM1068493 = Value for GSM1068493: Bronchial Epithelial Cells MMA biological 0
37 #GSM1068494 = Value for GSM1068494: Bronchial Epithelial Cells MMA biological 0
38 #GSM1068495 = Value for GSM1068495: Bronchial Epithelial Cells MMA biological 0
39 #GSM1068496 = Value for GSM1068496: Bronchial Epithelial Cells MMA biological 1
40 #GSM1068497 = Value for GSM1068497: Bronchial Epithelial Cells MMA biological 1
41 #GSM1068498 = Value for GSM1068498: Bronchial Epithelial Cells MMA biological 1
42 #GSM1068499 = Value for GSM1068499: Bronchial Epithelial Cells MMA biological 1
43 #GSM1068500 = Value for GSM1068500: Bronchial Epithelial Cells MMA biological 1
44 #GSM1068501 = Value for GSM1068501: Bronchial Epithelial Cells MMA biological 2
45 #GSM1068502 = Value for GSM1068502: Bronchial Epithelial Cells MMA biological 2
46 #GSM1068503 = Value for GSM1068503: Bronchial Epithelial Cells MMA biological 2
47 #GSM1068504 = Value for GSM1068504: Bronchial Epithelial Cells MMA biological 2
48 #GSM1068505 = Value for GSM1068505: Bronchial Epithelial Cells MMA biological 2
49 #GSM1068506 = Value for GSM1068506: Bronchial Epithelial Cells MMA biological 2
50 #GSM1068507 = Value for GSM1068507: Bronchial Epithelial Cells MMA biological 2
51 #GSM1068508 = Value for GSM1068508: Bronchial Epithelial Cells MMA biological 2
52 #GSM1068509 = Value for GSM1068509: Bronchial Epithelial Cells MMA biological 1
53 #GSM1068510 = Value for GSM1068510: Bronchial Epithelial Cells MMA biological 2
54 #GSM1068511 = Value for GSM1068511: Bronchial Epithelial Cells MMA biological 2
55 #GSM1068512 = Value for GSM1068512: Bronchial Epithelial Cells MMA biological 2
56 #GSM1068513 = Value for GSM1068513: Bronchial Epithelial Cells MMA biological 2
57 #GSM1068514 = Value for GSM1068514: Bronchial Epithelial Cells MMA biological 2
58 #GSM1068515 = Value for GSM1068515: Bronchial Epithelial Cells MMA biological 0
59 #GSM1068516 = Value for GSM1068516: Bronchial Epithelial Cells MMA biological 0

60 #GSM1068517 = Value for GSM1068517: Bronchial Epithelial Cells MMA biological 1
61 #GSM1068518 = Value for GSM1068518: Bronchial Epithelial Cells MMA biological 1
62 #GSM1068519 = Value for GSM1068519: Bronchial Epithelial Cells MMA biological 1
63 #GSM1068520 = Value for GSM1068520: Bronchial Epithelial Cells MMA biological 1
64 #GSM1068521 = Value for GSM1068521: Bronchial Epithelial Cells MMA biological 1
65 #GSM1068522 = Value for GSM1068522: Bronchial Epithelial Cells MMA biological 1
66 #GSM1068523 = Value for GSM1068523: Bronchial Epithelial Cells MMA biological 1
67 #GSM1068524 = Value for GSM1068524: Bronchial Epithelial Cells MMA biological 1
68 #GSM1068525 = Value for GSM1068525: Bronchial Epithelial Cells MMA biological 2
69 #GSM1068526 = Value for GSM1068526: Bronchial Epithelial Cells MMA biological 2
70 #GSM1068527 = Value for GSM1068527: Bronchial Epithelial Cells MMA biological 2
71 #GSM1068528 = Value for GSM1068528: Bronchial Epithelial Cells SA biological 0
72 #GSM1068529 = Value for GSM1068529: Bronchial Epithelial Cells SA biological 1
73 #GSM1068530 = Value for GSM1068530: Bronchial Epithelial Cells SA biological 1
74 #GSM1068531 = Value for GSM1068531: Bronchial Epithelial Cells SA biological 1
75 #GSM1068532 = Value for GSM1068532: Bronchial Epithelial Cells SA biological 1
76 #GSM1068533 = Value for GSM1068533: Bronchial Epithelial Cells SA biological 1
77 #GSM1068534 = Value for GSM1068534: Bronchial Epithelial Cells SA biological 1
78 #GSM1068535 = Value for GSM1068535: Bronchial Epithelial Cells SA biological 1
79 #GSM1068536 = Value for GSM1068536: Bronchial Epithelial Cells SA biological 1
80 #GSM1068537 = Value for GSM1068537: Bronchial Epithelial Cells SA biological 1
81 #GSM1068538 = Value for GSM1068538: Bronchial Epithelial Cells SA biological 1
82 #GSM1068539 = Value for GSM1068539: Bronchial Epithelial Cells SA biological 1
83 #GSM1068540 = Value for GSM1068540: Bronchial Epithelial Cells SA biological 1
84 #GSM1068541 = Value for GSM1068541: Bronchial Epithelial Cells SA biological 1
85 #GSM1068542 = Value for GSM1068542: Bronchial Epithelial Cells SA biological 1
86 #GSM1068543 = Value for GSM1068543: Bronchial Epithelial Cells SA biological 1
87 #GSM1068544 = Value for GSM1068544: Bronchial Epithelial Cells SA biological 1
88 #GSM1068545 = Value for GSM1068545: Bronchial Epithelial Cells SA biological 1
89 #GSM1068546 = Value for GSM1068546: Bronchial Epithelial Cells SA biological 1
90 #GSM1068547 = Value for GSM1068547: Bronchial Epithelial Cells SA biological 1
91 #GSM1068548 = Value for GSM1068548: Bronchial Epithelial Cells SA biological 1
92 #GSM1068549 = Value for GSM1068549: Bronchial Epithelial Cells SA biological 1
93 #GSM1068550 = Value for GSM1068550: Bronchial Epithelial Cells SA biological 1
94 #GSM1068551 = Value for GSM1068551: Bronchial Epithelial Cells SA biological 1
95 #GSM1068552 = Value for GSM1068552: Bronchial Epithelial Cells SA biological 1
96 #GSM1068553 = Value for GSM1068553: Bronchial Epithelial Cells SA biological 2


```

## 97      #GSM1068554 = Value for GSM1068554: Bronchial Epithelial Cells SA biological 2
## 98      #GSM1068555 = Value for GSM1068555: Bronchial Epithelial Cells SA biological 2
## 99      #GSM1068556 = Value for GSM1068556: Bronchial Epithelial Cells SA biological 2
## 100     #GSM1068557 = Value for GSM1068557: Bronchial Epithelial Cells SA biological 2
## 101     #GSM1068558 = Value for GSM1068558: Bronchial Epithelial Cells SA biological 2
## 102     #GSM1068559 = Value for GSM1068559: Bronchial Epithelial Cells SA biological 2
## 103     #GSM1068560 = Value for GSM1068560: Bronchial Epithelial Cells SA biological 2
## 104     #GSM1068561 = Value for GSM1068561: Bronchial Epithelial Cells SA biological 2
## 105     #GSM1068562 = Value for GSM1068562: Bronchial Epithelial Cells SA biological 2
## 106     #GSM1068563 = Value for GSM1068563: Bronchial Epithelial Cells SA biological 2
## 107     #GSM1068564 = Value for GSM1068564: Bronchial Epithelial Cells SA biological 2
## 108     #GSM1068565 = Value for GSM1068565: Bronchial Epithelial Cells SA biological 2
##          ID gender
## 1  control female
## 2  control female
## 3  control female
## 4  control female
## 5  control  male
## 6  control  male
## 7  control female
## 8  control  male
## 9  control  male
## 10 control  male
## 11 control female
## 12 control  male
## 13 control  male
## 14 control  male
## 15 control female
## 16 control female
## 17 control female
## 18 control  male
## 19 control female
## 20 control female
## 21      MMA female
## 22      MMA female
## 23      MMA  male
## 24      MMA female

```

## 25	MMA female
## 26	MMA female
## 27	MMA male
## 28	MMA male
## 29	MMA female
## 30	MMA female
## 31	MMA female
## 32	MMA male
## 33	MMA female
## 34	MMA female
## 35	MMA female
## 36	MMA female
## 37	MMA female
## 38	MMA female
## 39	MMA female
## 40	MMA male
## 41	MMA female
## 42	MMA female
## 43	MMA female
## 44	MMA male
## 45	MMA female
## 46	MMA female
## 47	MMA male
## 48	MMA female
## 49	MMA female
## 50	MMA female
## 51	MMA female
## 52	MMA male
## 53	MMA female
## 54	MMA male
## 55	MMA female
## 56	MMA female
## 57	MMA female
## 58	MMA male
## 59	MMA male
## 60	MMA female
## 61	MMA female

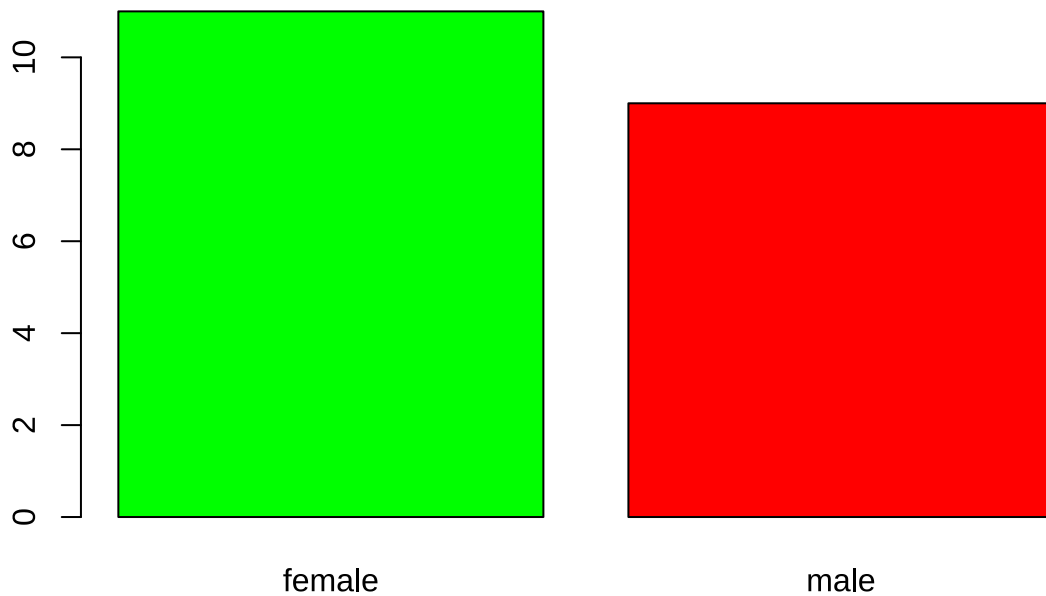
## 62	MMA	male
## 63	MMA	female
## 64	MMA	female
## 65	MMA	female
## 66	MMA	male
## 67	MMA	female
## 68	MMA	male
## 69	MMA	male
## 70	MMA	female
## 71	SA	female
## 72	SA	male
## 73	SA	male
## 74	SA	female
## 75	SA	female
## 76	SA	female
## 77	SA	male
## 78	SA	female
## 79	SA	male
## 80	SA	female
## 81	SA	female
## 82	SA	female
## 83	SA	female
## 84	SA	male
## 85	SA	female
## 86	SA	female
## 87	SA	female
## 88	SA	female
## 89	SA	female
## 90	SA	female
## 91	SA	female
## 92	SA	female
## 93	SA	female
## 94	SA	female
## 95	SA	female
## 96	SA	male
## 97	SA	male
## 98	SA	female

```
## 99      SA female
## 100     SA female
## 101     SA  male
## 102     SA  male
## 103     SA female
## 104     SA female
## 105     SA female
## 106     SA female
## 107     SA  male
## 108     SA female
```

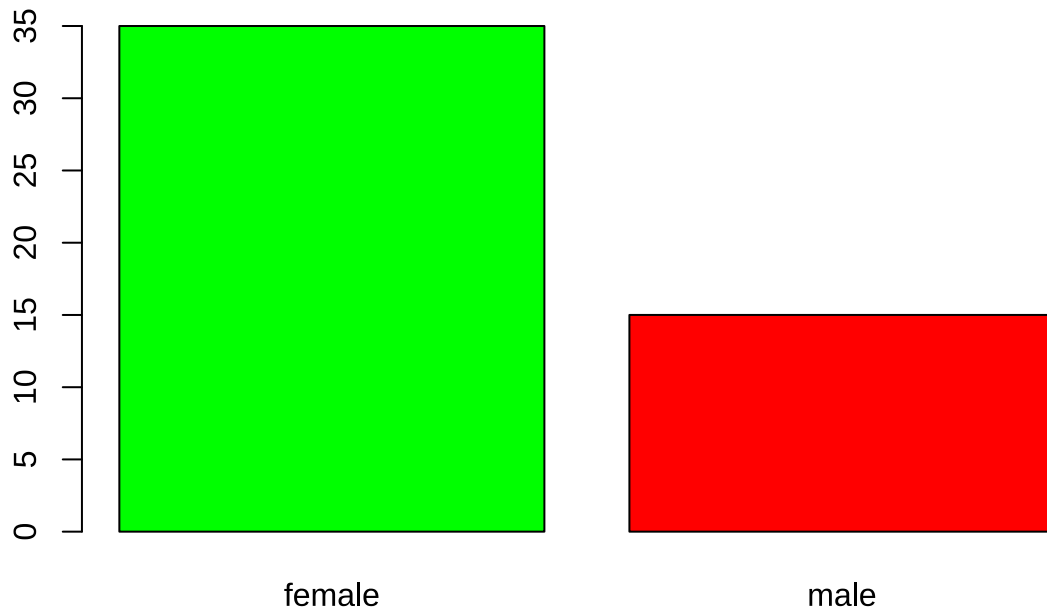
```
write.csv(genderdata,file ='GDS5037/genderdata.csv' )

par(mfrow=c(1,3))

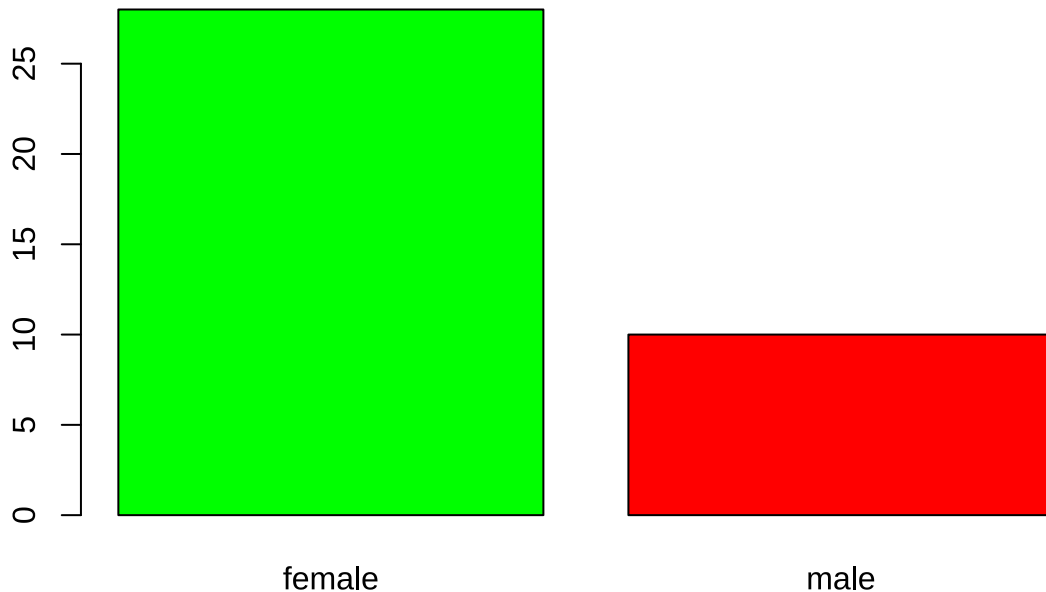
row <- 1
value <- c(su[row,3]$Counts,su[row+1,3]$Counts)
gender <- c('female','male')
barplot(value,names.arg = gender,col = c('green','red'))
```



```
row <- 3
value <- c(su[row,3]$Counts,su[row+1,3]$Counts)
gender <- c('female','male')
barplot(value,names.arg = gender,col = c('green','red'))
```



```
row <- 5
value <- c(su[row,3]$Counts,su[row+1,3]$Counts)
gender <- c('female','male')
barplot(value,names.arg = gender,col = c('green','red'))
```



```
#2.4 group by patients' status, calculate the sample mean and
#variance of IDENTIFIER FAM174bB in each group
data_table <- read.csv('GDS5037/data_table.csv', nrow = 1);
#I know the data I need is the first
ttable <- t(data_table)
d <- ttable
names <- rownames(d)
rownames(d) <- NULL
ttable <- cbind(names, d); ttable <- ttable[-1,]; ttable <- ttable[-1,]
ttable <- data.frame(ttable)
names(ttable) <- c('ID_REF', 'value')
data <- merge(ttable, description, by = 'ID_REF');
data <- data %>% mutate_at(.vars = vars('value'), .fun = as.numeric)
#data$value=as.data.frame(lapply(data$value, as.numeric))
mean <- aggregate(data[,2], list(data[,4]), mean); mean
```

```
## Group.1      x
## 1 control 53.50000
## 2      MMA 58.74000
```

```
## 3      SA 47.76316
```

```
sd <- aggregate(data[,2],list(data[,4]),sd);sd
```

```
##   Group.1      x
## 1 control 26.44856
## 2     MMA 31.18268
## 3      SA 31.47838
```

```
n <- aggregate(data[,2],list(data[,4]),length);n
```

```
##   Group.1  x
## 1 control 20
## 2     MMA 50
## 3      SA 38
```

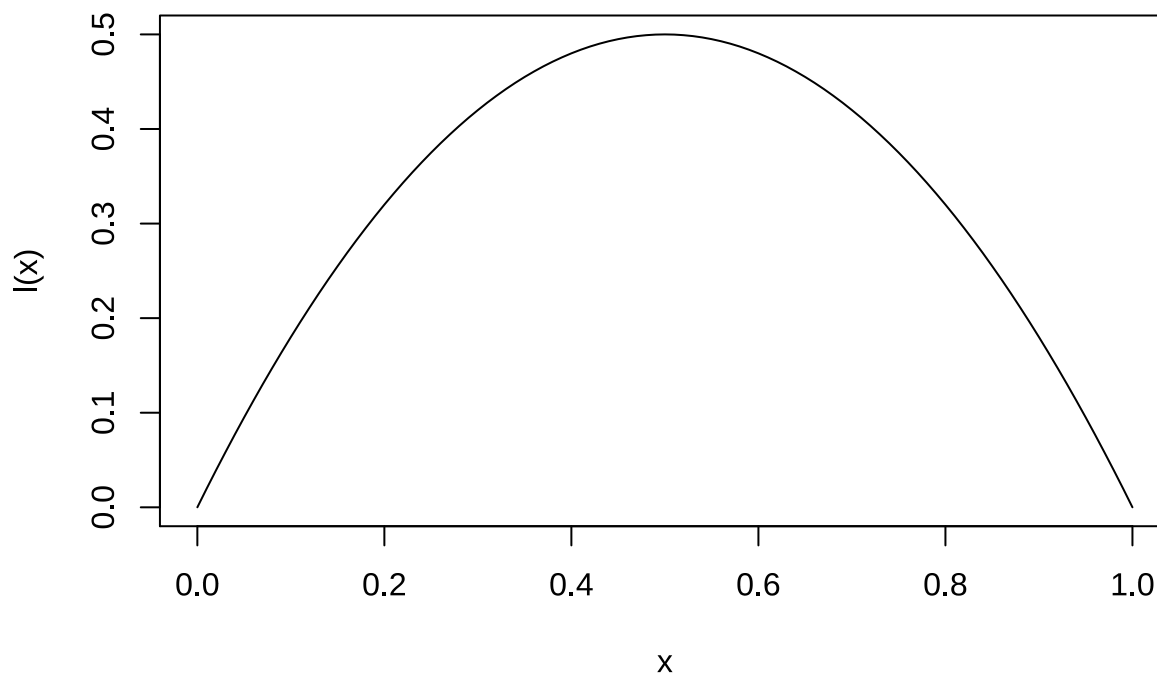
```
z <- (mean[1,2]-mean[3,2])/sqrt(sd[1,2]^2/n[1,2]+sd[3,2]^2/n[3,2])
p <- 1-pnorm(z);p
```

```
## [1] 0.2314096
```

#so the result show that the mean of the two groups are equal under 5% significant level

- 1.4 的 d 题有一个画图题

```
l <- function(x) 2*x*(1-x)
curve(l,0,1)
```



```
l <- function(x) -2*x*(1-x)
optimize(l,c(0,1))
```

```
## $minimum
## [1] 0.5
##
## $objective
## [1] -0.5
```

- so the result show that the minvalue is 0.5 when $\pi = 0.5$
- 1.12

```
library(cdabookfunc)
binom_inference(0,25,0.5,method = 'wald')
```

```
## $z
## [1] -Inf
##
## $method
## [1] "wald"
```