

Bio-Rad Take Home Project

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```
library(tidyverse)
library(data.table)
library(here)
library(gridExtra)
library(gt)
library(ggtext)
library(ggdark)
```

```
count_tbl_filt <- data.table::fread(here("data", "count_tbl_filt.csv"))
```

Data set dimensions (rows, columns)

```
dim(count_tbl_filt)
```

```
## [1] 128804 5580
```

First 4 rows and 2 columns of count_tbl_filt.

```
count_tbl_filt[1:4, 1:2]
```

```
##                V1 alignments.possorted.tagged_BC00001_N01
## 1: hg19_chr1:713971-714221                                0
## 2: hg19_chr1:714480-714730                                0
## 3: hg19_chr1:762720-762970                                0
## 4: hg19_chr1:825991-826241                                0
```

Create useful variables from the gene ID (V1) column.

```
count_tbl_filt_2 <- count_tbl_filt %>%
  separate(V1, c("org_chr", "bp_loc"), ":", remove = FALSE) %>%
  separate(org_chr, c("ref_genome", "chromosome"), "_") %>%
  mutate(chromosome = str_remove(chromosome, 'chr')) %>%
  mutate(chromosome = factor(chromosome,
                             levels = c('1', '2', '3', '4', '5', '6', '7', '8', '9', '10',
                                           '11', '12', '13', '14', '15', '16', '17', '18',
                                           '19', '20', '21', '22', 'X'),
                             ordered = TRUE)) %>%
  separate(bp_loc, c("from_loc", "to_loc")) %>%
  mutate_at(c("from_loc", "to_loc"), as.numeric) %>%
  select(ref_genome, chromosome, from_loc, to_loc, drops_with_gene, 6:last_col())

rm(count_tbl_filt)
```

Now we can group_by, filter, ... count_tbl_filt_2 by reference genome, chromosome, or locus. The drops_with_gene variable was created earlier when filtering out genes present in < 1% of droplets.

First 4 rows and 5 columns of count_tbl_filt_2.

```
count_tbl_filt_2[1:4, 1:5]
```

```
##   ref_genome chromosome from_loc to_loc drops_with_gene
## 1:      hg19          1   713971 714221             630
## 2:      hg19          1   714480 714730              88
## 3:      hg19          1   762720 762970             189
## 4:      hg19          1   825991 826241              56
```

Find total gene read counts for both reference genomes in each droplet.

```
count_tbl_sum <- count_tbl_filt_2 %>%
  group_by(ref_genome) %>%
  summarise(across(5:last_col(), .fns = sum))
```

View total read counts for both reference genomes in first 2 droplets.

```
count_tbl_sum[,1:3]
```

```
## # A tibble: 2 x 3
##   ref_genome alignments.possorted.tagged_BC00001_N01 alignments.possorted.tagge~
##   <chr>                                <int>                                <int>
## 1 hg19                                  51                                  146
## 2 mm10                                 962                                 3724
```

Reshape count_tbl_sum so droplets are observations (rows). Filter to retain droplets that have > 800 read counts in at least 1 reference genome.

```
count_tbl_sum_long <- count_tbl_sum %>%
  pivot_longer(cols = -ref_genome) %>%
  pivot_wider(names_from = ref_genome, values_from = value) %>%
  rename(total_hg19_counts = hg19, total_mm10_counts = mm10) %>%
  filter(total_hg19_counts > 800 | total_mm10_counts > 800)
```

View first 4 rows in count_tbl_sum_long.

```
count_tbl_sum_long[1:4,]
```

```
## # A tibble: 4 x 3
##   name                                total_hg19_counts total_mm10_counts
##   <chr>                                <int>                                <int>
## 1 alignments.possorted.tagged_BC00001_N01             51              962
## 2 alignments.possorted.tagged_BC00002_N02            146             3724
## 3 alignments.possorted.tagged_BC00003_N03              78             2067
## 4 alignments.possorted.tagged_BC00004_N03           12796             149
```

Create a summary table to tally unique genes expressed in each droplet. This is different than total counts. If a gene in the count_tbl is >= 1, count_tbl_tally will count it as 1.

```
count_tbl_tally <- count_tbl_filt_2 %>%
  group_by(ref_genome) %>%
  summarise(across(5:last_col(), ~ length(which(.x != 0))))
```

Create `drops_filter` to get names of droplets with > 800 read counts in at least 1 reference genome. Then use `drops_filter` to filter `count_tbl_tally`.

```
drops_filter <- count_tbl_sum_long %>% pull(name)

count_tbl_tally_long <- count_tbl_tally %>%
  pivot_longer(cols = -ref_genome) %>%
  pivot_wider(names_from = ref_genome, values_from = value) %>%
  rename(unique_hg19_genes = hg19, unique_mm10_genes = mm10) %>%
  filter(name %in% drops_filter)
```

View first 4 rows in `count_tbl_tally_long`.

```
count_tbl_tally_long[1:4,]
```

```
## # A tibble: 4 x 3
##   name                                unique_hg19_genes unique_mm10_genes
##   <chr>                                <int>             <int>
## 1 alignments.possorted.tagged_BC00001_N01          51             935
## 2 alignments.possorted.tagged_BC00002_N02         144            3554
## 3 alignments.possorted.tagged_BC00003_N03          78            1935
## 4 alignments.possorted.tagged_BC00004_N03       10853             141
```

Join `count_tbl_sum_long` and `count_tbl_tally_long` summary tables. Create 2 variables for the ratios of `unique_genes/total_counts`, one for each reference genome.

```
count_tbl_join <- count_tbl_sum_long %>%
  left_join(count_tbl_tally_long, by = "name") %>%
  mutate(hg19_unique_count_ratio = log10(unique_hg19_genes)/log10(total_hg19_counts),
         mm10_unique_count_ratio = log10(unique_mm10_genes)/log10(total_mm10_counts))
```

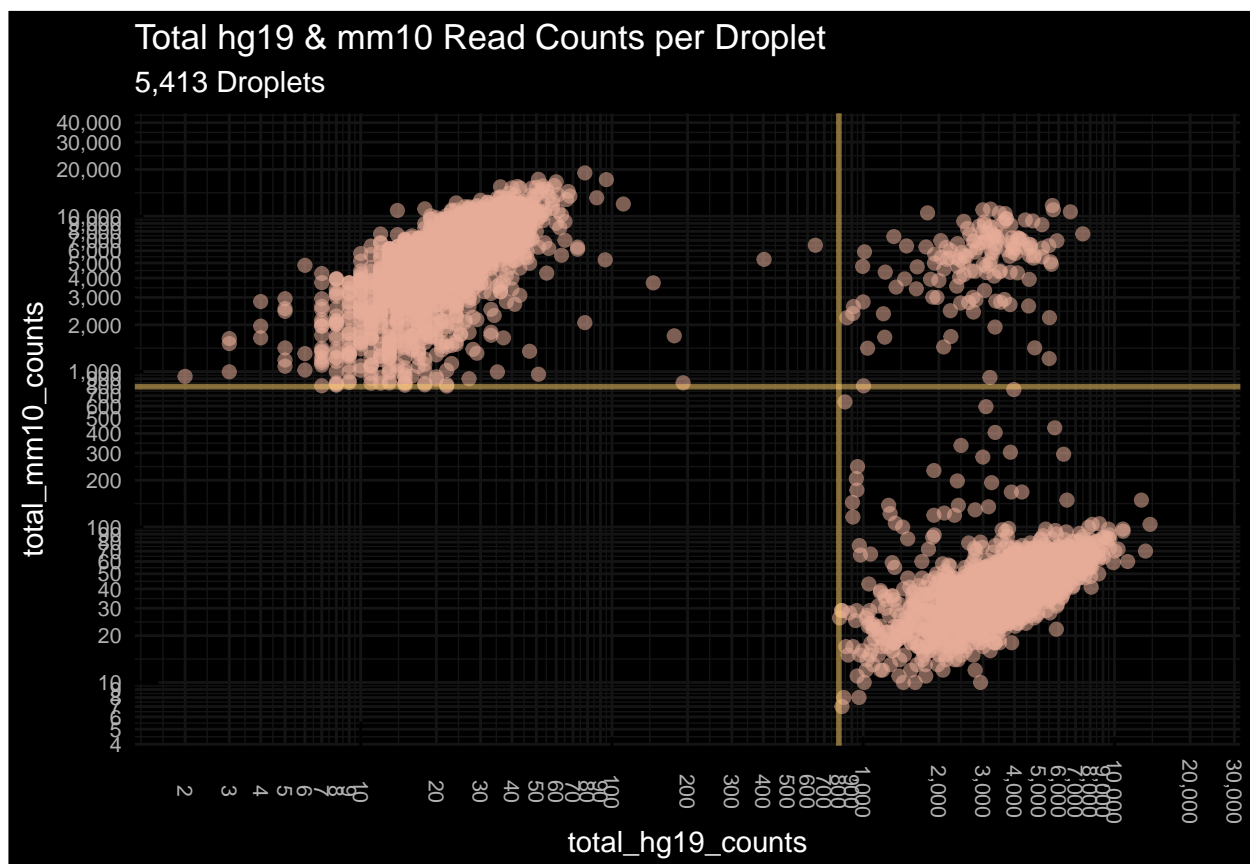
Add a plot here of $\log_{10}(\text{hg19 Counts})$ vs $\log_{10}(\text{mm10 Counts})$, no grouping. Just to show its easy to identify Mouse + Human doublets. Count those and use that number to find the number of expected Human + Human and Mouse + Doublets.

```
count_tbl_join %>%
  ggplot(aes(x = total_hg19_counts, y = total_mm10_counts)) +
  geom_point(alpha = 0.55, size = 2, color = "#e7ad99") +
  scale_y_log10(breaks = scales::breaks_log(n = 38),
               labels = scales::number_format(accuracy = 1, big.mark = ","),
               limit = c(6, 30000)) +
  scale_x_log10(breaks = scales::breaks_log(n = 40),
               labels = scales::number_format(accuracy = 1, big.mark = ","),
               limit = c(2, 20000)) +
  geom_hline(yintercept = c(800),
            color = "#FFCF6A",
            alpha = 0.5,
```

```

      size      = 1) +
geom_vline(xintercept = c(800),
           color      = "#FFCF6A",
           alpha      = 0.5,
           size      = 1) +
ggdark::dark_theme_minimal() +
theme(axis.text.x    = element_text(angle = -90, vjust = 0.5, size = rel(0.9)),
      axis.text.y    = element_text(size = rel(0.9)),
      title          = element_text(size = rel(1.0)),
      legend.position = c(0.4, 0.3),
      legend.text     = element_text(size = rel(0.8))) +
annotation_logticks(scaled = FALSE) +
labs(title    = "Total hg19 & mm10 Read Counts per Droplet",
     subtitle = "5,413 Droplets") +
guides(shape = "none")

```



We now need to identify multiplets (droplets containing multiple cells) Multiplets containing a human and mouse cell are easiest to identify: Droplets with total read counts for both reference genomes > 800 or all droplets in quadrant I.

Count droplets in each quadrant and use those counts to find expected number of multiplets in each quadrant.

```

quad_1_count <- count_tbl_join %>%
  filter(total_hg19_counts > 800 & total_mm10_counts > 800) %>%
  nrow()

quad_4_count <- count_tbl_join %>%

```

```

  filter(total_hg19_counts > 800 & total_mm10_counts < 800) %>%
  nrow()

quad_2_count <- count_tbl_join %>%
  filter(total_hg19_counts < 800 & total_mm10_counts > 800) %>%
  nrow()

ratio_quad_2_4 <- quad_2_count/quad_4_count
ratio_quad_4_2 <- quad_4_count/quad_2_count

expect_quad_2_multi_count <- round(quad_1_count*ratio_quad_2_4, dig = 0)
expect_quad_4_multi_count <- round(quad_1_count*ratio_quad_4_2, dig = 0)

```

`expect_quad_2_multi_count` is the expected n doublets in quadrant II. Droplets containing 2+ human OR 2+ mouse cells can be identified by having lower ratio of unique genes to total read counts in the hg19 or mm10 reference genome, respectively. Droplets containing 1+ human AND 1+ mouse cells will have typical ratios of unique genes to total read counts. Find the value for nth lowest `unique_count_ratio`. Droplets with values \leq to the nth lowest `unique_count_ratio` will be classified as multiplets. Remember to filter counts for quadrant II and IV.

```

hg19_unique_count_ratio_cutoff <- count_tbl_join %>%
  filter(total_hg19_counts > 800) %>%
  select(hg19_unique_count_ratio) %>%
  arrange(hg19_unique_count_ratio) %>%
  pluck(1, expect_quad_4_multi_count)

mm10_unique_count_ratio_cutoff <- count_tbl_join %>%
  filter(total_mm10_counts > 800) %>%
  select(mm10_unique_count_ratio) %>%
  arrange(mm10_unique_count_ratio) %>%
  pluck(1, expect_quad_2_multi_count)

```

Create new variables to label multiplets in the different quadrants:

- Logical variables to label Human + Human in quadrant IV (`hg19_multiplet`), Mouse + Mouse in quadrant II (`mm10_multiplet`), and Human + Mouse in quadrant I (`hg19_mm10_multiplet`).
- Logical variable that labels any multiplet droplet (`Multiplet`).
- Character variable that labels the type of multiplet (`Multiplet_Type`).
- Character variable that labels multiplet type and singlet type (`Droplet_Contents`).

```

count_tbl_summary <- count_tbl_join %>%
  mutate(
    hg19_multiplet = ifelse(hg19_unique_count_ratio <= hg19_unique_count_ratio_cutoff
      & total_hg19_counts > 800, TRUE, FALSE),
    mm10_multiplet = ifelse(mm10_unique_count_ratio <= mm10_unique_count_ratio_cutoff
      & total_mm10_counts > 800, TRUE, FALSE),
    hg19_mm10_multiplet = ifelse(total_hg19_counts > 800
      & total_mm10_counts > 800, TRUE, FALSE),
    Multiplet = ifelse(hg19_multiplet
      | mm10_multiplet

```

```

      | hg19_mm10_multiplet, TRUE, FALSE),
  Multiplet_Type = ifelse(hg19_mm10_multiplet,
    "Human + Mouse",
    ifelse(Multiplet == TRUE & total_hg19_counts > 800
      & total_mm10_counts < 800, "Human + Human",
      ifelse(Multiplet == TRUE & total_mm10_counts > 800
        & total_hg19_counts < 800, "Mouse + Mouse",
        "Singlet"))),
  Droplet_Contents = ifelse(Multiplet_Type == "Human + Mouse", "Human + Mouse",
    ifelse(Multiplet_Type == "Human + Human", "Human + Human",
      ifelse(Multiplet_Type == "Mouse + Mouse",
        "Mouse + Mouse",
        ifelse(Multiplet_Type == "Singlet"
          & total_hg19_counts < 800, "Mouse",
          "Human")))))

```

View new variables in count_tbl_summary.

```
glimpse(count_tbl_summary)
```

```

## Rows: 5,413
## Columns: 13
## $ name <chr> "alignments.possorted.tagged_BC00001_N01", "al~
## $ total_hg19_counts <int> 51, 146, 78, 12796, 8706, 49, 63, 39, 5323, 38~
## $ total_mm10_counts <int> 962, 3724, 2067, 149, 105, 11183, 5610, 10488,~
## $ unique_hg19_genes <int> 51, 144, 78, 10853, 7761, 48, 59, 35, 4829, 34~
## $ unique_mm10_genes <int> 935, 3554, 1935, 141, 104, 9559, 5208, 9176, 9~
## $ hg19_unique_count_ratio <dbl> 1.0000000, 0.9972323, 1.0000000, 0.9825851, 0.~
## $ mm10_unique_count_ratio <dbl> 0.9958556, 0.9943175, 0.9913555, 0.9889714, 0.~
## $ hg19_multiplet <lgl> FALSE, FALSE, FALSE, TRUE, TRUE, FALSE, FALSE,~
## $ mm10_multiplet <lgl> FALSE, FALSE, FALSE, FALSE, FALSE, FALSE, FALS~
## $ hg19_mm10_multiplet <lgl> FALSE, FALSE, FALSE, FALSE, FALSE, FALSE, FALS~
## $ Multiplet <lgl> FALSE, FALSE, FALSE, TRUE, TRUE, FALSE, FALSE,~
## $ Multiplet_Type <chr> "Singlet", "Singlet", "Singlet", "Human + Huma~
## $ Droplet_Contents <chr> "Mouse", "Mouse", "Mouse", "Human + Human", "H~

```

Make sure the number of expected multiplets we calculated in expect_quad_4_multi_count and expect_quad_2_multi_count match the number of TRUE in count_tbl_summary\$hg19_multiplet and count_tbl_summary\$mm10_multiplet.

```

expect_quad_4_multi_count == count_tbl_summary %>%
  filter(hg19_multiplet == TRUE) %>%
  nrow()

```

```
## [1] TRUE
```

```

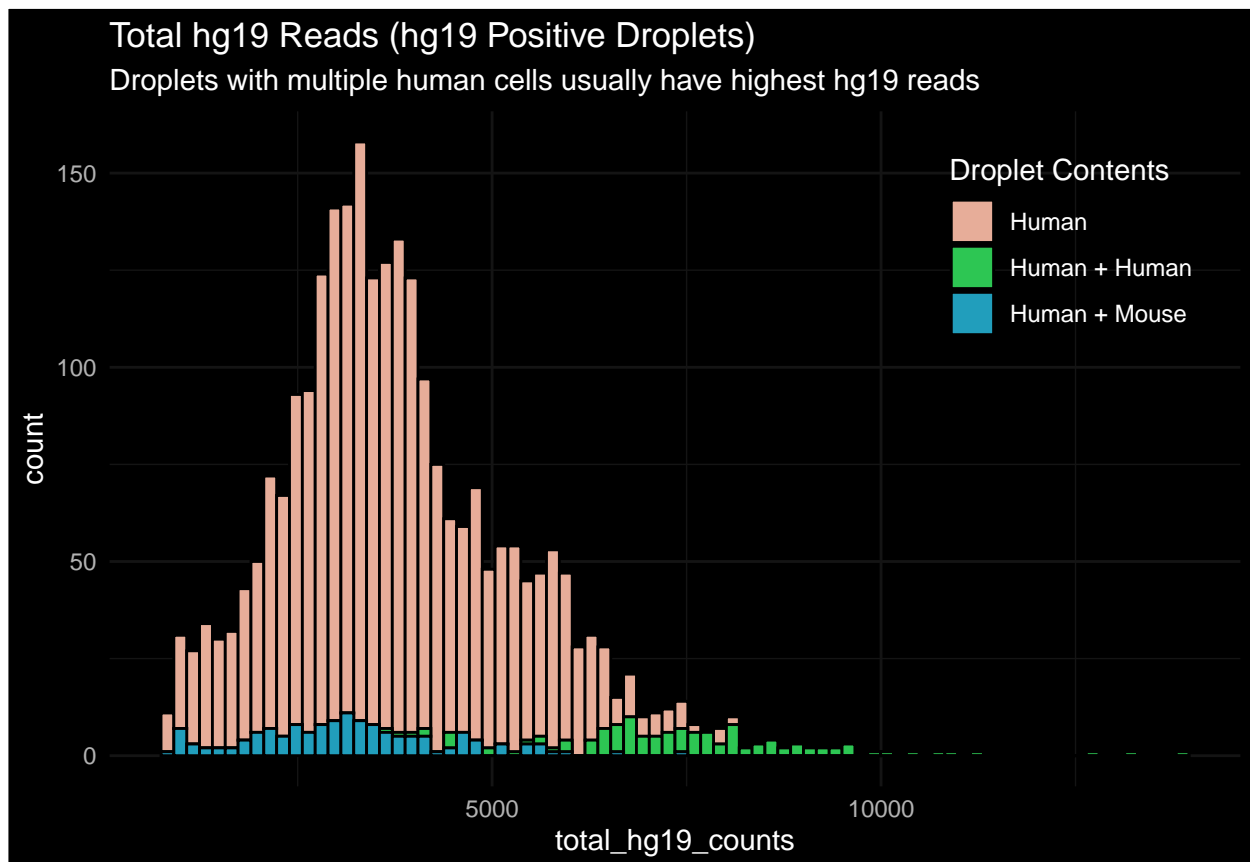
expect_quad_2_multi_count == count_tbl_summary %>%
  filter(mm10_multiplet == TRUE) %>%
  nrow()

```

```
## [1] TRUE
```

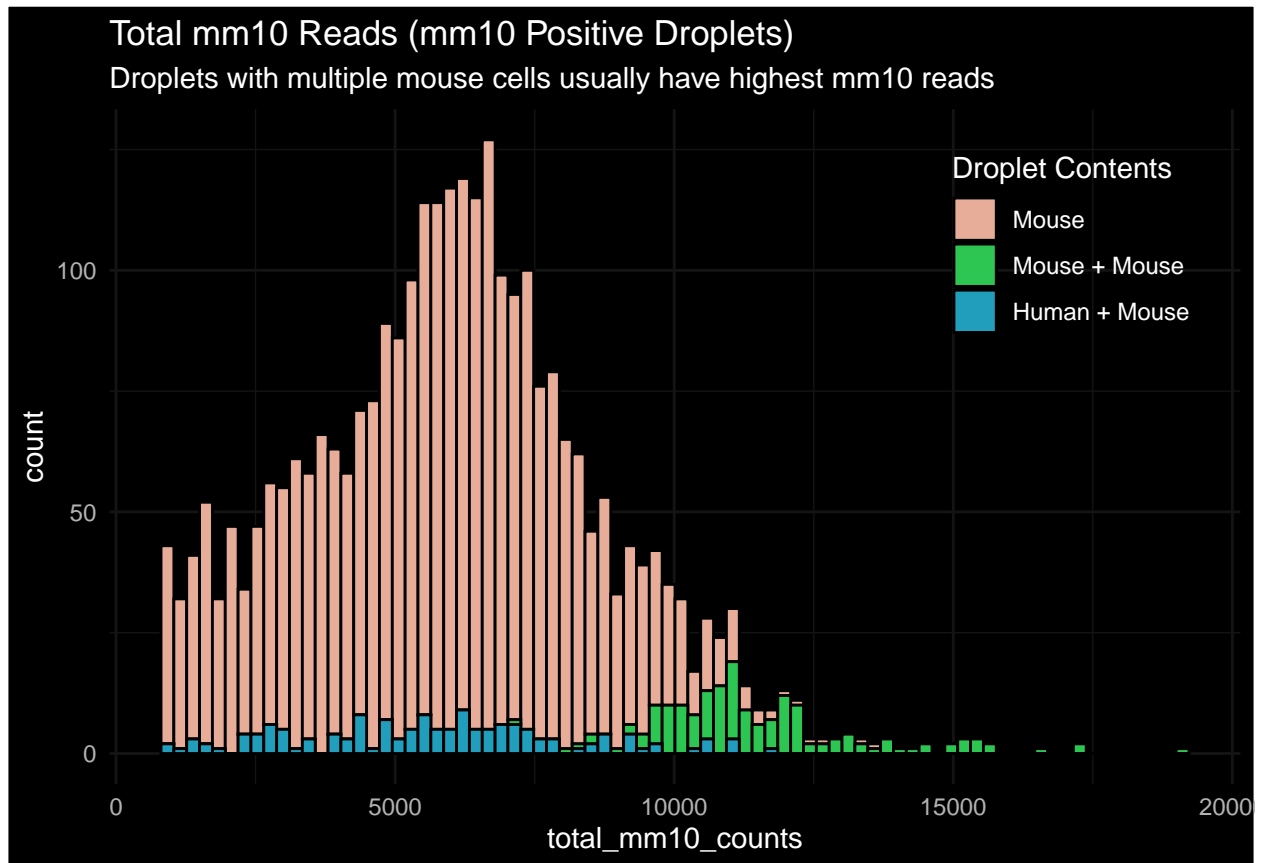
Droplets with 2+ Human OR 2+ Mouse cells typically have higher read counts per droplet for that reference genome. Droplets with 1+ Human AND 1+ Mouse cells have typical read counts for each reference genome. View histogram of hg19 and mm10 reads on separate plots.

```
count_tbl_summary %>%
  filter(total_hg19_counts > 800) %>%
  ggplot(aes(x = total_hg19_counts, fill = Droplet_Contents)) +
  geom_histogram(bins = 80, color = "black") +
  scale_fill_manual(values = c("#e7ad99", "#2dc653", "#219ebc")) +
  ggdark::dark_theme_minimal() +
  labs(title = "Total hg19 Reads (hg19 Positive Droplets)",
       subtitle = "Droplets with multiple human cells usually have highest hg19 reads",
       fill = "Droplet Contents") +
  theme(legend.position = c(0.85, 0.8),
        legend.text = element_text(size = rel(0.8)))
```



```
count_tbl_summary %>%
  filter(total_mm10_counts > 800) %>%
  ggplot(aes(x = total_mm10_counts,
             fill = factor(Droplet_Contents, levels = c("Mouse",
                                                         "Mouse + Mouse",
                                                         "Human + Mouse")))) +
  geom_histogram(bins = 80, color = "black") +
  scale_fill_manual(values = c("#e7ad99", "#2dc653", "#219ebc")) +
  ggdark::dark_theme_minimal() +
```

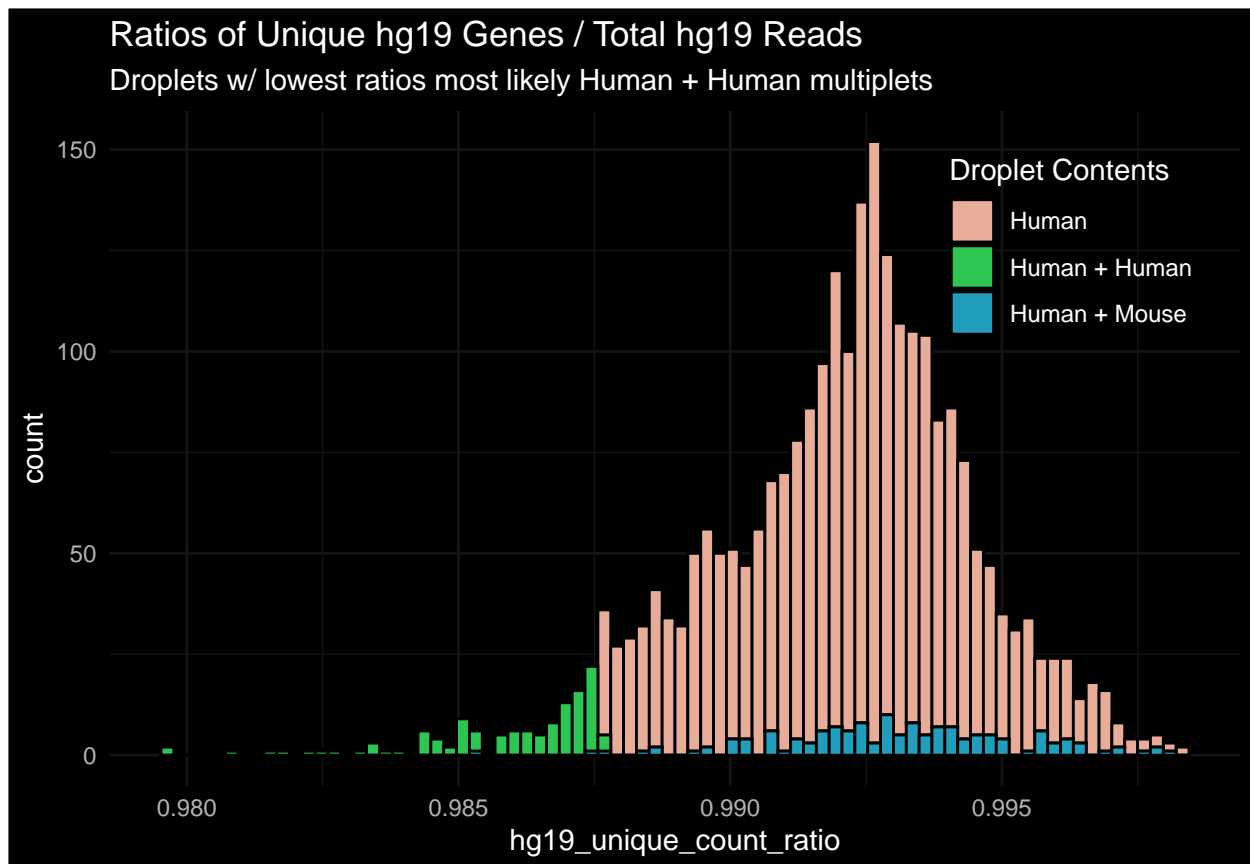
```
labs(title = "Total mm10 Reads (mm10 Positive Droplets)",
     subtitle = "Droplets with multiple mouse cells usually have highest mm10 reads",
     fill = "Droplet Contents") +
theme(legend.position = c(0.85, 0.8),
     legend.text = element_text(size = rel(0.8)))
```



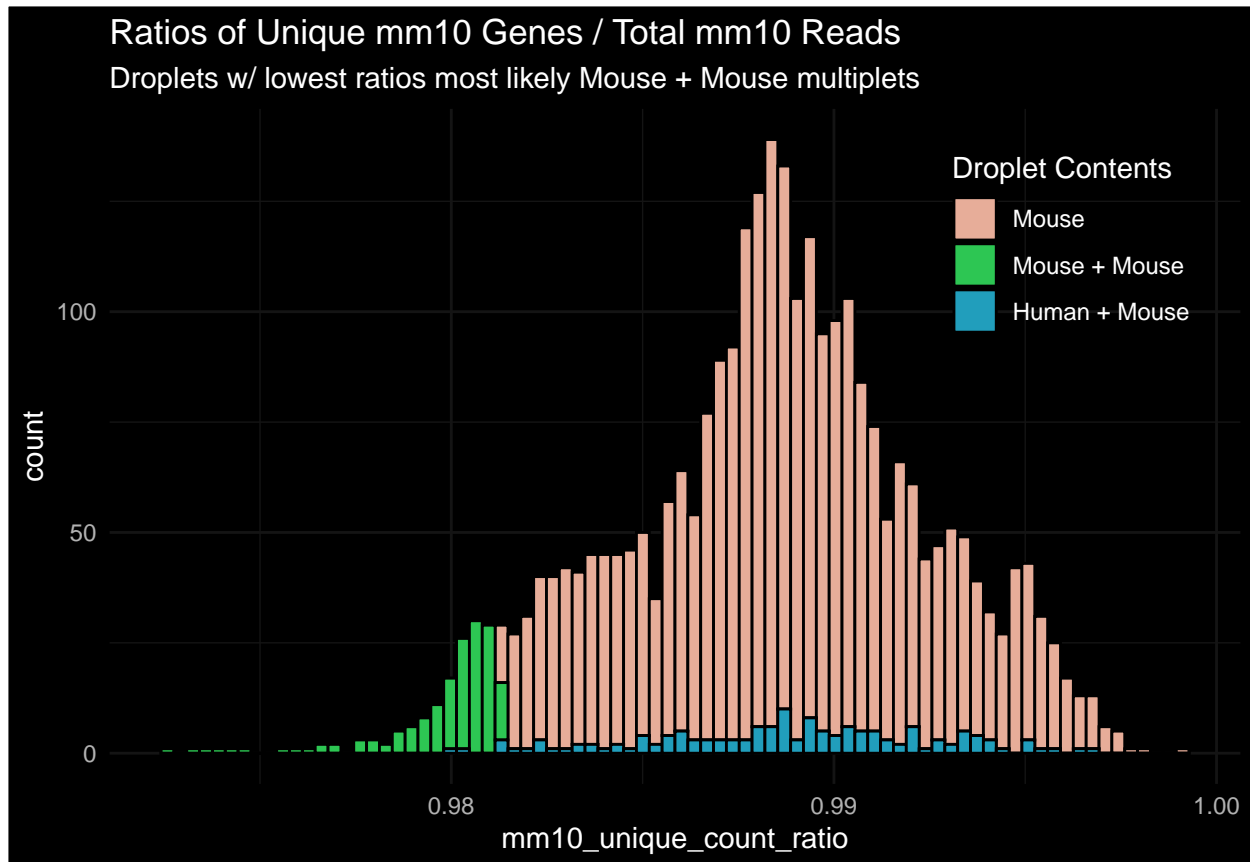
Droplets with 2+ Human OR 2+ Mouse cells typically have the lowest unique gene count to total read count ratios for that reference genome. Droplets with 1+ Human AND 1+ Mouse cells have typical ratios for each reference genome.

View histogram of hg19 and mm10 ratios on separate plots.

```
count_tbl_summary %>%
  filter(total_hg19_counts > 800) %>%
  ggplot(aes(x = hg19_unique_count_ratio, fill = Droplet_Contents)) +
  geom_histogram(bins = 80, color = "black") +
  scale_fill_manual(values = c("#e7ad99", "#2dc653", "#219ebc")) +
  ggdark::dark_theme_minimal() +
  labs(title = "Ratios of Unique hg19 Genes / Total hg19 Reads",
       subtitle = "Droplets w/ lowest ratios most likely Human + Human multiplets",
       fill = "Droplet Contents") +
  theme(legend.position = c(0.85, 0.8),
       legend.text = element_text(size = rel(0.8)))
```

```
count_tbl_summary %>%
  filter(total_mm10_counts > 800) %>%
  ggplot(aes(x = mm10_unique_count_ratio,
             fill = factor(Droplet_Contents, levels = c("Mouse",
                                                         "Mouse + Mouse",
                                                         "Human + Mouse")))) +
  geom_histogram(bins = 80, color = "black") +
  scale_fill_manual(values = c("#e7ad99", "#2dc653", "#219ebc")) +
  ggdark::dark_theme_minimal() +
  labs(title = "Ratios of Unique mm10 Genes / Total mm10 Reads",
       subtitle = "Droplets w/ lowest ratios most likely Mouse + Mouse multiplets",
       fill = "Droplet Contents") +
  theme(legend.position = c(0.85, 0.8),
       legend.text = element_text(size = rel(0.8)))
```



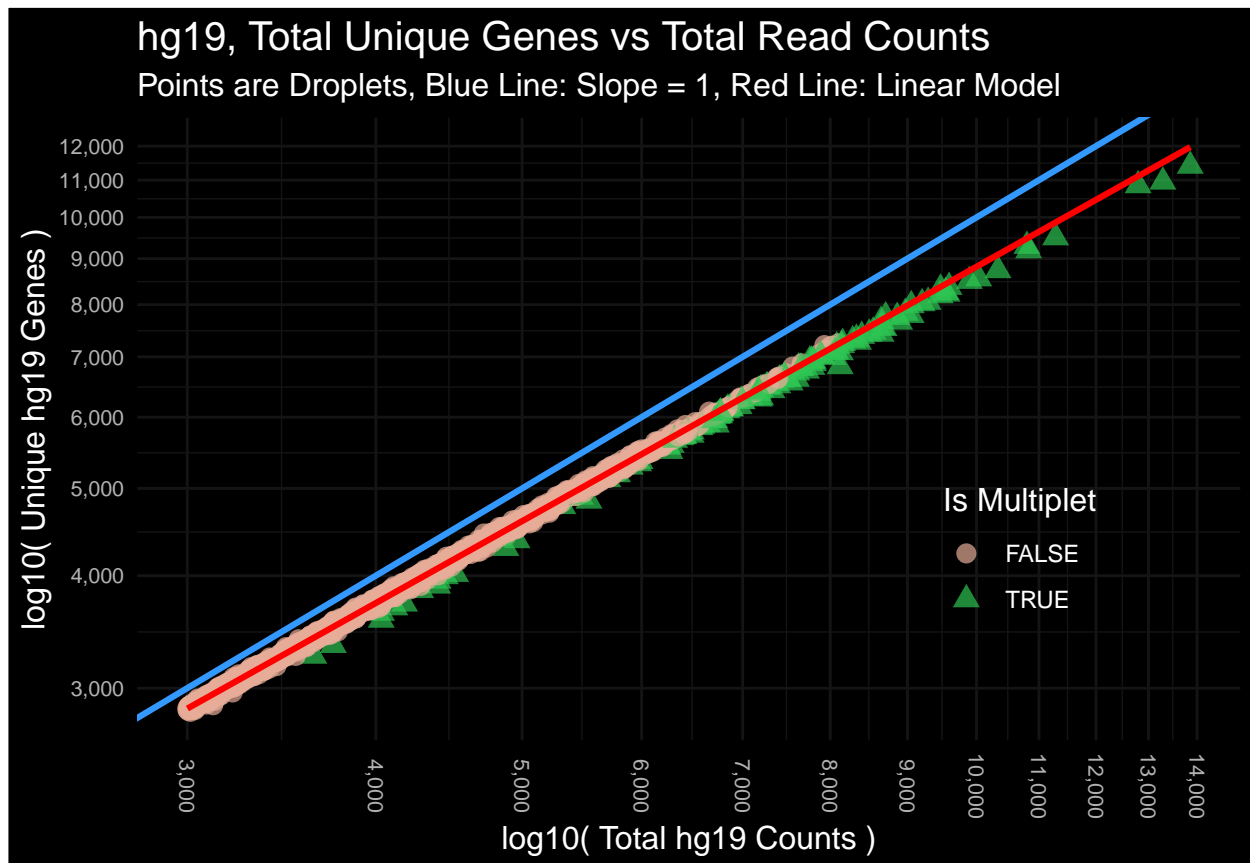
Another view of how unique gene count to total gene count ratios and total read counts are good indicators of multipliers. Fit a linear model on `total_hg19_counts` for low count droplets, and see how high count droplets deviate from that model.

```
count_tbl_summary %>%
  filter(total_hg19_counts > 3000) %>%
  ggplot(aes(x = total_hg19_counts, y = unique_hg19_genes, color = hg19_multipler)) +
    geom_point(aes(shape = hg19_multipler), alpha = 0.7, size = 3.2) +
    geom_smooth(data = count_tbl_summary %>% filter(total_hg19_counts < 6000
                                                    & total_hg19_counts > 3000),
               method = "lm",
               fullrange = TRUE,
               color = "red",
               size = 1.1) +
    scale_color_manual(values = c("#e7ad99", "#2dc653")) +
    scale_y_log10(breaks = scales::breaks_log(n = 12),
                  labels = scales::number_format(accuracy = 1, big.mark = ",")) +
    scale_x_log10(breaks = scales::breaks_log(n = 12),
                  labels = scales::number_format(accuracy = 1, big.mark = ",")) +
    geom_abline(slope = 1,
                color = "#3399FF",
                size = 1.1) +
    ggdark::dark_theme_minimal() +
    theme(axis.text.x = element_text(angle = -90, vjust = 0.5, size = rel(0.9)),
          axis.text.y = element_text(size = rel(0.9)),
          title = element_text(size = rel(1.1)),
```

```

legend.position = c(0.8, 0.3)) +
labs(title = "hg19, Total Unique Genes vs Total Read Counts",
      subtitle = "Points are Droplets, Blue Line: Slope = 1, Red Line: Linear Model",
      x = "log10( Total hg19 Counts )",
      y = "log10( Unique hg19 Genes )",
      color = "Is Multiplet",
      shape = "Is Multiplet")

```



Same for total_mm10_counts.

```

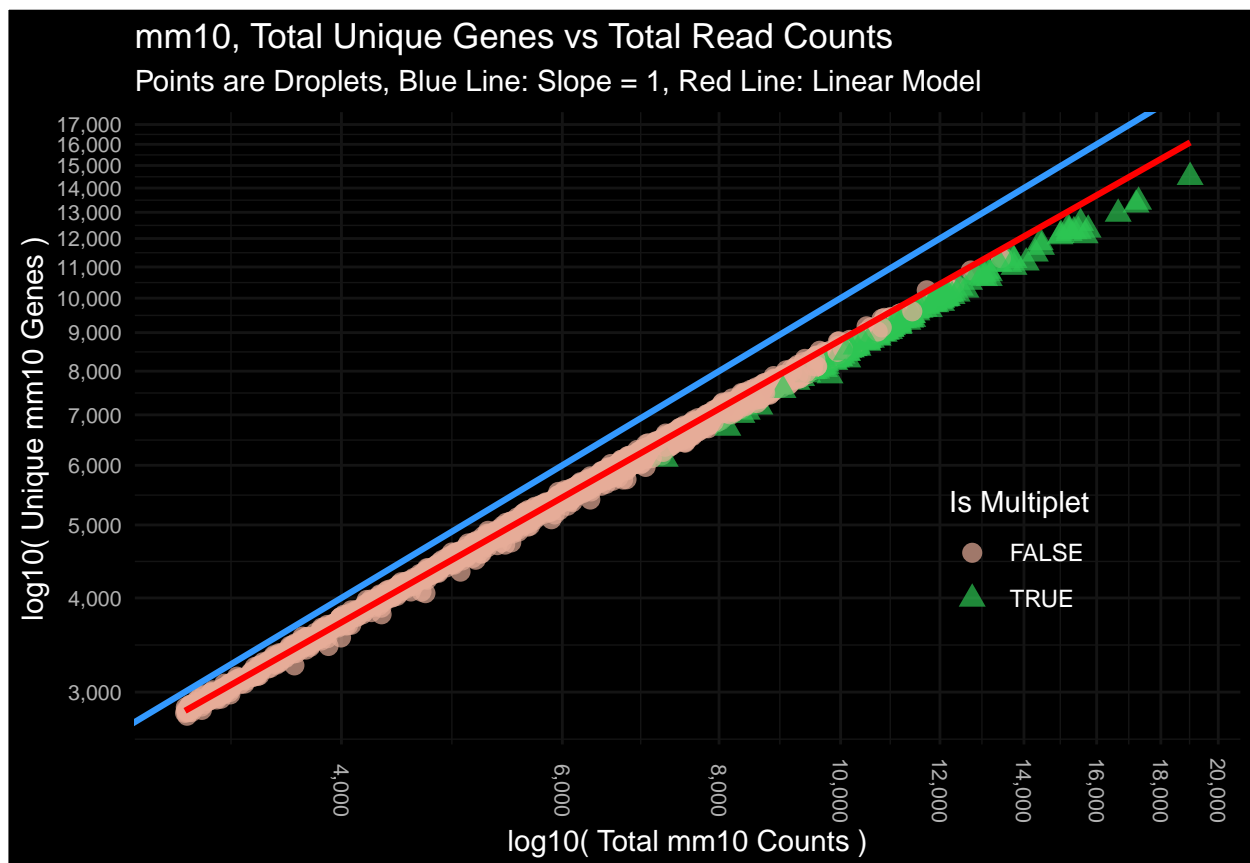
count_tbl_summary %>%
  filter(total_mm10_counts > 3000) %>%
  ggplot(aes(x = total_mm10_counts, y = unique_mm10_genes, color = mm10_multiplet)) +
    geom_point(aes(shape = mm10_multiplet), alpha = 0.7, size = 3.2) +
    geom_smooth(data = count_tbl_summary %>% filter(total_mm10_counts < 6000
                                                    & total_mm10_counts > 3000),
               method = "lm",
               fullrange = TRUE,
               color = "red",
               size = 1.1) +
    scale_color_manual(values = c("#e7ad99", "#2dc653")) +
    scale_y_log10(breaks = scales::breaks_log(n = 12),
                  labels = scales::number_format(accuracy = 1, big.mark = ",")) +
    scale_x_log10(breaks = scales::breaks_log(n = 12),
                  labels = scales::number_format(accuracy = 1, big.mark = ",")) +
    geom_abline(slope = 1,

```

```

        color = "#3399FF",
        size = 1.1) +
ggdark::dark_theme_minimal() +
theme(axis.text.x      = element_text(angle = -90, vjust = 0.5, size = rel(0.9)),
      axis.text.y      = element_text(size = rel(0.9)),
      title            = element_text(size = rel(1.0)),
      legend.position = c(0.8, 0.3)) +
labs(title      = "mm10, Total Unique Genes vs Total Read Counts",
     subtitle   = "Points are Droplets, Blue Line: Slope = 1, Red Line: Linear Model",
     x          = "log10( Total mm10 Counts )",
     y          = "log10( Unique mm10 Genes )",
     color      = "Is Multiplet",
     shape      = "Is Multiplet")

```



View total_hg19_counts vs total_mm10_counts without log scaled values.

```

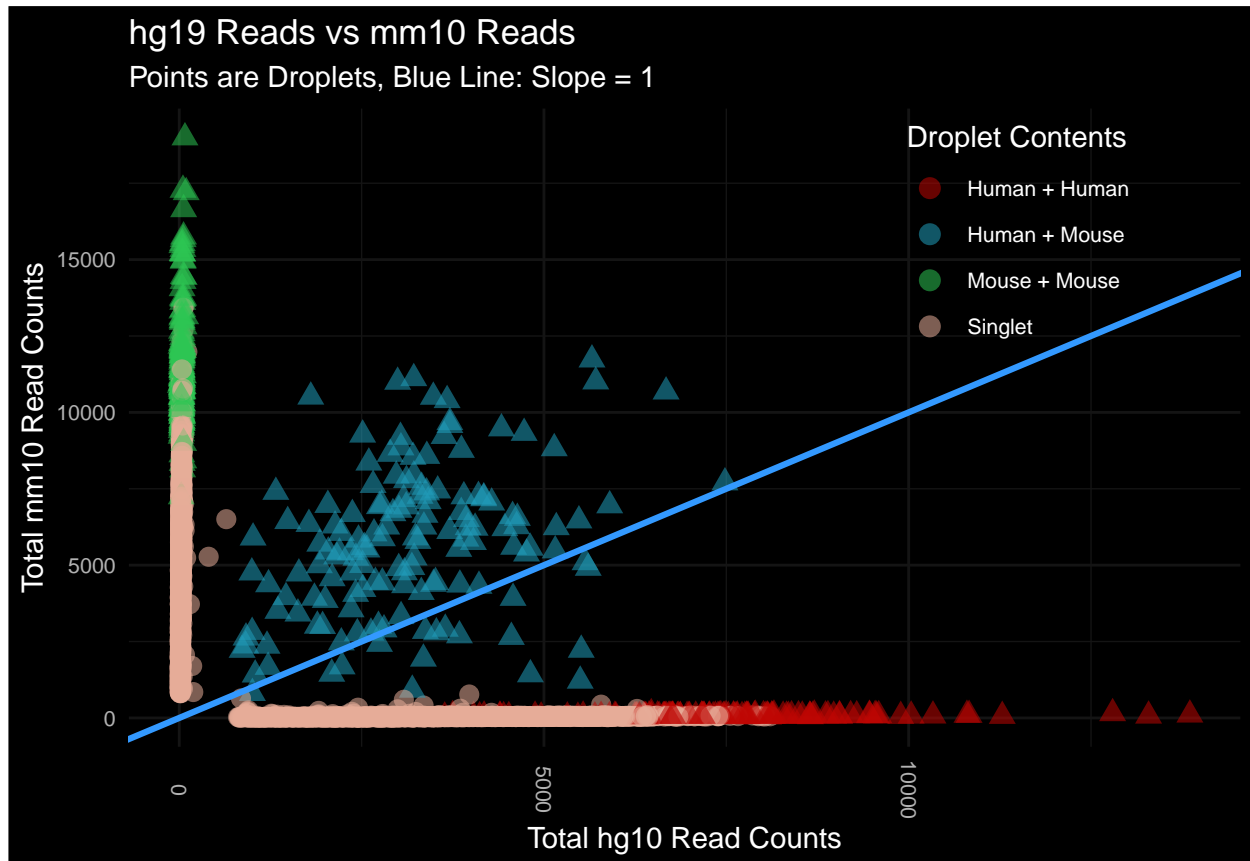
count_tbl_summary %>%
ggplot(aes(x = total_hg19_counts, y = total_mm10_counts, color = Multiplet_Type)) +
  geom_point(aes(shape = Multiplet), alpha = 0.55, size = 3.2) +
  scale_color_manual(values = c("#bf0603", "#219ebc", "#2dc653", "#e7ad99")) +
  geom_abline(slope = 1,
             color = "#3399FF",
             size = 1.1) +
ggdark::dark_theme_minimal() +
theme(axis.text.x      = element_text(angle = -90, vjust = 0.5, size = rel(0.9)),
      axis.text.y      = element_text(size = rel(0.9)),

```

```

title          = element_text(size = rel(1.0)),
legend.position = c(0.80, 0.80),
legend.text    = element_text(size = rel(0.75))) +
labs(title      = "hg19 Reads vs mm10 Reads",
     subtitle   = "Points are Droplets, Blue Line: Slope = 1",
     x          = "Total hg10 Read Counts",
     y          = "Total mm10 Read Counts",
     color      = "Droplet Contents") +
guides(shape = "none")

```



View `total_hg19_counts` vs `total_mm10_counts` with log scaled values.

```

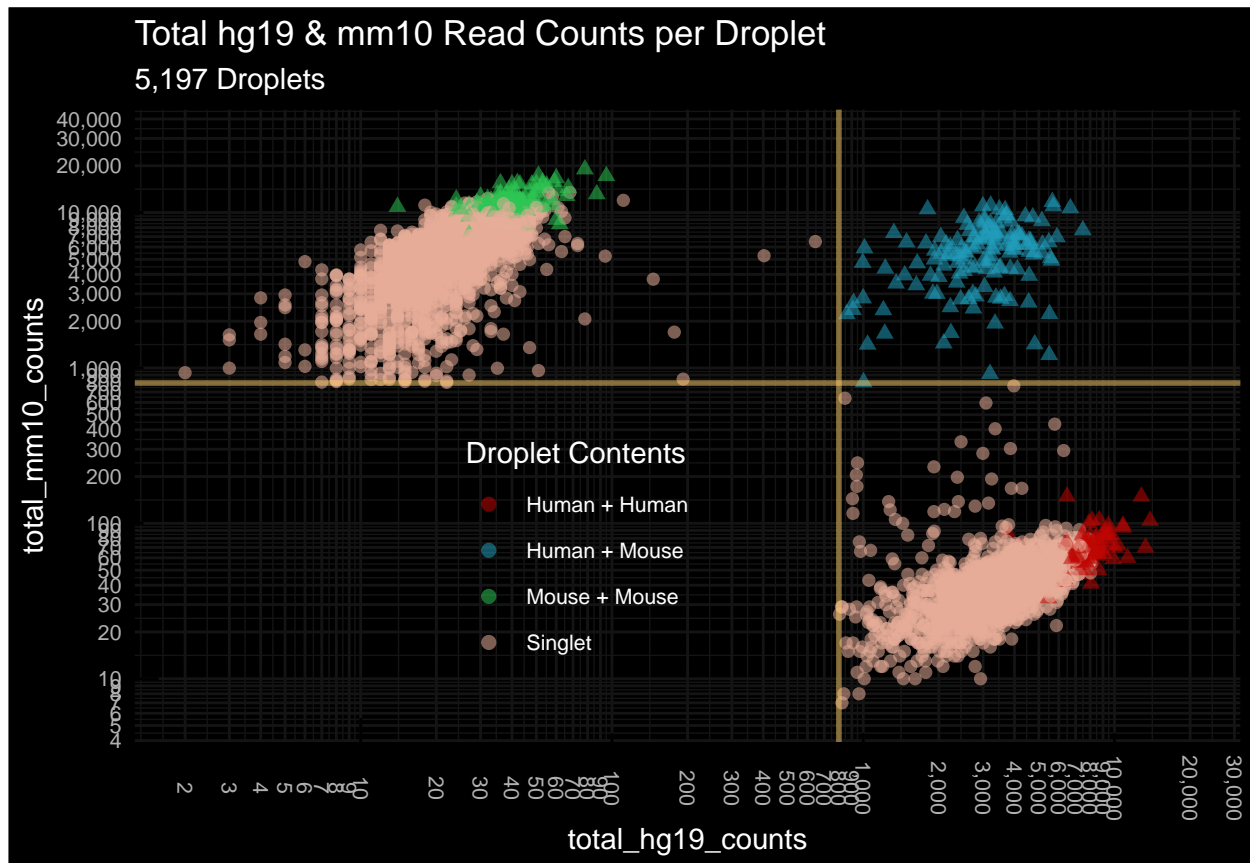
count_tbl_summary %>%
  ggplot(aes(x      = total_hg19_counts,
             y      = total_mm10_counts,
             color  = Multiplet_Type,
             shape  = Multiplet)) +
  geom_point(alpha = 0.55, size = 2) +
  scale_y_log10(breaks = scales::breaks_log(n = 38),
               labels = scales::number_format(accuracy = 1, big.mark = ","),
               limit = c(6, 30000)) +
  scale_x_log10(breaks = scales::breaks_log(n = 40),
               labels = scales::number_format(accuracy = 1, big.mark = ","),
               limit = c(2, 20000)) +
  scale_color_manual(values = c("#bf0603", "#219ebc", "#2dc653", "#e7ad99")) +
  geom_hline(yintercept = c(800),

```

```

        color      = "#FFCF6A",
        alpha      = 0.5,
        size       = 1) +
geom_vline(xintercept = c(800),
          color      = "#FFCF6A",
          alpha      = 0.5,
          size       = 1) +
ggdark::dark_theme_minimal() +
theme(axis.text.x    = element_text(angle = -90, vjust = 0.5, size = rel(0.9)),
      axis.text.y    = element_text(size = rel(0.9)),
      title          = element_text(size = rel(1.0)),
      legend.position = c(0.4, 0.3),
      legend.text     = element_text(size = rel(0.75))) +
annotation_logticks(scaled = FALSE) +
labs(title      = "Total hg19 & mm10 Read Counts per Droplet",
     subtitle   = "5,197 Droplets",
     color      = "Droplet Contents") +
guides(shape = "none")

```



View `total_hg19_counts` vs `total_mm10_counts` with log scaled values, but just the multiplets.

```

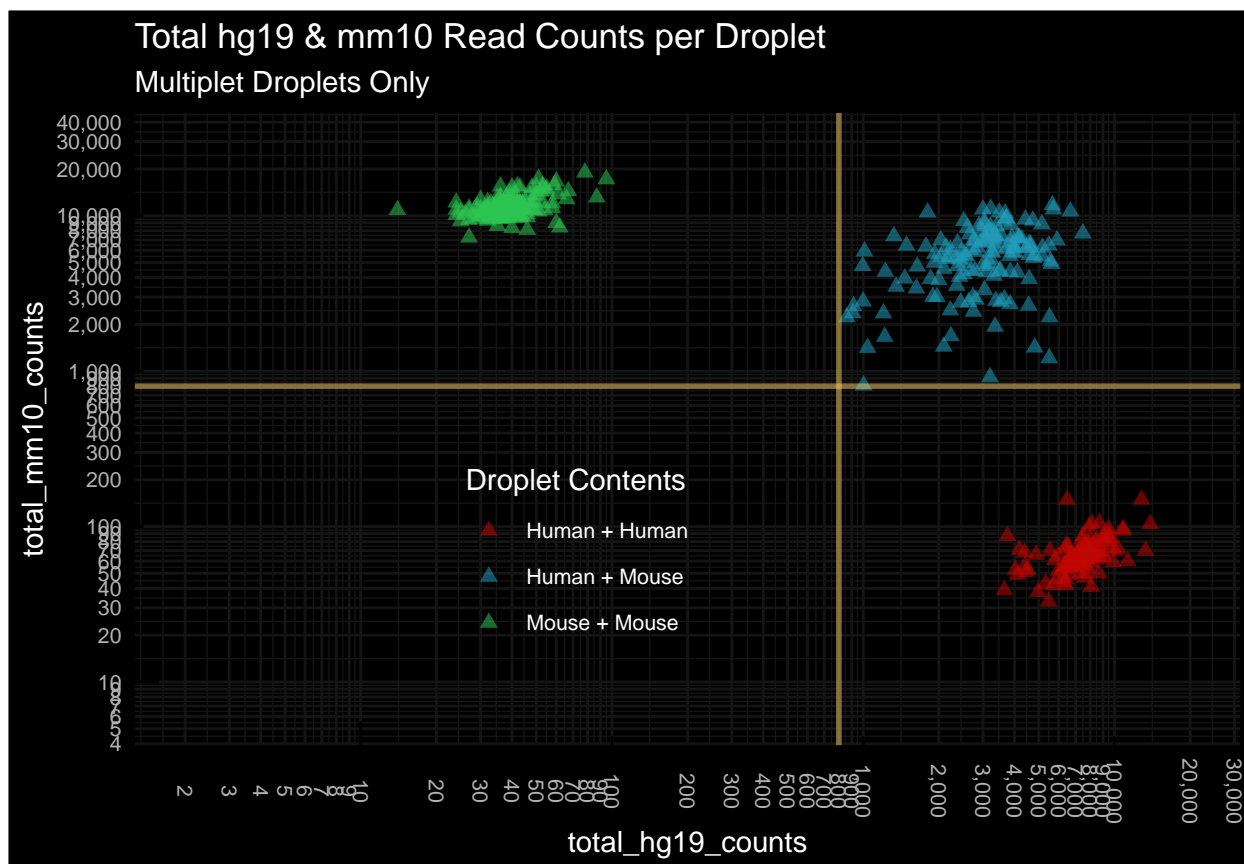
count_tbl_summary %>% filter(Multiplet == TRUE) %>%
  ggplot(aes(x = total_hg19_counts, y = total_mm10_counts)) +
  geom_point(aes(color = Multiplet_Type),
            alpha = 0.55,
            size = 2,

```

```

      shape = 17) +
scale_y_log10(breaks = scales::breaks_log(n = 38),
              labels = scales::number_format(accuracy = 1, big.mark = ","),
              limit = c(6, 30000)) +
scale_x_log10(breaks = scales::breaks_log(n = 40),
              labels = scales::number_format(accuracy = 1, big.mark = ","),
              limit = c(2, 20000)) +
scale_color_manual(values = c("#bf0603", "#219ebc", "#2dc653")) +
geom_hline(yintercept = c(800),
           color       = "#FFCF6A",
           alpha       = 0.5,
           size        = 1) +
geom_vline(xintercept = c(800),
           color       = "#FFCF6A",
           alpha       = 0.5,
           size        = 1) +
ggdark::dark_theme_minimal() +
theme(axis.text.x    = element_text(angle = -90, vjust = 0.5, size = rel(0.9)),
      axis.text.y    = element_text(size = rel(0.9)),
      title          = element_text(size = rel(1.0)),
      legend.position = c(0.4, 0.3),
      legend.text     = element_text(size = rel(0.75))) +
annotation_logticks(scaled = FALSE) +
labs(title      = "Total hg19 & mm10 Read Counts per Droplet",
     subtitle   = "Multiplet Droplets Only",
     color      = "Droplet Contents") +
guides(shape = "none")

```



Summary table of singlet and multiplet droplets identified.

```
count_tbl_summary %>%
  group_by(Droplet_Contents) %>%
  summarise(n = n()) %>%
  gt(rowname_col = "Droplet_Contents") %>%
  tab_header(title = "Droplet Contents",
             subtitle = "5413 Total Droplets") %>%
  tab_options(heading.subtitle.font.size = 12,
              heading.align = "left",
              table.border.top.color = "black",
              column_labels.border.bottom.color = "black",
              column_labels.border.bottom.width = px(3),
              column_labels.border.top.width = px(4),
              table.border.top.width = px(3))
```

Droplet Contents	
5413 Total Droplets	
	n
Human	2328
Human + Human	124
Human + Mouse	145
Mouse	2652
Mouse + Mouse	164

Save `count_tbl_summary` for further exploratory data analysis.

```
data.table::fwrite(count_tbl_summary, here("data", "count_tbl_summary.csv"))
```