

# Statistical summaries

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BIOF 339



# Where we've been

1. Understand what tidy data is
2. Manipulate data to make it tidy (tidyr, dplyr)
3. Transform particular variables
4. Write basic functions
5. High-throughput analyses
  - Lists of data sets
  - `map` to apply similar processes to each data set
  - for-loops to repeat same recipe on multiple data sets or objects

# Where we're going

1. Creating data summaries
2. Basic statistical comparisons between groups
3. Creating tables
  - Table 1
  - Tables for analytic results

The basic assumption we'll make is that we will start with a tidy data set.

# Statistical summaries

# Univariate summaries

## Single summaries

- Mean (`mean`)
- Variance (`var`)
- Standard deviation (`sd`)
- Count (`nrow` or `dplyr::n` or `dplyr::n_distinct`)
- Median ('`median`')
- Inter-quartile range (`IQR`)
- Mean absolute deviation (`mad`)
- Minimum (`min`) and Maximum (`max`)

## Multiple summaries

- Quantiles (`quantile`)
- Range (`range`)

# Summarizing the breast cancer expression dataset

# Mean

```
brca <- rio::import('../data/BreastCancer_Expression.csv')
brca %>%
  summarize(across(starts_with('NP'),
                    mean, na.rm=T))
```

```
NP_958782 NP_958785 NP_958786 NP_000436 NP_958781 NP_958780 NP_958783
1 0.3202321 0.3269153 0.3264254 0.3236833 0.3270832 0.3263382 0.3259212
NP_958784 NP_112598 NP_001611
1 0.3259995 -0.3074577 0.4578748
```



# Median

```
brca %>%  
  summarize(across(starts_with('NP'),  
                    median, na.rm=T))
```

```
NP_958782 NP_958785 NP_958786 NP_000436 NP_958781 NP_958780 NP_958783  
1 0.3236627 0.3269726 0.3269726 0.3302826 0.3269726 0.3269726 0.3269726  
NP_958784 NP_112598 NP_001611  
1 0.3269726 -0.6021319 0.6948104
```

# Standard deviation

```
brca %>%  
  summarize(across(starts_with('NP'),  
                    sd, na.rm=T))
```

```
NP_958782 NP_958785 NP_958786 NP_000436 NP_958781 NP_958780 NP_958783  
1 0.9767777 0.9800721 0.9799358 0.9784656 0.9806001 0.9796277 0.9806739  
NP_958784 NP_112598 NP_001611  
1 0.9807512 2.024663 1.496951
```

# Multiple summaries together

```
brca %>%  
  summarize(across(starts_with('NP'),  
                    c(mean,  
                      median,  
                      sd), na.rm=T))
```

```
NP_958782_1 NP_958782_2 NP_958782_3 NP_958785_1 NP_958785_2 NP_958785_3  
1 0.3202321 0.3236627 0.9767777 0.3269153 0.3269726 0.9800721  
NP_958786_1 NP_958786_2 NP_958786_3 NP_000436_1 NP_000436_2 NP_000436_3  
1 0.3264254 0.3269726 0.9799358 0.3236833 0.3302826 0.9784656  
NP_958781_1 NP_958781_2 NP_958781_3 NP_958780_1 NP_958780_2 NP_958780_3  
1 0.3270832 0.3269726 0.9806001 0.3263382 0.3269726 0.9796277  
NP_958783_1 NP_958783_2 NP_958783_3 NP_958784_1 NP_958784_2 NP_958784_3  
1 0.3259212 0.3269726 0.9806739 0.3259995 0.3269726 0.9807512  
NP_112598_1 NP_112598_2 NP_112598_3 NP_001611_1 NP_001611_2 NP_001611_3  
1 -0.3074577 -0.6021319 2.024663 0.4578748 0.6948104 1.496951
```

# Multiple summaries together

```
brca %>%  
  summarize(across(-1, # got tired of typing  
    c('Mean'=mean,  
      'Median' = median,  
      'SD'=sd), na.rm=T))
```

```
NP_958782_Mean NP_958782_Median NP_958782_SD NP_958785_Mean NP_958785_Median  
1 0.3202321 0.3236627 0.9767777 0.3269153 0.3269726  
NP_958785_SD NP_958786_Mean NP_958786_Median NP_958786_SD NP_000436_Mean  
1 0.9800721 0.3264254 0.3269726 0.9799358 0.3236833  
NP_000436_Median NP_000436_SD NP_958781_Mean NP_958781_Median NP_958781_SD  
1 0.3302826 0.9784656 0.3270832 0.3269726 0.9806001  
NP_958780_Mean NP_958780_Median NP_958780_SD NP_958783_Mean NP_958783_Median  
1 0.3263382 0.3269726 0.9796277 0.3259212 0.3269726  
NP_958783_SD NP_958784_Mean NP_958784_Median NP_958784_SD NP_112598_Mean  
1 0.9806739 0.3259995 0.3269726 0.9807512 -0.3074577  
NP_112598_Median NP_112598_SD NP_001611_Mean NP_001611_Median NP_001611_SD  
1 -0.6021319 2.024663 0.4578748 0.6948104 1.496951
```

# Multiple summaries together

```
brca %>%
  summarize(across(-1,
    c('Mean' = mean,
      'Median' = median,
      'SD' = sd), na.rm=T)) %>%
  pivot_longer(cols=everything(),
    names_to='variable',
    values_to='value') %>%
  # extract(variable, c('ID','Statistic'),
  #   # regex = '(NP_\\d+)_(\\[A-Za-z]+)') %>%
  separate(variable,
    c("Type","ID","Statistic"), sep='_') %>%
  pivot_wider(names_from = Statistic, values_from = value) %>%
  unite(ID, c('Type','ID'), sep='_')
```

```
# A tibble: 10 × 4
  ID           Mean Median  SD
  <chr>      <dbl> <dbl> <dbl>
1 NP_958782  0.320  0.324 0.977
2 NP_958785  0.327  0.327 0.980
3 NP_958786  0.326  0.327 0.980
4 NP_000436  0.324  0.330 0.978
5 NP_958781  0.327  0.327 0.981
6 NP_958780  0.326  0.327 0.980
7 NP_958783  0.326  0.327 0.981
8 NP_958784  0.326  0.327 0.981
9 NP_112598 -0.307 -0.602 2.02
10 NP_001611  0.458  0.695 1.50
```

You could replace the highlighted code with

```
extract(variable,
  c('ID','Statistic'),
  regex = '(NP_\\d+)_(\\[A-Za-z]+)') %>%
  pivot_wider(
    names_from=Statistic,
    values_from=value)
```

# Summarizing a data set

# Data set summary

There is a function `summary` that will give you summaries of all the variables. It's nice for looking at the data, but the output format isn't very good for further manipulation

```
summary(brca[,-1]) # Omit first column
```

```
NP_958782      NP_958785      NP_958786      NP_000436
Min.   :-1.9478  Min.   :-1.9527  Min.   :-1.9552  Min.   :-1.9478
1st Qu.:-0.4549  1st Qu.:-0.4421  1st Qu.:-0.4440  1st Qu.:-0.4385
Median : 0.3237  Median : 0.3270  Median : 0.3270  Median : 0.3303
Mean   : 0.3202  Mean   : 0.3269  Mean   : 0.3264  Mean   : 0.3237
3rd Qu.: 0.9181  3rd Qu.: 0.9238  3rd Qu.: 0.9238  3rd Qu.: 0.9180
Max.   : 2.7651  Max.   : 2.7797  Max.   : 2.7797  Max.   : 2.7980
NP_958781      NP_958780      NP_958783      NP_958784
Min.   :-1.9576  Min.   :-1.9552  Min.   :-1.9552  Min.   :-1.9552
1st Qu.:-0.4440  1st Qu.:-0.4458  1st Qu.:-0.4440  1st Qu.:-0.4440
Median : 0.3270  Median : 0.3270  Median : 0.3270  Median : 0.3270
Mean   : 0.3271  Mean   : 0.3263  Mean   : 0.3259  Mean   : 0.3260
3rd Qu.: 0.9277  3rd Qu.: 0.9238  3rd Qu.: 0.9238  3rd Qu.: 0.9238
Max.   : 2.7870  Max.   : 2.7797  Max.   : 2.7834  Max.   : 2.7834
NP_112598      NP_001611
Min.   :-4.9527  Min.   :-2.5751
1st Qu.:-1.6741  1st Qu.:-0.5216
Median : -0.6021  Median : 0.6948
Mean   : -0.3075  Mean   : 0.4579
3rd Qu.: 0.8696  3rd Qu.: 1.4394
Max.   : 4.9557  Max.   : 3.4365
```

**Maybe an easier way?**



# The tableone package

The `tableone` package is meant to create, you guessed it, Table 1.

It is quite a convenient package for most purposes and saves gobs of time

# The tableone package

```
library(tableone)
tab1 <- CreateTableOne(data=brca[,-1])
tab1
```

```
Overall
n      83
NP_958782 (mean (SD)) 0.32 (0.98)
NP_958785 (mean (SD)) 0.33 (0.98)
NP_958786 (mean (SD)) 0.33 (0.98)
NP_000436 (mean (SD)) 0.32 (0.98)
NP_958781 (mean (SD)) 0.33 (0.98)
NP_958780 (mean (SD)) 0.33 (0.98)
NP_958783 (mean (SD)) 0.33 (0.98)
NP_958784 (mean (SD)) 0.33 (0.98)
NP_112598 (mean (SD)) -0.31 (2.02)
NP_001611 (mean (SD)) 0.46 (1.50)
```

# The tableone package

```
library(tableone)
tab1 <- CreateTableOne(data = brca[-1])
print(tab1, nonnormal = names(brca)[-1])
```

You have to give the variable names of those you think are non-normally distributed and need to be summarized by the median

```
Overall
n 83
NP_958782 (median [IQR]) 0.32 [-0.45, 0.92]
NP_958785 (median [IQR]) 0.33 [-0.44, 0.92]
NP_958786 (median [IQR]) 0.33 [-0.44, 0.92]
NP_000436 (median [IQR]) 0.33 [-0.44, 0.92]
NP_958781 (median [IQR]) 0.33 [-0.44, 0.93]
NP_958780 (median [IQR]) 0.33 [-0.45, 0.92]
NP_958783 (median [IQR]) 0.33 [-0.44, 0.92]
NP_958784 (median [IQR]) 0.33 [-0.44, 0.92]
NP_112598 (median [IQR]) -0.60 [-1.67, 0.87]
NP_001611 (median [IQR]) 0.69 [-0.52, 1.44]
```

# The tableone package

```
library(tableone)
tab1 <- CreateTableOne(data = brca[-1])
kableone(print(tab1, nonnormal = names(brca)[-1]),
          format='html')
```

---

	Overall
n	83
NP_958782 (median [IQR])	0.32 [-0.45, 0.92]
NP_958785 (median [IQR])	0.33 [-0.44, 0.92]
NP_958786 (median [IQR])	0.33 [-0.44, 0.92]
NP_000436 (median [IQR])	0.33 [-0.44, 0.92]
NP_958781 (median [IQR])	0.33 [-0.44, 0.93]
NP_958780 (median [IQR])	0.33 [-0.45, 0.92]
NP_958783 (median [IQR])	0.33 [-0.44, 0.92]
NP_958784 (median [IQR])	0.33 [-0.44, 0.92]
NP_112598 (median [IQR])	-0.60 [-1.67, 0.87]
NP_001611 (median [IQR])	0.69 [-0.52, 1.44]

---

# Mixed data

Let's first put the expression and clinical data together

```
library(rio)
brca1 <- import('../data/clinical_data_breast_cancer_hw.csv')
brca2 <- import('../data/BreastCancer_Expression.csv')
brca <- left_join(brca1, brca2, by=c('Complete.TCGA.ID' = 'TCGA_ID')) %>%
  mutate(Age.at.Initial.Pathologic.Diagnosis =
         as.numeric(Age.at.Initial.Pathologic.Diagnosis)) %>%
  mutate(ER.Status = ifelse(ER.Status %in% c('Positive', 'Negative'),
                           ER.Status, NA))

summary(brca)
```

```
Complete.TCGA.ID      Gender      Age.at.Initial.Pathologic.Diagnosis
Length:108            Length:108      Min.      :30.00
Class :character      Class :character 1st Qu.   :49.00
Mode  :character      Mode  :character Median    :58.00
                                Mean     :58.72
                                3rd Qu. :66.50
                                Max.    :88.00
                                NA's    :1
ER.Status             PR.Status      HER2.Final.Status      Tumor
Length:108            Length:108      Length:108              Length:108
Class :character      Class :character  Class :character        Class :character
Mode  :character      Mode  :character  Mode  :character        Mode  :character

Node                  Metastasis     AJCC.Stage           Vital.Status
Length:108            Length:108      Length:108             Length:108
Class :character      Class :character  Class :character        Class :character
Mode  :character      Mode  :character  Mode  :character        Mode  :character
```

Let's first put the expression and clinical data together

```
library(rio)
brca1 <- import('../data/clinical_data_breast_cancer_hw.csv')
brca2 <- import('../data/BreastCancer_Expression.csv')
brca <- left_join(brca1, brca2, by=c('Complete.TCGA.ID' = 'TCGA_ID')) %>%
  mutate(Age.at.Initial.Pathologic.Diagnosis =
    as.numeric(Age.at.Initial.Pathologic.Diagnosis)) %>%
  mutate(ER.Status = ifelse(ER.Status %in% c('Positive', 'Negative'),
    ER.Status, NA),
    HER2.Final.Status = ifelse(HER2.Final.Status == 'Equivocal',
    NA, HER2.Final.Status)) %>%
  mutate(across(is.character, as.factor)) %>%
  mutate(Complete.TCGA.ID = as.character(Complete.TCGA.ID))

str(brca)
```

```
'data.frame': 108 obs. of 23 variables:
 $ Complete.TCGA.ID : chr "TCGA-A2-A0T2" "TCGA-A2-A0CM" "TCGA-BH-A18V" "TCGA-BH-A18Q" ...
 $ Gender : Factor w/ 2 levels "FEMALE", "MALE": 1 1 1 1 1 1 1 1 1 1 1 ...
 $ Age.at.Initial.Pathologic.Diagnosis: num 66 40 48 56 38 57 74 60 61 NA ...
 $ ER.Status : Factor w/ 2 levels "Negative", "Positive": 1 1 1 1 1 1 1 1 1 1 1 ...
 $ PR.Status : Factor w/ 2 levels "Negative", "Positive": 1 1 1 1 1 1 1 1 1 1 1 ...
 $ HER2.Final.Status : Factor w/ 2 levels "Negative", "Positive": 1 1 1 1 1 1 1 1 1 1 1 ...
 $ Tumor : Factor w/ 4 levels "T1", "T2", "T3", ...: 3 2 2 2 3 2 3 2 2 2 ...
 $ Node : Factor w/ 4 levels "N0", "N1", "N2", ...: 4 1 2 2 4 1 1 1 1 1 ...
 $ Metastasis : Factor w/ 2 levels "M0", "M1": 2 1 1 1 1 1 1 1 1 1 ...
 $ AJCC.Stage : Factor w/ 11 levels "Stage I", "Stage IA", ...: 11 5 6 6 10 5 6 5 5 5 ...
 $ Vital.Status : Factor w/ 2 levels "DECEASED", "LIVING": 1 1 1 1 2 2 2 2 2 2 ...
 $ Days.to.Date.of.Last.Contact : int 240 754 1555 1692 133 309 425 643 775 964 ...
 $ Days.to.date.of.Death : int 240 754 1555 1692 NA NA NA NA NA NA ...
 $ NP_958782 : num NA 0.683 NA 0.195 NA ...
 $ NP_958785 : num NA 0.694 NA 0.215 NA ...
 $ NP_958786 : num NA 0.698 NA 0.215 NA ...
```

Identify which variables are categorical (factors) and which are continuous (numeric)

```
catvars <- brca %>% select(where(is.factor)) %>% names()  
ctsvars <- brca %>% select(where(is.numeric)) %>% names()
```



```
CreateCatTable(vars = catvars, data = brca)
```

```
Overall
n 108
Gender = MALE (%) 2 ( 1.9)
ER.Status = Positive (%) 69 (64.5)
PR.Status = Positive (%) 55 (50.9)
HER2.Final.Status = Positive (%) 28 (26.2)
Tumor (%)
  T1 16 (14.8)
  T2 67 (62.0)
  T3 19 (17.6)
  T4 6 ( 5.6)
Node (%)
  N0 54 (50.0)
  N1 30 (27.8)
  N2 15 (13.9)
  N3 9 ( 8.3)
Metastasis = M1 (%) 2 ( 1.9)
AJCC.Stage (%)
  Stage I 3 ( 2.8)
  Stage IA 7 ( 6.5)
  Stage IB 2 ( 1.9)
  Stage II 11 (10.2)
  Stage IIA 32 (29.6)
  Stage IIB 23 (21.3)
  Stage III 4 ( 3.7)
  Stage IIIA 12 (11.1)
  Stage IIIB 6 ( 5.6)
  Stage IIIC 6 ( 5.6)
  Stage IV 2 ( 1.9)
Vital.Status = LIVING (%) 97 (89.8)
```

```
CreateContTable(vars = ctsvars, data = brca)
```

```

n
Age.at.Initial.Pathologic.Diagnosis (mean (SD))
Days.to.Date.of.Last.Contact (mean (SD))
Days.to.date.of.Death (mean (SD))
NP_958782 (mean (SD))
NP_958785 (mean (SD))
NP_958786 (mean (SD))
NP_000436 (mean (SD))
NP_958781 (mean (SD))
NP_958780 (mean (SD))
NP_958783 (mean (SD))
NP_958784 (mean (SD))
NP_112598 (mean (SD))
NP_001611 (mean (SD))

Overall
108
58.72 (13.21)
806.37 (667.70)
1254.45 (678.05)
0.32 (0.99)
0.33 (1.00)
0.33 (1.00)
0.32 (0.99)
0.33 (1.00)
0.33 (1.00)
0.33 (1.00)
0.33 (1.00)
-0.30 (2.06)
0.38 (1.46)
```

```

brca <- brca %>%
  rename(
    'Age'='Age.at.Initial.Pathologic.Diagnosis',
    'Last.Contact' = 'Days.to.Date.of.Last.Contact',
    'Death' = 'Days.to.date.of.Death'
  )
ctsvars <- brca %>%
  select(where(is.numeric))%>% names()
CreateContTable(vars = ctsvars, data = brca)

```

	Overall
n	108
Age (mean (SD))	58.72 (13.21)
Last.Contact (mean (SD))	806.37 (667.70)
Death (mean (SD))	1254.45 (678.05)
NP_958782 (mean (SD))	0.32 (0.99)
NP_958785 (mean (SD))	0.33 (1.00)
NP_958786 (mean (SD))	0.33 (1.00)
NP_000436 (mean (SD))	0.32 (0.99)
NP_958781 (mean (SD))	0.33 (1.00)
NP_958780 (mean (SD))	0.33 (1.00)
NP_958783 (mean (SD))	0.33 (1.00)
NP_958784 (mean (SD))	0.33 (1.00)
NP_112598 (mean (SD))	-0.30 (2.06)
NP_001611 (mean (SD))	0.38 (1.46)

# Putting it together

```
CreateTableOne(vars = c(catvars, ctsvars),  
              data = brca)
```

```
Overall  
n 108  
Gender = MALE (%) 2 ( 1.9)  
ER.Status = Positive (%) 69 (64.5)  
PR.Status = Positive (%) 55 (50.9)  
HER2.Final.Status = Positive (%) 28 (26.2)  
Tumor (%)  
  T1 16 (14.8)  
  T2 67 (62.0)  
  T3 19 (17.6)  
  T4 6 ( 5.6)  
Node (%)  
  N0 54 (50.0)  
  N1 30 (27.8)  
  N2 15 (13.9)  
  N3 9 ( 8.3)  
Metastasis = M1 (%) 2 ( 1.9)  
AJCC.Stage (%)  
  Stage I 3 ( 2.8)  
  Stage IA 7 ( 6.5)  
  Stage IB 2 ( 1.9)  
  Stage II 11 (10.2)  
  Stage IIA 32 (29.6)  
  Stage IIB 23 (21.3)  
  Stage III 4 ( 3.7)
```

# Putting it together

```
CreateTableOne(data = brca[,-1])
```

```
Overall
n 108
Gender = MALE (%) 2 ( 1.9)
Age (mean (SD)) 58.72 (13.21)
ER.Status = Positive (%) 69 (64.5)
PR.Status = Positive (%) 55 (50.9)
HER2.Final.Status = Positive (%) 28 (26.2)
Tumor (%)
  T1 16 (14.8)
  T2 67 (62.0)
  T3 19 (17.6)
  T4 6 ( 5.6)
Node (%)
  N0 54 (50.0)
  N1 30 (27.8)
  N2 15 (13.9)
  N3 9 ( 8.3)
Metastasis = M1 (%) 2 ( 1.9)
AJCC.Stage (%)
  Stage I 3 ( 2.8)
  Stage IA 7 ( 6.5)
  Stage IB 2 ( 1.9)
  Stage II 11 (10.2)
  Stage IIA 32 (29.6)
  Stage IIB 23 (21.3)
  Stage III 4 ( 3.7)
```

# Grouped summaries

```
brca %>%
  group_by(ER.Status) %>%
  summarize(across(starts_with('NP'),
                    mean))
```

```
# A tibble: 3 × 11
  ER.Status NP_958782 NP_958785 NP_958786 NP_000436 NP_958781 NP_958780
  <fct>      <dbl>      <dbl>      <dbl>      <dbl>      <dbl>      <dbl>
1 Negative      NA          NA          NA          NA          NA          NA
2 Positive      NA          NA          NA          NA          NA          NA
3 <NA>          NA          NA          NA          NA          NA          NA
# ... with 4 more variables: NP_958783 <dbl>, NP_958784 <dbl>, NP_112598 <dbl>,
#   NP_001611 <dbl>
```

There are missing values now, so we have to use `na.rm=T`.

```
brca %>%
  group_by(ER.Status) %>%
  summarize(across(starts_with('NP'),
                    mean, na.rm=T))
```

We still have a row for the missing values of ER.Status

```
# A tibble: 3 × 11
  ER.Status NP_958782 NP_958785 NP_958786 NP_000436 NP_958781 NP_958780
  <fct>      <dbl>      <dbl>      <dbl>      <dbl>      <dbl>      <dbl>
1 Negative    0.429      0.438      0.439      0.432      0.436      0.436
2 Positive    0.267      0.273      0.272      0.271      0.274      0.273
3 <NA>        NaN        NaN        NaN        NaN        NaN        NaN
# ... with 4 more variables: NP_958783 <dbl>, NP_958784 <dbl>, NP_112598 <dbl>,
#   NP_001611 <dbl>
```



```
brca %>%  
  filter(!is.na(ER.Status)) %>%  
  group_by(ER.Status) %>%  
  summarize(across(starts_with('NP'),  
                    mean, na.rm=T))
```

```
# A tibble: 2 × 11  
  ER.Status NP_958782 NP_958785 NP_958786 NP_000436 NP_958781 NP_958780  
  <fct>      <dbl>     <dbl>     <dbl>     <dbl>     <dbl>     <dbl>  
1 Negative    0.429     0.438     0.439     0.432     0.436     0.436  
2 Positive    0.267     0.273     0.272     0.271     0.274     0.273  
# ... with 4 more variables: NP_958783 <dbl>, NP_958784 <dbl>, NP_112598 <dbl>,  
#   NP_001611 <dbl>
```

How about reversing the rows and columns for readability

```
brca %>%
  filter(!is.na(ER.Status)) %>%
  group_by(ER.Status) %>%
  summarize(across(starts_with('NP'),
                    mean, na.rm=T)) %>%
  pivot_longer(names_to='ID', values_to='value',
               cols = c(-ER.Status)) %>%
  pivot_wider(names_from = ER.Status,
              values_from=value)
```

```
# A tibble: 10 × 3
  ID      Negative Positive
  <chr>    <dbl>    <dbl>
1 NP_958782  0.429    0.267
2 NP_958785  0.438    0.273
3 NP_958786  0.439    0.272
4 NP_000436  0.432    0.271
5 NP_958781  0.436    0.274
6 NP_958780  0.436    0.273
7 NP_958783  0.436    0.272
8 NP_958784  0.436    0.273
9 NP_112598 -0.197   -0.357
10 NP_001611 -0.566    0.840
```

# Using tableone

```
CreateTableOne(  
  data = brca %>% filter(!is.na(ER.Status)),  
  vars = brca %>%  
    select(starts_with('NP')) %>%  
    names(),  
  strata = 'ER.Status', # single quotes, not backticks  
  test = F)
```

	Stratified by ER.Status	
	Negative	Positive
n	38	69
NP_958782 (mean (SD))	0.43 (1.13)	0.27 (0.93)
NP_958785 (mean (SD))	0.44 (1.14)	0.27 (0.93)
NP_958786 (mean (SD))	0.44 (1.14)	0.27 (0.93)
NP_000436 (mean (SD))	0.43 (1.14)	0.27 (0.93)
NP_958781 (mean (SD))	0.44 (1.14)	0.27 (0.93)
NP_958780 (mean (SD))	0.44 (1.14)	0.27 (0.93)
NP_958783 (mean (SD))	0.44 (1.14)	0.27 (0.93)
NP_958784 (mean (SD))	0.44 (1.14)	0.27 (0.93)
NP_112598 (mean (SD))	-0.20 (2.28)	-0.36 (1.97)
NP_001611 (mean (SD))	-0.57 (1.54)	0.84 (1.19)

# Alternatives to tableone

- table1
- gtsummary
- flextable
- arsenal

# arsenal

```
library(arsenal)
summary(tableby(ER.Status ~ ., data = brca[,-1])) # Here . implies all other variables.
```

	Negative (N=38)	Positive (N=69)	Total (N=107)	p value
<b>Gender</b>				0.289
FEMALE	38 (100.0%)	67 (97.1%)	105 (98.1%)	
MALE	0 (0.0%)	2 (2.9%)	2 (1.9%)	
<b>Age</b>				0.101
N-Miss	1	0	1	
Mean (SD)	55.919 (12.269)	60.348 (13.573)	58.802 (13.245)	
Range	36.000 - 82.000	30.000 - 88.000	30.000 - 88.000	
<b>PR.Status</b>				< 0.001
Negative	38 (100.0%)	14 (20.3%)	52 (48.6%)	
Positive	0 (0.0%)	55 (79.7%)	55 (51.4%)	
<b>HER2.Final.Status</b>				0.281
N-Miss	0	1	1	
Negative	26 (68.4%)	53 (77.9%)	79 (74.5%)	