

# TNBC\_Phenotype\_Distribution

2022-10-12

```
library(readr)
library(dplyr)

##
## Attaching package: 'dplyr'
## The following objects are masked from 'package:stats':
##
##   filter, lag
## The following objects are masked from 'package:base':
##
##   intersect, setdiff, setequal, union
library(ggplot2)
library(hrbrthemes)

## NOTE: Either Arial Narrow or Roboto Condensed fonts are required to use these themes.
##       Please use hrbrthemes::import_roboto_condensed() to install Roboto Condensed and
##       if Arial Narrow is not on your system, please see https://bit.ly/arialnarrow
library(RColorBrewer)
```

## CSV

Read in the TNBC revised csv file.

```
TNBC <- readr::read_csv("/Users/henzhwang/Desktop/TNBC_training/MIBI-TNBC_scddata_counts_mm_matlab_revised.csv")

## Rows: 179194 Columns: 64
## -- Column specification -----
## Delimiter: ","
## chr (3): SITE_02, RECURRENCE_LABEL, mm
## dbl (61): sample_id, patient_id, AGE_AT_DX, STAGE, LATERAL, GRADE, Survival_...
##
## i Use `spec()` to retrieve the full column specification for this data.
## i Specify the column types or set `show_col_types = FALSE` to quiet this message.
```

## Some statistics of the dataset

```
# Column names in the dataset
names(TNBC)
```

```
## [1] "sample_id"      "patient_id"      "AGE_AT_DX"
## [4] "STAGE"          "SITE_02"         "LATERAL"
## [7] "GRADE"         "RECURRENCE_LABEL" "Survival_days_capped"
## [10] "cluster_id"     "mm"              "cell_type"
```

```

## [13] "ImageNb"           "cellLabelInImage"   "cellSize"
## [16] "cellRadius"        "centroidX"          "centroidY"
## [19] "majoraxis"         "eccentricity"       "Au"
## [22] "Background"        "betaCatenin"        "Ca"
## [25] "CD11b"             "CD11c"              "CD138"
## [28] "CD16"              "CD20"               "CD209"
## [31] "CD3"               "CD31"               "CD4"
## [34] "CD45"              "CD45RO"             "CD56"
## [37] "CD63"              "CD68"               "CD8"
## [40] "dsDNA"             "EGFR"                "Fe"
## [43] "FoxP3"             "H3K27me3"           "H3K9ac"
## [46] "HLA_Class_1"       "HLADR"               "IDO"
## [49] "Keratin17"         "Keratin6"           "Ki67"
## [52] "Lag3"              "MPO"                 "Na"
## [55] "P"                 "p53"                 "panKeratin"
## [58] "PD1"               "PDL1"                "pS6"
## [61] "Si"                "SMA"                 "Ta"
## [64] "Vimentin"

print("-----")

## [1] "-----"
# Check if patient_id and sample_id have the same number
count_patient <- n_distinct(TNBC$patient_id)
count_sample <- n_distinct(TNBC$sample_id)
dplyr::setequal(count_sample, count_patient)

## [1] TRUE
print(paste("There are equal number of", count_patient, "patients and samples in the dataset.))

## [1] "There are equal number of 39 patients and samples in the dataset."
print("-----")

## [1] "-----"
# Checking whether there are duplicates in cellLabelInImage column
dplyr::select(TNBC, cellLabelInImage) %>%
  duplicated() %>% sum()

## [1] 169164
print("-----")

## [1] "-----"
# Number of unique cell types and their names
## There are total of 16 different cells in the dataset
type_counts <- TNBC %>%
  group_by(mm) %>%
  summarise(count = n_distinct(cellLabelInImage))
  summarise(n = n(), NumOfSamples = n_distinct(sample_id))
type_counts

## # A tibble: 16 x 3
##   mm           n NumOfSamples
##   <chr>       <int>       <int>

```

```
## 1 B 17084 31
## 2 CD3 T 1135 22
## 3 CD4 T 9918 36
## 4 CD8 T 13376 36
## 5 DC 2381 34
## 6 DC/Mono 1280 28
## 7 Endothelial 279 25
## 8 Epithelial 31871 39
## 9 Mac 5552 38
## 10 Mesenchymal 27698 39
## 11 Mono/Neu 835 36
## 12 Neu 1365 36
## 13 NK 285 26
## 14 Other 56603 39
## 15 Other immune 8669 36
## 16 T reg 863 22
```

## Extract needed columns

Extract columns that are useful for our analysis.

```
TNBC <- TNBC %>%
  dplyr::select(c(sample_id, patient_id, Survival_days_capped,
                  cluster_id, mm, cell_type, ImageNb, cellLabelInImage,
                  cellSize, cellRadius, centroidX, centroidY))
head(TNBC)

## # A tibble: 6 x 12
##   sample-1 patie-2 Survi-3 clust-4 mm cell_~5 ImageNb cellL-6 cellS-7 cellR-8
##   <dbl> <dbl> <dbl> <dbl> <chr> <dbl> <dbl> <dbl> <dbl> <dbl>
## 1 1 30824 2612 34 B 5 1 10 211 7.82
## 2 1 30824 2612 11 B 5 1 17 184 7.20
## 3 1 30824 2612 31 B 5 1 18 277 8.94
## 4 1 30824 2612 33 B 5 1 47 564 13.0
## 5 1 30824 2612 31 B 5 1 49 402 11.1
## 6 1 30824 2612 31 B 5 1 65 705 14.7
## # ... with 2 more variables: centroidX <dbl>, centroidY <dbl>, and abbreviated
## # variable names 1: sample_id, 2: patient_id, 3: Survival_days_capped,
## # 4: cluster_id, 5: cell_type, 6: cellLabelInImage, 7: cellSize,
## # 8: cellRadius
```

## Range of the spatial corrdinate x and y

First we want to find the range of the spatial corrdinate  $x$  and  $y$  in the whole dataset.

```
centroidX <- TNBC$centroidX
centroidY <- TNBC$centroidY

# Range for the whole dataset
range_wholeX <- c(min(centroidX), max(centroidX), median(centroidX), sd(centroidX))
range_wholeY <- c(min(centroidY), max(centroidY), median(centroidY), sd(centroidY))

print(paste("The range of centroid X in the whole dataset is (", range_wholeX[1], ",", range_wholeX[2],
## [1] "The range of centroid X in the whole dataset is ( 4.511905 , 2043.474 ), the median is 1054.774
```

```
print(paste("The range of centroid X in the whole dataset is (", range_wholeY[1], ",", range_wholeY[2],
```

```
## [1] "The range of centroid X in the whole dataset is ( 3.514286 , 2044.483 ), the median is 1055.116
```

Now we want to find the range of the spatial coordinate for each cell types in the dataset.

```
# Find the range of each cell types in the dataset
```

```
## centroidX
```

```
cells_corrX <- TNBC %>%
```

```
  group_by(mm) %>%
```

```
  summarise(n = n(), min = min(centroidX), max = max(centroidX), median = median(centroidX), sd = sd(centroidX))
```

```
## # A tibble: 16 x 6
```

##	mm	n	min	max	median	sd
##	<chr>	<int>	<dbl>	<dbl>	<dbl>	<dbl>
##	1 B	17084	4.58	2043.	1069.	541.
##	2 CD3 T	1135	6.78	2043.	1425.	501.
##	3 CD4 T	9918	4.61	2043.	1324.	560.
##	4 CD8 T	13376	4.69	2043.	1110.	570.
##	5 DC	2381	4.57	2043.	949.	646.
##	6 DC/Mono	1280	9.00	2043.	1309.	499.
##	7 Endothelial	279	28.7	2023.	1310.	536.
##	8 Epithelial	31871	4.57	2043.	1118.	566.
##	9 Mac	5552	4.69	2043.	1194.	559.
##	10 Mesenchymal	27698	4.56	2043.	1029.	576.
##	11 Mono/Neu	835	10.0	2043.	1266.	547.
##	12 Neu	1365	8.70	2043.	1154.	569.
##	13 NK	285	15.8	2029.	1343.	523.
##	14 Other	56603	4.51	2043.	914.	588.
##	15 Other immune	8669	4.64	2043.	1056.	572.
##	16 T reg	863	9.92	2040.	1233.	512.

```
## centroidY
```

```
cells_corrY <- TNBC %>%
```

```
  group_by(mm) %>%
```

```
  summarise(n = n(), min = min(centroidY), max = max(centroidY), median = median(centroidY), sd = sd(centroidY))
```

```
## # A tibble: 16 x 6
```

##	mm	n	min	max	median	sd
##	<chr>	<int>	<dbl>	<dbl>	<dbl>	<dbl>
##	1 B	17084	4.6	2044.	1144.	577.
##	2 CD3 T	1135	14.9	2044.	1070.	480.
##	3 CD4 T	9918	3.86	2044.	1107.	553.
##	4 CD8 T	13376	4.04	2044.	1096.	575.
##	5 DC	2381	5.20	2043.	1122.	619.
##	6 DC/Mono	1280	3.65	2043.	1004.	535.
##	7 Endothelial	279	20.4	2011.	902.	494.
##	8 Epithelial	31871	4.49	2044.	949.	558.
##	9 Mac	5552	5.91	2043.	977.	569.
##	10 Mesenchymal	27698	3.78	2044.	979.	580.
##	11 Mono/Neu	835	11.0	2044.	1257.	532.
##	12 Neu	1365	7.00	2043.	1214.	538.
##	13 NK	285	18.1	2042.	1041.	540.
##	14 Other	56603	3.51	2044.	1121.	595.

```
## 15 Other immune 8669 4.25 2044. 1041. 588.
## 16 T reg        863 10.0 2038. 973. 526.
```

Next we want to find the range of spatial coordinate for each cell types in each sample.

```
# Find range of corrdinate for each cell types in each sample
## centroidX
cells_corrdX_perSample <- TNBC %>%
  group_by(sample_id, mm) %>%
  summarise(n = n(), min = min(centroidX), max = max(centroidX), median = median(centroidX), sd = sd(centroidX))
cells_corrdX_perSample
```

```
## # A tibble: 523 x 7
## # Groups:   sample_id [39]
##   sample_id mm          n    min    max median    sd
##   <dbl> <chr>    <int> <dbl> <dbl> <dbl> <dbl>
## 1         1 B          734   34.6 2022. 1062. 321.
## 2         1 CD4 T      152  100. 2043. 1566. 516.
## 3         1 CD8 T      147   9.66 2041. 1147. 615.
## 4         1 DC           1  238.  238.  238.  NA
## 5         1 Epithelial    20  44.7 2042. 1742. 584.
## 6         1 Mac           10  347. 2003. 1520. 528.
## 7         1 Mesenchymal 289  36.4 2042. 1074. 518.
## 8         1 Mono/Neu       2 1540. 1938. 1739. 281.
## 9         1 Neu           2 1159. 1572. 1365. 292.
## 10        1 NK            4  610. 1141.  644. 255.
## # ... with 513 more rows
```

```
## centroidY
cells_corrdY_perSample <- TNBC %>%
  group_by(sample_id, mm) %>%
  summarise(n = n(), min = min(centroidY), max = max(centroidY), median = median(centroidY), sd = sd(centroidY))
cells_corrdY_perSample
```

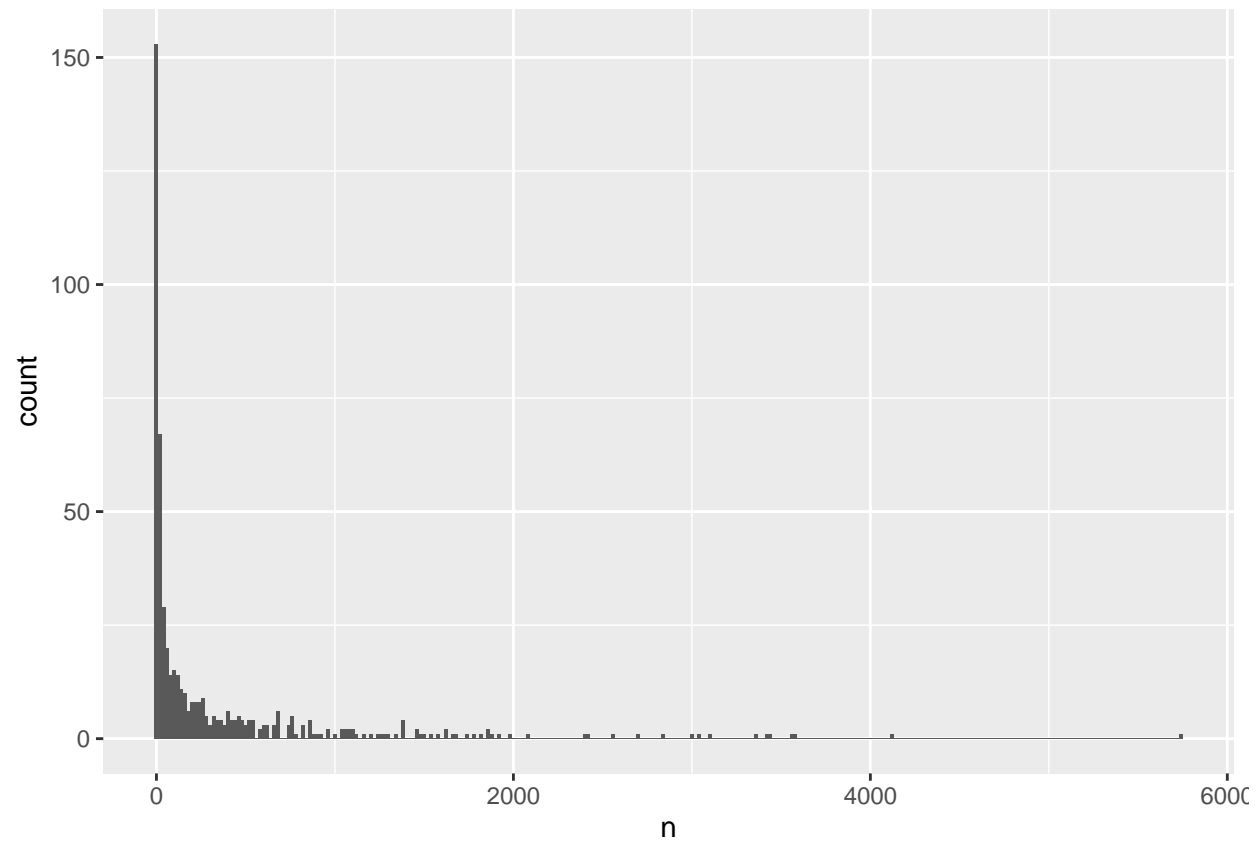
```
## # A tibble: 523 x 7
## # Groups:   sample_id [39]
##   sample_id mm          n    min    max median    sd
##   <dbl> <chr>    <int> <dbl> <dbl> <dbl> <dbl>
## 1         1 B          734   6.73 2043. 1793. 525.
## 2         1 CD4 T      152  12.7 2000. 1005. 635.
## 3         1 CD8 T      147  12.1 1998. 1074. 624.
## 4         1 DC           1  412.  412.  412.  NA
## 5         1 Epithelial    20   6.44 818.  240. 284.
## 6         1 Mac           10  418. 1767. 1100. 461.
## 7         1 Mesenchymal 289   8.05 2038.  596. 538.
## 8         1 Mono/Neu       2  170. 1415.  793. 880.
## 9         1 Neu           2  988. 1686. 1337. 494.
## 10        1 NK            4  301. 1086.  872. 349.
## # ... with 513 more rows
```

## Phenotype distribution for each sample

Want to make a heatmap of the phenotype distribution where sample number vs. cell types. We first want to plot a distribution plot of the number of cells in the dataset.

```
# hist 1
cell_counts <- cells_corrdX_perSample
```

```
ggplot(cell_counts, aes(x = n)) +  
  geom_histogram(binwidth = 20)
```



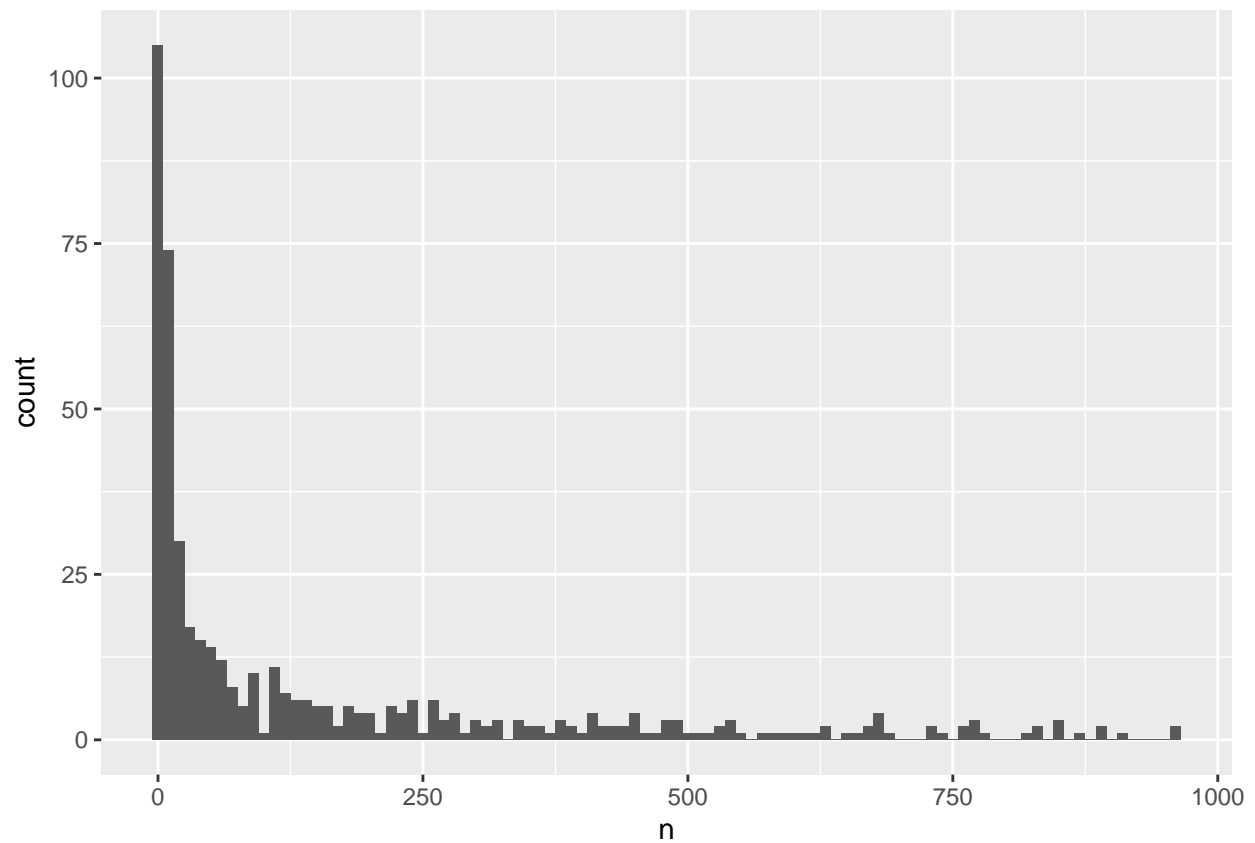
*## We notice that majority of the distribution is less than 1000 count, we then want to draw a new hist*

*# hist 2*

```
cell_counts$n[cell_counts$n >= 1000] <- NA
```

```
ggplot(cell_counts, aes(x = n)) +  
  geom_histogram(binwidth = 10)
```

## Warning: Removed 55 rows containing non-finite values (stat\_bin).



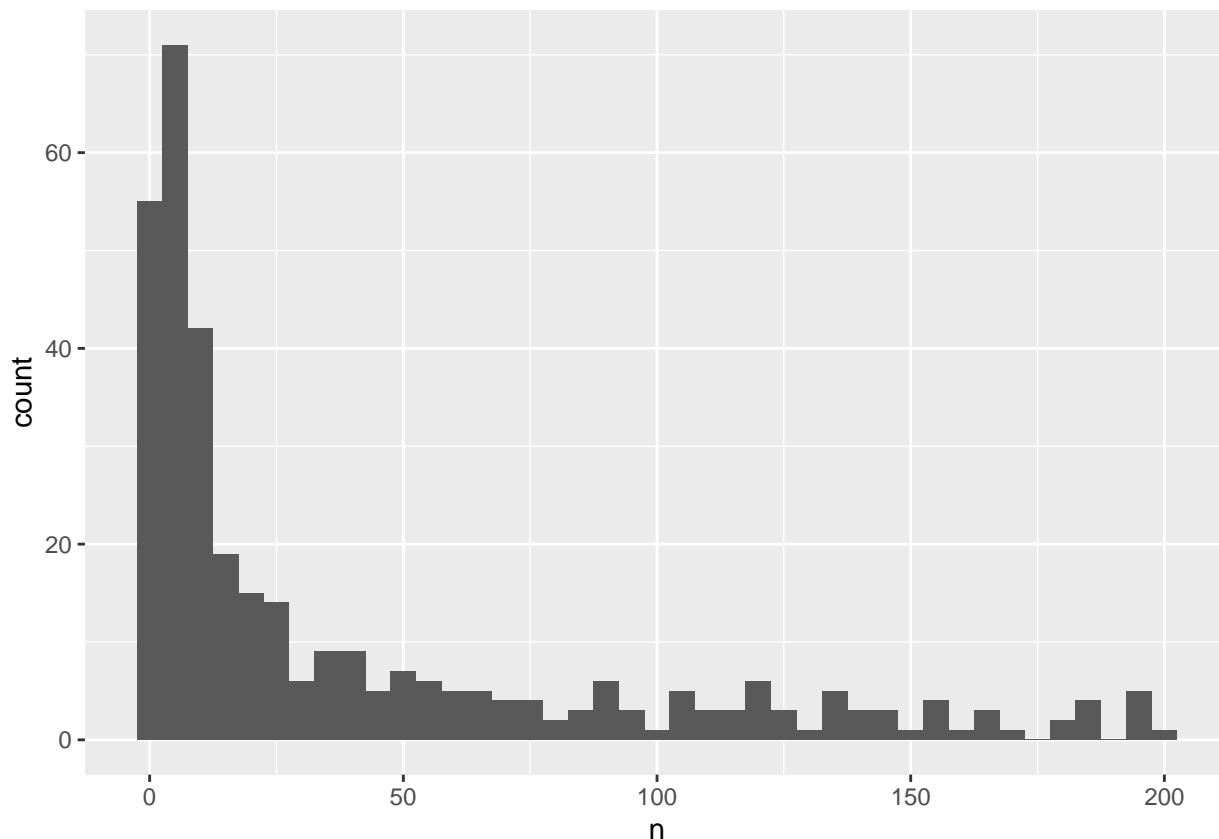
*## We notice that again the majority of the distribution is less than 200, we want to take a closer look*

*# hist 3*

```
cell_counts$n[cell_counts$n >= 200] <- NA
```

```
ggplot(cell_counts, aes(x = n)) +  
  geom_histogram(binwidth = 5)
```

## Warning: Removed 178 rows containing non-finite values (stat\_bin).



(Why the sample\_id is outlier for B cell?) (How well the sequencing for the sample\_id?)

```
# Heatmap
#sample_number <- paste("Sample", sort(unique(TNBC$sample_id)))
#sample_number <- c("Sample 1", "Sample 2", "Sample 3", "Sample 4")

TNBC$sample_id[TNBC$sample_id < 10] <- paste0("0", TNBC$sample_id[TNBC$sample_id < 10])

TNBC %>%
  group_by(sample_id, mm) %>%
  mutate(n = n()) %>%
  mutate(sample_id = paste("Sample", sample_id)) %>%
  mutate(sample_id = factor(sample_id, levels = rev(sort(unique(sample_id))))) %>%
  select(c(sample_id, mm, n)) %>%
  #as.data.frame() %>%

  mutate(mm = factor(mm, levels = rev(sort(unique(mm))))) %>%
  mutate(count = cut(n, breaks = c(-1, 1.1, 5.1, 10.1,
                                25.1, 50.1, 100.1, 250.1, 500.1, 1000.1, 2000.1, max(n, na.rm = TRUE)),
                                labels = c("0-1", "1-5", "5-10", "10-25", "25-50",
                                             "50-100", "100-250", "250-500", "500-1000", "1000-2000", ">2000"))) %>%
  mutate(count = factor(as.character(count), levels = rev(levels(count)))) %>%

  ggplot(mapping = aes(x = mm, y = sample_id, fill = count)) +
  geom_tile(colour = "white", size = 0.3) +
  guides(fill=guide_legend(title = "Number of Cells \nin Each Sample"))+
  labs(x="", y="", title = "Phenotype Distribution of Triple Negative Breast Cancer")+
  scale_y_discrete(expand = c(0, 0)) +
```



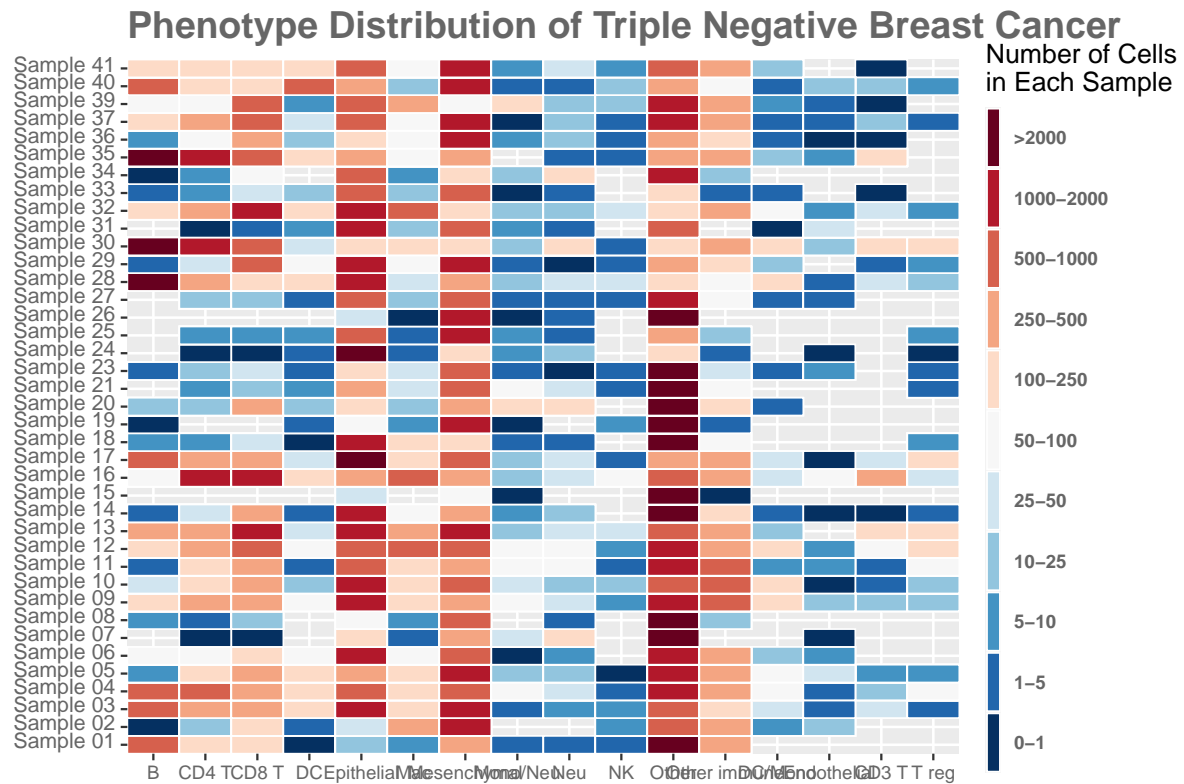
```

scale_x_discrete(expand = c(0, 0)) +

#scale_fill_manual(values = rev(brewer.pal(11, "RdGy")), na.value = "azure4") + #RdGy RdBu
scale_fill_manual(values = brewer.pal(11, "RdBu"), na.value = "azure4") +
#scale_fill_manual(values=c("#d53e4f", "#f46d43", "#fdae61", "#fee08b",
#                             "#e6f598", "#abdda4", "#dddfda"), na.value = "grey90") +

#theme_grey(base_size=6)+
# #theme options
#   theme(
#     #bold font for legend text
#     legend.text=element_text(face="bold"),
#     #set thickness of axis ticks
#     axis.ticks=element_line(size=0.4),
#     #remove plot background
#     plot.background=element_blank(),
#     #remove plot border
#     panel.border=element_blank()
#   )
theme_grey(base_size = 10)+
theme(legend.position = "right", legend.direction = "vertical",
      legend.title = element_text(colour = "black"),
      legend.margin = margin(grid::unit(0, "cm")),
      legend.text = element_text(colour = "grey40", size = 7, face = "bold"),
      legend.key.height = grid::unit(0.8, "cm"),
      legend.key.width = grid::unit(0.2, "cm"),
      axis.text.x = element_text(size = 7, colour = "grey40"),
      axis.text.y = element_text(vjust = 0.2, colour = "grey40"),
      axis.ticks = element_line(size = 0.4),
      plot.background = element_blank(),
      panel.border = element_blank(),
      plot.margin = margin(0.7, 0.4, 0.1, 0.2, "cm"),
      plot.title = element_text(colour = "grey40", hjust = 0, size = 14, face = "bold")
)

```



Reference: <https://www.royfrancis.com/a-guide-to-elegant-tiled-heatmaps-in-r-2019/>

Next:

To explore whether we can colour the cell in the tiff file based on cell type. Chap 11 Modern Stats Modern Bio