Power analysis on skew normal fitting for right tail

Some notation

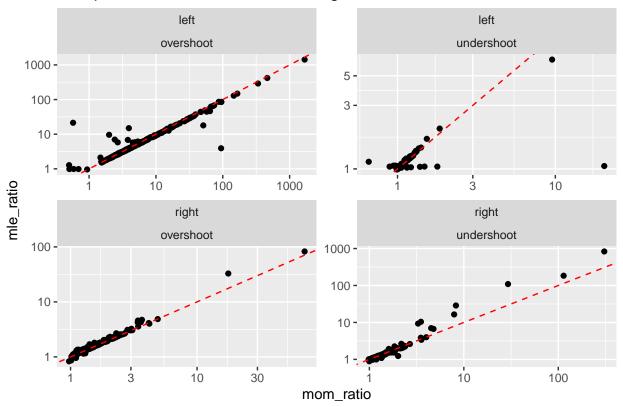
In this report, define A = subsample fitted probability; B = subsample empirical tail probability; C = full sample fitted probability and D = full sample empirical tail probability. I consider the analysis for A/D(D/A) and A/B(B/A).

1.Load results for fitting comparison

```
library(dplyr)
library(ggplot2)
library(tidyverse)
undershoot <- read_csv("undershoot_joint_fit.csv")[,-1]</pre>
overshoot <- read_csv("overshoot_joint_fit.csv")[,-1]</pre>
quantile_list <- seq(0.01, 0.99, length.out = 10)
no_sam <- round(exp(seq(log(1e3), log(5e4), length.out = 10)))</pre>
# rearrange the data frame
B <- 100
undershoot_df <- data.frame(id = rep(1:B, 10*10),
                             ratio_value = 0,
                             no_sam = 0,
                             ratio quantile = 0)
overshoot_df <- data.frame(id = rep(1:B, 10*10),</pre>
                            ratio value = 0,
                            no_sam = 0,
                            ratio_quantile = 0)
# i: quantile; j: no of sample
for (i in 1:10) {
  for (j in 1:10) {
    start <- (j - 1 + (i-1)*10)*B +1
    end <- (j + (i-1)*10)*B
    undershoot_df[start:end, 2] <- as.vector(undershoot[(((j-1)*B+1):(j*B)), (i-1)*3+32])[[1]]
    undershoot_df[start:end, 3] <- rep(no_sam[j], B)</pre>
    undershoot_df[start:end, 4] <- rep(quantile_list[i], B)</pre>
    overshoot_df[start:end, 2] \leftarrow as.vector(overshoot[(((j-1)*B+1):(j*B)), (i-1)*3+32])[[1]]
    overshoot_df[start:end, 3] <- rep(no_sam[j], B)</pre>
    overshoot_df[start:end, 4] <- rep(quantile_list[i], B)</pre>
  }
}
# load the oracle ratio
mle_param_nc <- read_csv("figures/power_exploration/sknorm_tail_prob_500000_resamples_0.96_percentile/p
mom_param_nc <- read_csv("figures/power_exploration/sknorm_tail_prob_500000_resamples_0.96_percentile/m
mle_param_twosides <- t(mle_param_nc[,-1])</pre>
mom_param_twosides <- t(mom_param_nc[,-1])</pre>
mle_overshoot_ratio <- as.numeric(mle_param_twosides[, 6])</pre>
```

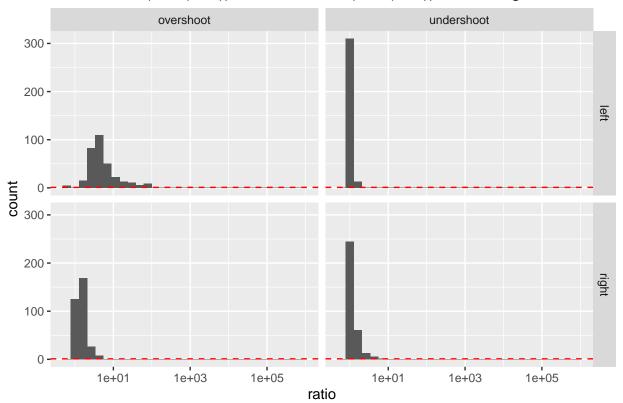
```
mle_undershoot_ratio <- as.numeric(mle_param_twosides[, 7])</pre>
mom_overshoot_ratio <- as.numeric(mom_param_twosides[, 6])</pre>
mom_undershoot_ratio <- as.numeric(mom_param_twosides[, 7])</pre>
quantile_list \leftarrow seq(0.01, 0.99, length.out = 10)
overshoot_set <- data.frame(index = numeric(10), ratio = numeric(10))</pre>
undershoot_set <- data.frame(index = numeric(10), ratio = numeric(10))</pre>
# find distributions based on right tail
for (r in 1:10){
  dist <- abs(mle_overshoot_ratio[331:660] - quantile(mle_overshoot_ratio[331:660], quantile_list[r]))
  overshoot_set[r, 1] <- which(dist == min(dist))</pre>
  overshoot_set[r, 2] <- mom_overshoot_ratio[which(dist == min(dist)) + 330]</pre>
  dist <- abs(mle undershoot ratio[331:660] - quantile(mle undershoot ratio[331:660], quantile list[r])
  undershoot_set[r, 1] <- which(dist == min(dist))</pre>
  undershoot_set[r, 2] <- mom_undershoot_ratio[which(dist == min(dist)) + 330]
}
# accuracy using 5e5 mom estimate
mle_groundtruth <- data.frame(ratio = c(mle_overshoot_ratio, mle_undershoot_ratio),</pre>
                               type = c(rep("overshoot", 660), rep("undershoot", 660)),
                               tail = c(rep(c(rep("left", 330),rep("right", 330)) , 2)),
                               id = c(rep(1:330, 2)))
mom_groundtruth <- data.frame(ratio = c(mom_overshoot_ratio, mom_undershoot_ratio),</pre>
                               type = c(rep("overshoot", 660), rep("undershoot", 660)),
                               tail = c(rep(c(rep("left", 330),rep("right", 330)), 2)),
                               id = c(rep(1:330, 2)))
groundtruth_comp <- data.frame(mom_ratio = c(mom_overshoot_ratio, mom_undershoot_ratio),</pre>
                                mle_ratio = c(mle_overshoot_ratio, mle_undershoot_ratio),
                                type = c(rep("overshoot", 660), rep("undershoot", 660)),
                                tail = c(rep(c(rep("left", 330),rep("right", 330)) , 2)),
                                id = c(rep(1:330, 2)))
groundtruth_comp |>
  filter(mom_ratio / mle_ratio <= 100) |>
  ggplot(aes_string(x = "mom_ratio", y = "mle_ratio")) +
  facet_wrap(tail~type, scales = "free") +
  geom_point() +
  scale_y_log10() +
  scale_x_log10() +
  geom_abline(linetype = "dashed", color = "red") +
  labs(title = "Comparison: mom versus mle in negative control")
```

Comparison: mom versus mle in negative control



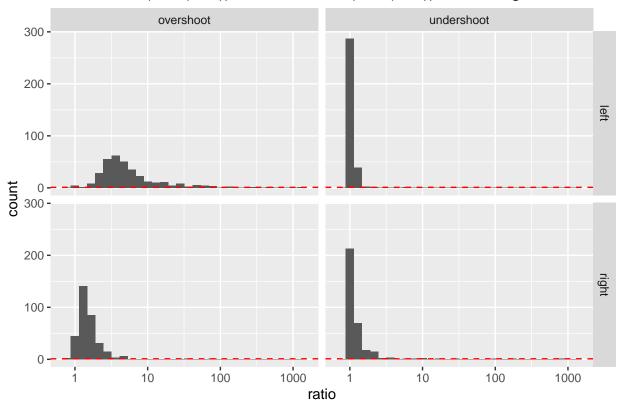
```
mom_groundtruth |>
    ggplot(aes_string(x = "ratio")) +
    facet_grid(tail~type) +
    geom_histogram() +
    scale_x_log10() +
    geom_hline(yintercept = 1, linetype = "dashed", colour = "red") +
    labs(title = "Undershoot (max(D/C)) and overshoot (max(C/D)) ratios: negative control")
```

Undershoot (max(D/C)) and overshoot (max(C/D)) ratios: negative control

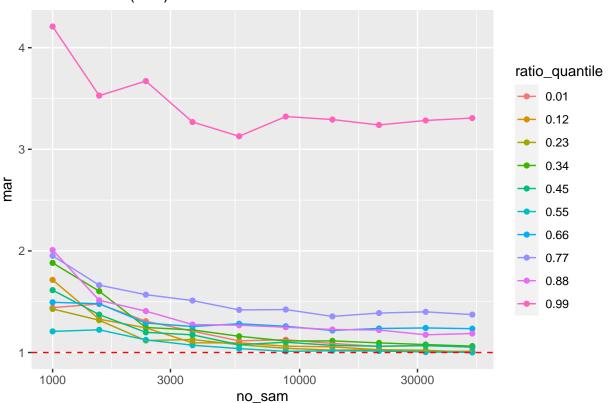


```
mle_groundtruth |>
    ggplot(aes_string(x = "ratio")) +
    facet_grid(tail~type) +
    geom_histogram() +
    scale_x_log10() +
    geom_hline(yintercept = 1, linetype = "dashed", colour = "red") +
    labs(title = "Undershoot (max(D/C)) and overshoot (max(C/D)) ratios: negative control")
```

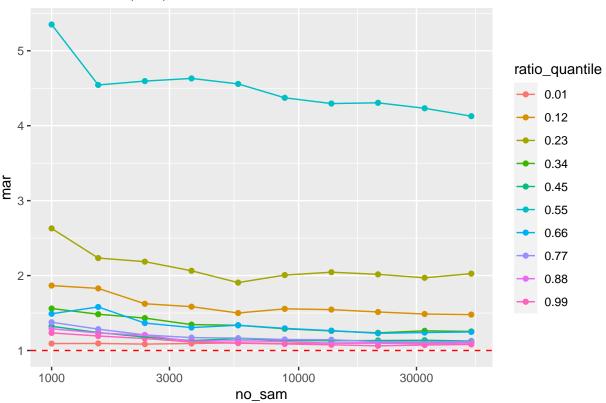
Undershoot (max(D/C)) and overshoot (max(C/D)) ratios: negative control



Ratio of max(D/A)



Ratio of max(A/D)



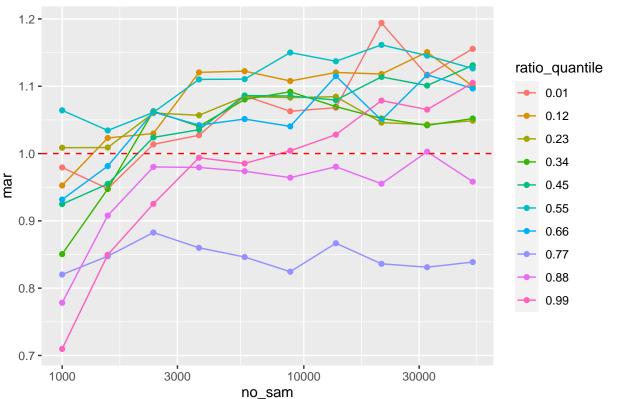
```
# store D/A (A/D)
gt_overshoot_curve <- overshoot_df$ratio_value
gt_undershoot_curve <- undershoot_df$ratio_value</pre>
```

Here we clearly see the ratio $\max[A/D](\max[D/A])$ approaches $\max[C/D](\max[D/C])$ very fast. This indicates the error decrease rather fast in estimating the tail probability at least in average sense. In the next section, we mainly consider the changes for the ratio $\max[A/B](\max[B/A])$ and how does this approach $\max[C/D](\max[D/C])$.

1.Load results for power comparison

```
for (i in 1:10) {
  for (j in 1:10) {
    start <- (j - 1 + (i-1)*10)*B +1
    end <- (j + (i-1)*10)*B
     undershoot_df[start:end, 2] \leftarrow as.vector(undershoot[(((j-1)*B+1):(j*B)), (i-1)*3+32])[[1]] 
    undershoot_df[start:end, 3] <- rep(no_sam[j], B)</pre>
    undershoot_df[start:end, 4] <- rep(quantile_list[i], B)</pre>
    overshoot_df[start:end, 2] \leftarrow as.vector(overshoot[(((j-1)*B+1):(j*B)), (i-1)*3+32])[[1]]
    overshoot_df[start:end, 3] <- rep(no_sam[j], B)</pre>
    overshoot_df[start:end, 4] <- rep(quantile_list[i], B)</pre>
  }
}
# load the oracle ratio
undershoot_acc <- matrix(abs(undershoot_df$ratio_value / gt_undershoot_curve), 100, 100)
undershoot_ame <- data.frame(mar = apply(undershoot_acc, 2, mean),</pre>
                              no_sam = rep(no_sam, 10),
                              ratio_quantile = as.character(round(rep(quantile_list, each = 10), 2)))
undershoot_ame |>
  ggplot(aes_string(x = "no_sam", y = "mar", colour = "ratio_quantile")) +
  scale_x_log10() +
  geom_point() +
  geom_line() +
  geom_hline(yintercept = 1, linetype = "dashed", colour = "red") +
  labs(title = "Ratio of max(B/A) / max(D/A)")
```

Ratio of max(B/A) / max(D/A)



Ratio of max(A/B) / max(A/D)

