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# set random seed start for consistant results
set.seed(123)
# call data file
longformdata <- read.table("201-107_DATAANALYSIS_longform_Rreadable.txt",
  header = T, sep = "\t")
# remove unneeded data fields for analysis
data <- subset(longformdata, select = c("TxSamp", "Tx", "Matrix",
  "Time_hr", "analyte", "Conc_nM"))
# order levels of analyte Pro>CDV>CDVPP for figure
# generation
data$analyte <- ordered(data$analyte, levels = c("Prodrug", "CDV",
  "PP"))
# rename the levels
levels(data$analyte) <- c("Prodrug", "CDV", "CDV-PP")
# order the test articles for figure generation
data$Tx <- ordered(data$Tx, levels = c("NPP669", "NPP666", "NPP663",
  "CDV", "CMX001", "USC505"))
# pull info from TxSamp column to indicate time collected and
# if sample is dose stability (ds), from medium (med), or
# from lysate (lys) and creates a new column 'SampleType'
data$SampleType <- gsub(".*_(T\\d+_\\w+)\\d+_\\w+", "\\1", data$TxSamp)
# volume of lysate in mL
lysate_mL <- 0.333
# volume of media in mL
media_mL <- 10

# the following funtion converts nanomolar to nanomoles
Nanomol <- function(x) {
  nmol <- ifelse(data$Matrix == "media", (x/1000 * media_mL),
    (x/1000 * lysate_mL))
  return(nmol)
}
# use above function to convert concentration to raw
# nanomoles for each sample in a new column
data$Nanomol <- Nanomol(data$Conc_nM)
# convert nanomoles to picomoles for each sample in a new
# column
data$Picomol <- data$Nanomol * 1000

# show structure of data and breakdowns by column
summary(data)

```

```

##           TxSamp           Tx           Matrix           Time_hr
## CDV_T0_ds1_CDV : 1 NPP669:27 lysate:51 Min. : 0.00
## CDV_T0_ds2_CDV : 1 NPP666:27 media :90 1st Qu.:72.00
## CDV_T0_ds3_CDV : 1 NPP663:27          Median :72.00
## CDV_T72_ds1_CDV: 1 CDV :15          Mean :56.68
## CDV_T72_ds2_CDV: 1 CMX001:27        3rd Qu.:72.00
## CDV_T72_ds3_CDV: 1 USC505:18        Max. :72.00
## (Other) :135
##      analyte      Conc_nM      SampleType      Nanomol
## Prodrug:51 Min. : 0.0 Length:141 Min. : 0.000
## CDV :72 1st Qu.: 0.0 Class :character 1st Qu.: 0.000
## CDV-PP :18 Median : 387.6 Mode :character Median : 0.152

```

```
##           Mean   : 558.6           Mean   : 3.104
##           3rd Qu.:1068.8          3rd Qu.: 5.802
##           Max.   :1784.5          Max.    :13.463
##
## Picomol
## Min.      :    0
## 1st Qu.   :    0
## Median    : 152
## Mean      : 3104
## 3rd Qu.   : 5802
## Max.      :13463
##
```

```
# select all data with lysate as the matrix to determine
# analyte per cells
lysate <- data[data$Matrix == "lysate", ]
# call in the cell count data
cellcountsdf <- read.table("201-107_rawcellcounts.txt", header = T,
  sep = "\t")

# the following function converts the raw hempcytometer
# counts to cells per flask
cells_per_flask <- function(rawcells) {
  cellspersflask <- rawcells/4 * 2 * 10000 * 5
  return(cellspersflask)
}

# use above function to add cells per flask column to cell
# count dataframe
cellcountsdf$cellspersflask <- cells_per_flask(cellcountsdf$rawcells)
# calculate the mean cell count by treatment group as a new
# dataframe
avecellcounts <- as.data.frame(cellcountsdf %>% group_by(Tx) %>%
  summarise(meancells = mean(cellspersflask)))
# summary of cell count table
summary(cellcountsdf)
```

```
##      Tx      rawcells      cellspersflask
## CDV   :3   Min.    : 50.00   Min.    :1250000
## CMX001:3   1st Qu.: 69.00   1st Qu.:1725000
## DMSO   :3   Median : 75.00   Median :1875000
## NPP663:3   Mean    : 81.24   Mean    :2030952
## NPP666:3   3rd Qu.: 95.00   3rd Qu.:2375000
## NPP669:3   Max.    :120.00   Max.    :3000000
## USC505:3
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# average counts by group
avecellcounts
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```
##      Tx meancells
## 1    CDV   2583333
## 2 CMX001   1575000
## 3    DMSO   2375000
## 4 NPP663   2075000
## 5 NPP666   2250000
## 6 NPP669   1566667
## 7 USC505   1791667
```

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# combine the lysate data with average cell count data by
# group
lysate <- merge(lysate, avecellcounts, by = "Tx")
# calculate picomole per 106 cells for each sample
lysate$pmolPERmillCell <- lysate$Picomol/(lysate$meancells/1e+06)

# summary of lysate table
summary(lysate)

##           Tx           TxSamp      Matrix      Time_hr      analyte
## NPP669:9   CDV_T72_lys1_CDV: 1 lysate:51   Min.      :72   Prodrug:15
## NPP666:9   CDV_T72_lys1_PP : 1 media : 0   1st Qu.:72   CDV      :18
## NPP663:9   CDV_T72_lys2_CDV: 1           Median :72   CDV-PP   :18
## CDV       :6 CDV_T72_lys2_PP : 1           Mean  :72
## CMX001:9   CDV_T72_lys3_CDV: 1           3rd Qu.:72
## USC505:9   CDV_T72_lys3_PP : 1           Max.   :72
##           (Other)           :45
##           Conc_nM           SampleType           Nanomol           Picomol
## Min.      : 0.0   Length:51   Min.      :0.00000   Min.      : 0.00
## 1st Qu.: 282.0   Class :character 1st Qu.:0.09391   1st Qu.: 93.91
## Median : 449.5   Mode  :character  Median :0.14968   Median :149.68
## Mean      : 709.8           Mean  :0.23636   Mean      :236.36
## 3rd Qu.:1319.8           3rd Qu.:0.43949   3rd Qu.:439.49
## Max.      :1784.5           Max.   :0.59424   Max.      :594.24
##
##           meancells           pmolPERmillCell
## Min.      :1566667   Min.      : 0.00
## 1st Qu.:1575000   1st Qu.: 53.35
## Median :1791667   Median : 84.83
## Mean      :1937745   Mean      :133.69
## 3rd Qu.:2250000   3rd Qu.:204.68
## Max.      :2583333   Max.      :379.30
##
# calculate average picomoles per 106 cells by treatment
# group
avepmolpcell <- lysate %>% group_by(Tx, analyte) %>% summarise(meanper = mean(pmolPERmillCell),
  sd = (sd(pmolPERmillCell)))
# show average picomole per 106 cells by gorup
avepmolpcell

## # A tibble: 17 x 4
## # Groups: Tx [?]
## Tx analyte meanper sd
## <ord> <ord> <dbl> <dbl>
## 1 NPP669 Prodrug 314. 57.2
## 2 NPP669 CDV 325. 35.9
## 3 NPP669 CDV-PP 288. 13.9
## 4 NPP666 Prodrug 132. 6.66
## 5 NPP666 CDV 129. 22.3
## 6 NPP666 CDV-PP 203. 3.13
## 7 NPP663 Prodrug 47.9 6.25
## 8 NPP663 CDV 37.1 5.84
## 9 NPP663 CDV-PP 65.4 3.36

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## 10 CDV      CDV      0.434  0.752
## 11 CDV      CDV-PP    0      0
## 12 CMX001 Prodrug 336.    4.09
## 13 CMX001 CDV      69.5   18.0
## 14 CMX001 CDV-PP   61.4    2.06
## 15 USC505 Prodrug 113.    15.0
## 16 USC505 CDV      69.3   16.8
## 17 USC505 CDV-PP   80.9    2.75

# isolate prodrug samples in lysate for statistical analysis
lys_pro <- lysate[lysate$analyte == "Prodrug", ]
# isolate CDV samples in lysate for statistical analysis
lys_cdv <- lysate[lysate$analyte == "CDV", ]
# isolate CDV-PP samples in lysate for statistical analysis
lys_pp <- lysate[lysate$analyte == "CDV-PP", ]

set.seed(123)
# verify normal distribution
norm_pp <- leveneTest(pmolPERmillCell ~ Tx, lys_pp)
# perform anova comparing picomoles per 106 cells by
# treatment group
aov_pp <- aov(pmolPERmillCell ~ Tx, lys_pp)
# anova results table
summary(aov_pp)

##           Df Sum Sq Mean Sq F value    Pr(>F)
## Tx           5 171988   34398   917.1 4.62e-15 ***
## Residuals    12    450     38
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

# performs Tukey post-hoc test for multiple comparisons and
# adjusts p-values accordingly, same for the below anovas
mct_pp <- TukeyHSD(aov_pp)
# display multiple comparison results
mct_pp

##      Tukey multiple comparisons of means
##      95% family-wise confidence level
##
## Fit: aov(formula = pmolPERmillCell ~ Tx, data = lys_pp)
##
## $Tx
##           diff          lwr          upr          p adj
## NPP666-NPP669 -84.752410 -101.548804  -67.95602 0.0000000
## NPP663-NPP669 -222.411192 -239.207586 -205.61480 0.0000000
## CDV-NPP669    -287.818277 -304.614670 -271.02188 0.0000000
## CMX001-NPP669 -226.433515 -243.229908 -209.63712 0.0000000
## USC505-NPP669 -206.888435 -223.684829 -190.09204 0.0000000
## NPP663-NPP666 -137.658782 -154.455176 -120.86239 0.0000000
## CDV-NPP666    -203.065867 -219.862260 -186.26947 0.0000000
## CMX001-NPP666 -141.681105 -158.477499 -124.88471 0.0000000
## USC505-NPP666 -122.136025 -138.932419 -105.33963 0.0000000
## CDV-NPP663    -65.407084  -82.203478  -48.61069 0.0000002
## CMX001-NPP663  -4.022322  -20.818716   12.77407 0.9612412

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## USC505-NPP663    15.522758   -1.273636   32.31915 0.0763444
## CMX001-CDV       61.384762   44.588368   78.18116 0.0000004
## USC505-CDV       80.929842   64.133448   97.72624 0.0000000
## USC505-CMX001    19.545080    2.748686   36.34147 0.0197853

# anova steps same as above but for CDV and Prodrug
set.seed(123)
norm_cdv <- leveneTest(pmolPERmillCell ~ Tx, lys_cdv)
aov_cdv <- aov(pmolPERmillCell ~ Tx, lys_cdv)
summary(aov_cdv)

##              Df Sum Sq Mean Sq F value    Pr(>F)
## Tx              5 201185   40237    99.5 2.43e-09 ***
## Residuals      12   4853     404
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

mct_cdv <- TukeyHSD(aov_cdv)
mct_cdv

##      Tukey multiple comparisons of means
##      95% family-wise confidence level
##
## Fit: aov(formula = pmolPERmillCell ~ Tx, data = lys_cdv)
##
## $Tx
##              diff            lwr            upr            p adj
## NPP666-NPP669 -195.9618667 -251.11258 -140.811156 0.0000006
## NPP663-NPP669 -287.9259759 -343.07669 -232.775265 0.0000000
## CDV-NPP669    -324.5740258 -379.72474 -269.423315 0.0000000
## CMX001-NPP669 -255.4620952 -310.61281 -200.311385 0.0000000
## USC505-NPP669 -255.7316093 -310.88232 -200.580899 0.0000000
## NPP663-NPP666  -91.9641092 -147.11482  -36.813399 0.0012511
## CDV-NPP666    -128.6121591 -183.76287  -73.461448 0.0000535
## CMX001-NPP666  -59.5002286 -114.65094  -4.349518 0.0320161
## USC505-NPP666  -59.7697426 -114.92045  -4.619032 0.0311408
## CDV-NPP663    -36.6480499  -91.79876   18.502661 0.2917495
## CMX001-NPP663   32.4638807  -22.68683   87.614591 0.4067818
## USC505-NPP663   32.1943666  -22.95634   87.345077 0.4150204
## CMX001-CDV      69.1119306   13.96122  124.262641 0.0119223
## USC505-CDV      68.8424165   13.69171  123.993127 0.0122555
## USC505-CMX001   -0.2695141  -55.42022   54.881197 1.0000000

set.seed(123)
norm_pro <- leveneTest(pmolPERmillCell ~ Tx, lys_pro)
aov_pro <- aov(pmolPERmillCell ~ Tx, lys_pro)
summary(aov_pro)

##              Df Sum Sq Mean Sq F value    Pr(>F)
## Tx              4 198439   49610   68.88 3.07e-07 ***
## Residuals      10   7202     720
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

mct_pro <- TukeyHSD(aov_pro)
mct_pro
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```

## Tukey multiple comparisons of means
## 95% family-wise confidence level
##
## Fit: aov(formula = pmolPERmillCell ~ Tx, data = lys_pro)
##
## $Tx
##              diff          lwr          upr          p adj
## NPP666-NPP669 -181.66717 -253.783338 -109.55099 0.0000656
## NPP663-NPP669 -266.20031 -338.316480 -194.08414 0.0000020
## CMX001-NPP669  22.04638  -50.069795   94.16255 0.8469444
## USC505-NPP669 -200.83179 -272.947961 -128.71562 0.0000270
## NPP663-NPP666  -84.53314 -156.649314  -12.41697 0.0207850
## CMX001-NPP666  203.71354  131.597371  275.82971 0.0000238
## USC505-NPP666  -19.16462  -91.280795   52.95155 0.8998891
## CMX001-NPP663  288.24669  216.130513  360.36286 0.0000010
## USC505-NPP663   65.36852   -6.747653  137.48469 0.0804850
## USC505-CMX001 -222.87817 -294.994338 -150.76199 0.0000105

# figure building label columns for statistical comparisons
# from Tukey tests
stats_labs <- data.frame(Tx = factor(rep(c("NPP669", "NPP666",
    "NPP663", "CDV", "CMX001", "USC505"), 3)), analyte = factor(c(rep("Prodrug",
    6), rep("CDV", 6), rep("CDV-PP", 6))), labs = factor(c("bcf",
    "ace", "abe", "sdgsna", "bcf", "ae", "bcdef", "acdef", "ab",
    "abef", "abd", "abd", "bcdef", "acdef", "abd", "abcef", "abdf",
    "abde")))

# annotate the column labels with letter for statistical
# comparisons
ann_text <- data.frame(Tx = factor(c("NPP669", "NPP666", "NPP663",
    "CDV", "CMX001", "USC505")), labs = factor(c("NPP-669\n(a)",
    "NPP-666\n(b)", "NPP-663\n(c)", "CDV\n(d)", "CMX-001\n(e)",
    "USC-505\n(f)")))

# label CDV-Prodrug as 'na' since there isnt a prodrug
noCDVpro <- data.frame(Tx = "CDV", analyte = "Prodrug", y = 20,
    lab = "italic(na)")

# merge label dataframe with avepmolcell dataframe for calling
# figure labels
labels <- merge(stats_labs, avepmolpcell, by = c("Tx", "analyte"),
    all = T)

set.seed(123)
# call color palette package for figure
library(RColorBrewer)
# select 3 complementary colors for 3 analytes from 'Pastel1'
# color palette
myColors <- brewer.pal(3, "Pastel1")
# assign a color to an analyte
names(myColors) <- levels(lysate$analyte)
# instruct ggplot how to color columns
colScale <- scale_fill_manual(name = "analyte", values = myColors)

# figure building: ggplot calls data and variables to

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```

# display, stat_summary selects bar chart display and
# errorbars using mean + SD, scale_x and geom_text add labels
# to x axis, scale_y does same for y axis, facet_grid breaks
# up figure by analyte and places the facet label on the
# left, geom_text labels CDV-Prodrug as na, colScale colors
# the columns and legend, theme_bw() instructs how figure
# background looks, remaining theme() calls remove the legend
# bolds the facet labels, titles the x axis and y axis
p <- ggplot(lysate, aes(Tx, pmolPERmillCell, fill = analyte)) +
  stat_summary(geom = "bar", fun.y = mean, position = "dodge",
    color = "black") + stat_summary(geom = "errorbar", fun.ymin = function(x) mean(x) -
    sd(x), fun.ymax = function(x) mean(x) + sd(x), position = "dodge",
    width = 0.5) + scale_x_discrete(breaks = ann_text$Tx, labels = ann_text$labs) +
  geom_text(labels, mapping = aes(Tx, y = (meanper + sd + 20),
    label = labs)) + scale_y_continuous(limits = c(0, 425),
    expand = c(0, 0)) + facet_grid(analyte ~ ., switch = "both") +
  geom_text(noCDVpro, mapping = aes(x = Tx, y = y, label = lab),
    parse = T) + colScale + theme_bw() + theme(legend.position = "none") +
  theme(strip.text = element_text(face = "bold")) + labs(x = "Treatment",
    y = bquote(paste("Picomoles Analyte per 10"6", " HFF-1 cells"))))

```