ddPCR

J.Schuv

30 4 2021

ddPCR Script The following script takes raw calculated copy number / μ L from ddPCR analysed in Quanta Soft software as input.

```
Built with 4.0.3
```

```
## Warning: package 'ggplot2' was built under R version 4.0.4
## Warning: package 'dplyr' was built under R version 4.0.4
## 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
## Loading required package: carData
## Attaching package: 'car'
## The following object is masked from 'package:dplyr':
##
##
      recode
#Working directory
# setwd(paste(rstudioapi::getSourceEditorContext()$path, "/../..", sep = ""))
#create folder for output graphs
if (!dir.exists(paste(getwd(), "/Graphs_", ProjectName, "/", sep=""))) {
 dir.create(paste(getwd(), "/Graphs_", ProjectName, "/", sep=""))
#create folder for extracted data like lists and dataframes
if (!dir.exists(paste(getwd(), "/ExtractedData_", ProjectName, "/", sep=""))) {
 dir.create(paste(getwd(), "/ExtractedData_", ProjectName, "/", sep=""))
#load all samples
```

```
if (file.exists(paste(getwd(), "/ExtractedData_", ProjectName, "/saveEnvironment.RData", sep=""))) {
  load(paste(getwd(), "/ExtractedData_", ProjectName, "/saveEnvironment.RData", sep=""))} else {
#import data, genes for current Project (here: allChr)
data_raw <- read_excel(pasteO(getwd(), "/../",inputname), tabname)</pre>
#clean input
data_clean <- data_raw[1:24,1:18]</pre>
#remove unnecessary columns
data_reduced <- data_clean[,c(1:4,6:8, 14)]</pre>
#save cleaned df
write_xlsx(data_reduced, paste0(getwd(),"/ExtractedData_", ProjectName, "/raw_cleaned.xlsx"))
#clean env
rm(data_raw)
save.image(file = paste(getwd(), "/ExtractedData_", ProjectName, "/saveEnvironment.RData", sep=""))
}
#show
head(data reduced)
## # A tibble: 6 x 8
     Well Sample Target 'Conc(copies/µL)' 'Accepted Droplets' Positives Negatives
##
     <chr> <chr> <chr>
                                                                                <dbl>
##
                                      <dbl>
                                                           <dbl>
                                                                     <dbl>
## 1 AO1
           N1
                  jct1
                                       85.2
                                                           19420
                                                                      1356
                                                                                18064
## 2 A02
           N1
                  jct1
                                       90.0
                                                           17416
                                                                      1282
                                                                                16134
## 3 A03
                                       90.5
                                                           19202
                                                                      1422
                                                                                17780
          N1
                  jct1
## 4 AO4
          N1
                  rpp30
                                      238.
                                                           17898
                                                                      3281
                                                                                14617
## 5 A05
          N1
                                      249.
                                                           18511
                                                                      3528
                                                                                14983
                  rpp30
## 6 A06
           N1
                  rpp30
                                      210.
                                                           18900
                                                                      3083
                                                                                15817
## # ... with 1 more variable: Threshold1 <dbl>
```

Normalise

Reference genes is avaraged per sample, then the conc of target is divided by that ref value yielding one normalized fraction of amplified structure per input. Here: biolg. triplicates -> three values

A second step includes averaging these three values per sample, returns mean and sd

```
## normalize ----
PlotName <- "ddPCR8"

#get backup
dd8 <- data_reduced

#derive reference value for each sample</pre>
```

```
dd8_mean <- dd8 %>%
  group_by(Sample, .add = TRUE) %>%
  group_by(Target, .add = TRUE) %>%
  summarise(mean = mean('Conc(copies/µL)'),
    SD = sd('Conc(copies/µL)'))
```

'summarise()' has grouped output by 'Sample'. You can override using the '.groups' argument.

```
#normalize each target with respective ref value
dd8_target <- dd8[dd8$Target == target,]</pre>
##loop for normalization
dd8_norm <- vector()
for (i in levels(factor(dd8_target$Sample))){
  subdat <- dd8_target[dd8_target$Sample == i,]</pre>
  #add ref value
  subdat$refmean <- dd8_mean$mean[dd8_mean$Target == reference &</pre>
                                    dd8 mean$Sample == i]
  #normalize for "i"
  subdat$concNorm <- subdat$'Conc(copies/μL)' / subdat$refmean
  #mean and sd for plot
  subdat$mean <- mean(subdat$concNorm)</pre>
  subdat$sd <- sd(subdat$concNorm)</pre>
  #output
  dd8_norm <- rbind(dd8_norm, subdat)
head(dd8_norm)
```

```
## # A tibble: 6 x 12
    Well Sample Target 'Conc(copies/µL)' 'Accepted Droplets' Positives Negatives
##
     <chr> <chr> <chr>
                                     <dbl>
                                                          <dbl>
                                                                    <dbl>
                                                                              <dbl>
## 1 D01
                                      70.7
                                                         17892
                                                                     1043
                                                                              16849
          blood
                  jct1
## 2 D02
                                      80.6
                                                         16921
                                                                     1121
          blood jct1
                                                                              15800
## 3 D03 blood jct1
                                      69.8
                                                         17904
                                                                     1032
                                                                              16872
## 4 CO1
          F1
                  jct1
                                     120.
                                                         17444
                                                                     1696
                                                                              15748
## 5 CO2
          F1
                  jct1
                                     131.
                                                         16743
                                                                     1765
                                                                              14978
## 6 CO3
                                     113.
                                                         18458
                                                                     1684
                                                                              16774
          F1
                  jct1
## # ... with 5 more variables: Threshold1 <dbl>, refmean <dbl>, concNorm <dbl>,
      mean <dbl>, sd <dbl>
```

Data render and export

Before continuing, we need to manipulate the data. The group *Sample* becomes a factor. The sd and mean will be manually calculated per Sample. This is for exporting the data

'summarise()' has grouped output by 'Sample'. You can override using the '.groups' argument.

```
#export numbers
write_xlsx(dd8_norm, path=paste(getwd(), "/ExtractedData_", ProjectName, "/", "numbersForPlot.xlsx", se
write_xlsx(dd8_norm2, path=paste(getwd(), "/ExtractedData_", ProjectName, "/", "numbersForPlot_short.xl
head(dd8_norm2)
## # A tibble: 4 x 4
```

```
## # A tibble: 4 x 4
## # Groups: Sample [4]
## Sample Target mean SD
## <fct> <chr> <dbl> <dbl> <dbl>
## 1 blood jct1 0.437 0.0356
## 2 F1 jct1 0.449 0.0344
## 3 I1 jct1 0.244 0.00875
## 4 N1 jct1 0.381 0.0127
```

Statistics

##

8

We use the one-way anova, because of 4 groups. First tests will check for normal distributed data and homogeneity. Followed by data exploring statistics.

```
#check for normality
for (i in levels(dd8_norm$Sample)){
    subdat <- dd8_norm[dd8_norm$Sample == i,]
    # subdat$normality <- shapiro.test(subdat$concNorm)$p.value
    print(paste(i, shapiro.test(subdat$concNorm)$p.value, sep = ": "))
}

## [1] "blood: 0.129735509517597"
## [1] "F1: 0.82518178713395"
## [1] "T1: 0.518509102187032"
## [1] "N1: 0.183197587004605"

# all tests are insignificant -> samples are normality distributed

#homogeneity of variance
leveneTest(dd8_norm$concNorm, dd8_norm$Sample, center = mean)

## Levene's Test for Homogeneity of Variance (center = mean)

## Df F value Pr(>F)
## group 3 2.5912 0.1252
```

```
#not significant -> variances are not different=there are homogenous
#anova, one way anova
celltype<-aov(dd8_norm$concNorm ~ dd8_norm$Sample, data = dd8_norm)</pre>
summary(celltype)
                   Df Sum Sq Mean Sq F value
##
                                                  Pr(>F)
## dd8 norm$Sample 3 0.07911 0.026368
                                       39.24 0.0000393 ***
## Residuals
                   8 0.00538 0.000672
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
#yes, there is a difference
#post-hoc, bonferroni (smaller sample size)
pairwise.t.test(dd8_norm$concNorm, dd8_norm$Sample, p.adjust.method = "bonferroni")
##
## Pairwise comparisons using t tests with pooled SD
## data: dd8_norm$concNorm and dd8_norm$Sample
##
##
     blood F1
                      I1
## F1 1.0000 -
## I1 0.0001 0.000066 -
## N1 0.1830 0.0780 0.0012
## P value adjustment method: bonferroni
pairwise.t.test(dd8_norm$concNorm, dd8_norm$Sample, p.adjust.method = "BH")
##
## Pairwise comparisons using t tests with pooled SD
## data: dd8_norm$concNorm and dd8_norm$Sample
##
##
     blood
              F1
                        I1
## F1 0.59295 -
## I1 0.000051 0.000051 -
## N1 0.03661 0.01950 0.00038
## P value adjustment method: BH
pairwise.t.test(dd8_norm$concNorm, dd8_norm$Sample, p.adjust.method = "none")
##
   Pairwise comparisons using t tests with pooled SD
##
##
## data: dd8_norm$concNorm and dd8_norm$Sample
##
              F1
##
     blood
                        T1
```

```
## F1 0.59295 -
## I1 0.000017 0.000011 -
## N1 0.03051 0.01300 0.00019
##
## P value adjustment method: none
#linear model (=baseline blood sample, only three tests, no correction)
summary(lm(dd8_norm$concNorm ~dd8_norm$Sample))
##
## Call:
## lm(formula = dd8_norm$concNorm ~ dd8_norm$Sample)
##
## Residuals:
##
         Min
                    1Q
                          Median
                                        3Q
                                                 Max
## -0.032408 -0.015466 -0.000554 0.007537 0.041045
##
## Coefficients:
                     Estimate Std. Error t value
##
                                                      Pr(>|t|)
                                 0.01497 29.191 0.00000000205 ***
## (Intercept)
                      0.43689
## dd8_norm$SampleF1 0.01178
                                 0.02117
                                           0.557
                                                         0.5929
## dd8_norm$SampleI1 -0.19266
                                 0.02117 -9.102 0.00001705075 ***
## dd8 norm$SampleN1 -0.05552
                                 0.02117 - 2.623
                                                         0.0305 *
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.02592 on 8 degrees of freedom
## Multiple R-squared: 0.9364, Adjusted R-squared: 0.9125
## F-statistic: 39.24 on 3 and 8 DF, p-value: 0.00003931
#t-test for comparison
test <- dd8_norm[dd8_norm$Sample %in% c("blood", "N1"),]
t.test(test$concNorm~test$Sample, var.equal = TRUE)
##
##
   Two Sample t-test
## data: test$concNorm by test$Sample
## t = 2.5425, df = 4, p-value = 0.06381
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.00510799 0.11614376
## sample estimates:
## mean in group blood
                          mean in group N1
##
             0.4368895
                                 0.3813716
#statistics for the comparison with the expected 0.5 perfect heterozygosity
stats <- as.data.frame(matrix(nrow = length(levels(dd8_norm$Sample)), ncol = 4))</pre>
names(stats)<- c("Sample", "pval", "padj", "sign")</pre>
for(m in 1:length(levels(dd8_norm$Sample))){
  subdat <- dd8_norm[dd8_norm$Sample == levels(dd8_norm$Sample)[m],]</pre>
```

```
stats$Sample[m]<- levels(dd8_norm$Sample)[m]</pre>
  stats$pval[m]<- t.test(subdat$concNorm, mu = 0.5, alternative = "less", var.equal = TRUE)$p.value
  stats$padj[m] <- p.adjust(stats$pval[m], method = "bonferroni", n = length(levels(dd8_norm$Sample)))</pre>
  #add pval ID
  if(stats$padj[m] < 0.001){</pre>
    stats$sign[m] <- "***"
  }else if (stats$padj[m] < 0.01){</pre>
    stats$sign[m] <- "**"
  }else if (stats$padj[m] < 0.05){</pre>
    stats$sign[m] <- "*"
  }else{
    stats$sign[m] <- "ns"
  # print(paste(levels(dd8_norm$Sample)[m], stats$siqn[m], stats$pval[m], stats$padj[m], sep = ": "))
tibble(stats)
## # A tibble: 4 x 4
##
     Sample
               pval
                          padj sign
     <chr>>
               <dbl>
                         <dbl> <chr>
## 1 blood 0.0459
                    0.184
                               ns
## 2 F1
            0.0613
                      0.245
## 3 I1
            0.000195 0.000780 ***
## 4 N1
            0.00190 0.00759 **
```

Plot

```
#plot ----
plot_norm_stat <- ggplot(dd8_norm,aes(x=Sample, y=concNorm)) +</pre>
 theme_classic(base_size=20) +
  geom_hline(yintercept = 0.5, linetype =2,size=0.5, colour = "black")+
  \# geom_point(aes(col=Sample),position=position_dodge(width=0.7), size=8, alpha=0.6, shape=16)+
  geom_jitter(aes(col=Sample), size=8, alpha=0.6, shape=16, width = 0.1)+
  scale_color_manual(values=colorsForLegend[c(5,9,3,8)])+
  scale_x_discrete(labels = c("Blood", "Fibroblasts", "iPSCs", "NESCs"))+
  stat_summary(fun.data="mean_se", geom="errorbar", width=0.1, size=1)+
  geom_boxplot(aes(x=Sample, y=mean), width=0.4, size=1, colour="black")+
  labs(x="", y = "Fraction of ring chromosome", fill = element_blank()) +
  scale_y_continuous(breaks = seq(0, 0.6, by=0.1), limits = c(0,0.6))+
  annotate(geom = "text", label = "expected fraction for perfect heterozygosity", x=0.2, y = 0.5, size
   plot.subtitle = element_text(colour="black", face="italic", size = 15),
   axis.ticks = element_line(colour="black"),
   axis.text.x = element_text(colour="black", hjust=0.5, vjust=1, angle=0),
   axis.text.y = element_text(colour="black", hjust=0.5),
   legend.title = element_text(colour = "black", size = 15),
    strip.background = element_rect(size=1.5, linetype="solid", colour ="black"),
    strip.text = element_text(size = 14),
    legend.position = "none"
  )+
```

Mosaic state of the ring chromosome i

