

Estimate Scaling Parameter in SUCCOTASH

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Abstract

I compare scaled-variance SUCCOTASH to procedures that do not change the variance and to the ad-hoc doubling of the variance. Estimating the scaling parameter results in a slightly anti-conservative procedure. It seems to perform best at $\pi_0 = 0.5$ and not as well as the ad-hoc variance inflation procedures at $\pi_0 = 0.9$ or 1. MSE and AUC for this new procedure are very competitive.

Results

```
library(knitr)
library(xtable)
library(dplyr)
library(reshape2)
library(ggplot2)
```

To view a description of these simulations and the results when the variance was not-inflated, please see http://dcgerard.github.io/flash_sims/analysis/flashr_v_succ.pdf.

In the plots below, “scale_succ1” refers to estimating the scaling parameter starting at a value of 1. “scale_succ2” is when the optimization starts at a value of 2. They give identical results.

Estimates of π_0 when estimating the scaling parameter are slightly anti-conservative. It seems that estimating the scaling parameter performs the best when $\pi_0 = 0.5$, but not as well as just doubling the variance when $\pi_0 = 0.9$ or 1.

MSE and AUC when estimating the scaling parameter are as good or better than every other method.

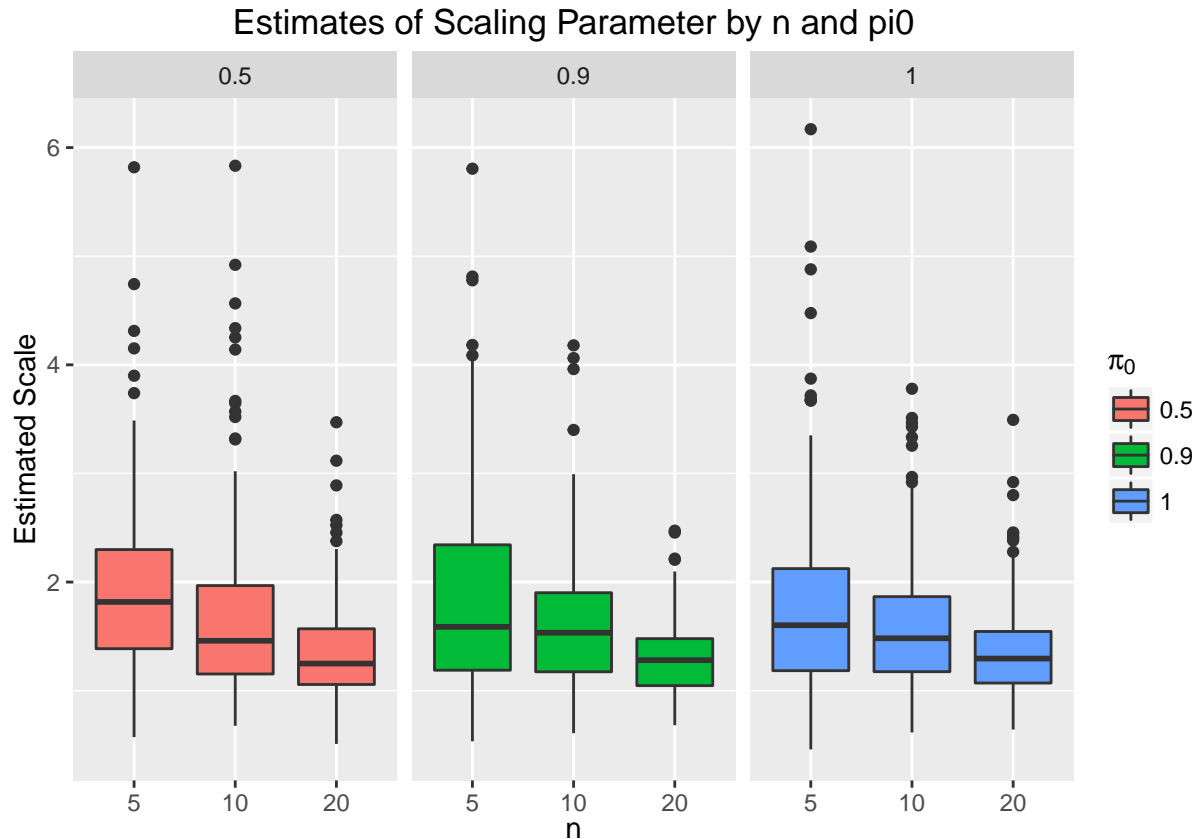
Why does inflating the variance by 2 seem to work so well for this data set? On average, the variance was always inflated. For small n , this inflation was actually very nearly 2. Perhaps 2 isn’t optimal, but better than no variance inflation.

```
scale_est_mat <- tbl_df(read.csv("scale_est_ssuc.csv", header = TRUE))
kable(aggregate(scale_suc1 ~ nullpi + nsamp, FUN = mean, data = scale_est_mat),
      col.names = c("$\\pi_0$", "$n$", "Mean Estimated Scale"), digits = 1)
```

π_0	n	Mean Estimated Scale
0.5	5	1.9
0.9	5	1.8
1.0	5	1.8
0.5	10	1.7
0.9	10	1.6
1.0	10	1.6
0.5	20	1.3
0.9	20	1.3
1.0	20	1.4

```
ggplot(data = scale_est_mat, mapping = aes(x = factor(nsamp), y = scale_suc1,
                                             fill = factor(nullpi))) +

  facet_grid(.~nullpi) +
  geom_boxplot() +
  xlab(expression(n)) + ylab("Estimated Scale") +
  scale_fill_discrete(name=expression(pi[0])) +
  ggtitle("Estimates of Scaling Parameter by n and pi0")
```



$\hat{\pi}_0$ Plots

```
double_pi0 <- read.csv("../double_succ/pi0_mat.csv")
reg_pi0 <- read.csv("../flash_v_rest_using_package/pi0_mat.csv")
scale_pi0 <- read.csv("pi0_ssuc.csv")
reg_pi0$inflate_succ <- double_pi0$succotash
reg_pi0$inflate_caterr_ash <- double_pi0$cate_rr_ash
reg_pi0$inflate_catenc_ash <- double_pi0$cate_nc_ash
reg_pi0$inflate_ols_ash <- double_pi0$ols_ash
reg_pi0$scale_suc1 <- scale_pi0$scale_suc1
reg_pi0$scale_suc2 <- scale_pi0$scale_suc2
reg_pi0 <- tbl_df(reg_pi0)
reg_pi0 <- reg_pi0[, c(1:2, 17, 3:4, 14, 18:19, 5:6, 15, 7:9, 16, 10:13)]
nsamp_seq <- unique(reg_pi0$nsamp)
nullpi_seq <- unique(reg_pi0$nullpi)
```

```

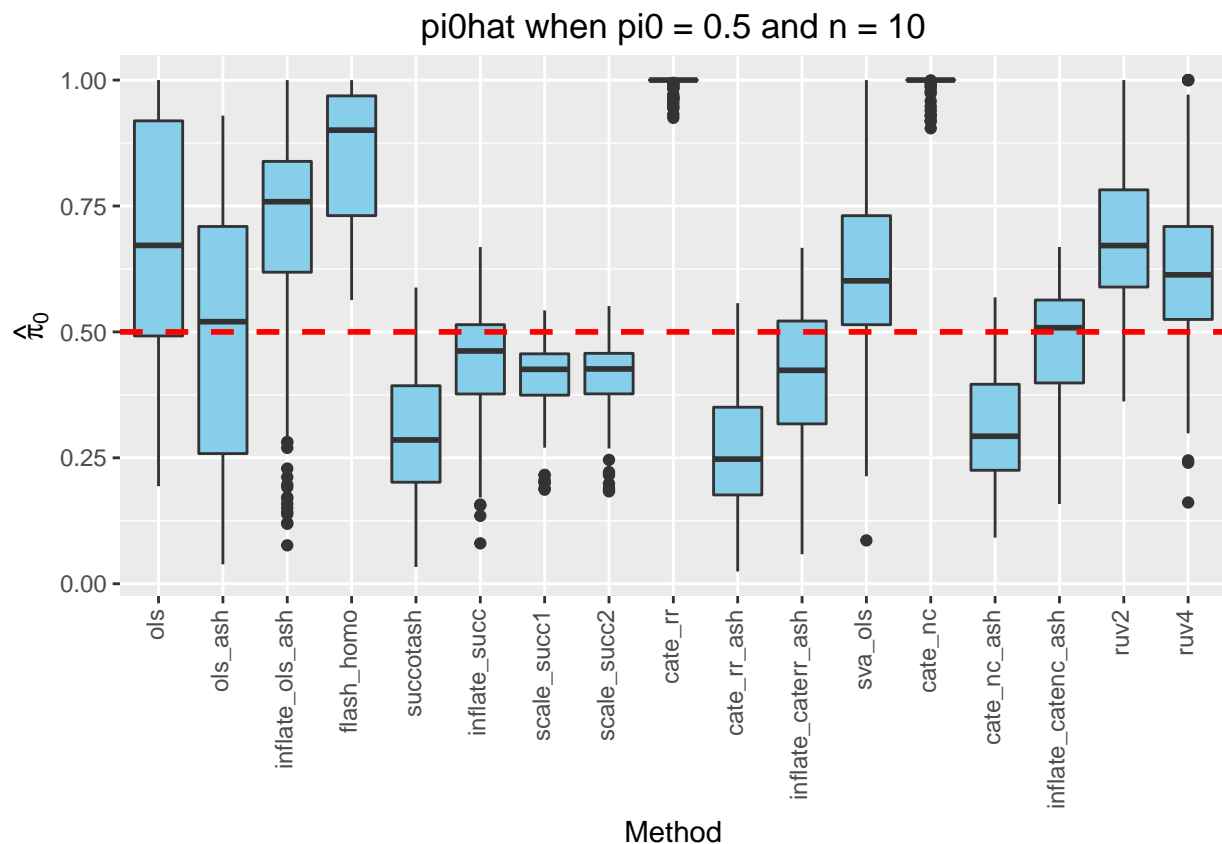
for (current_pi in nullpi_seq) {
  for (current_nsamp in nsamp_seq) {

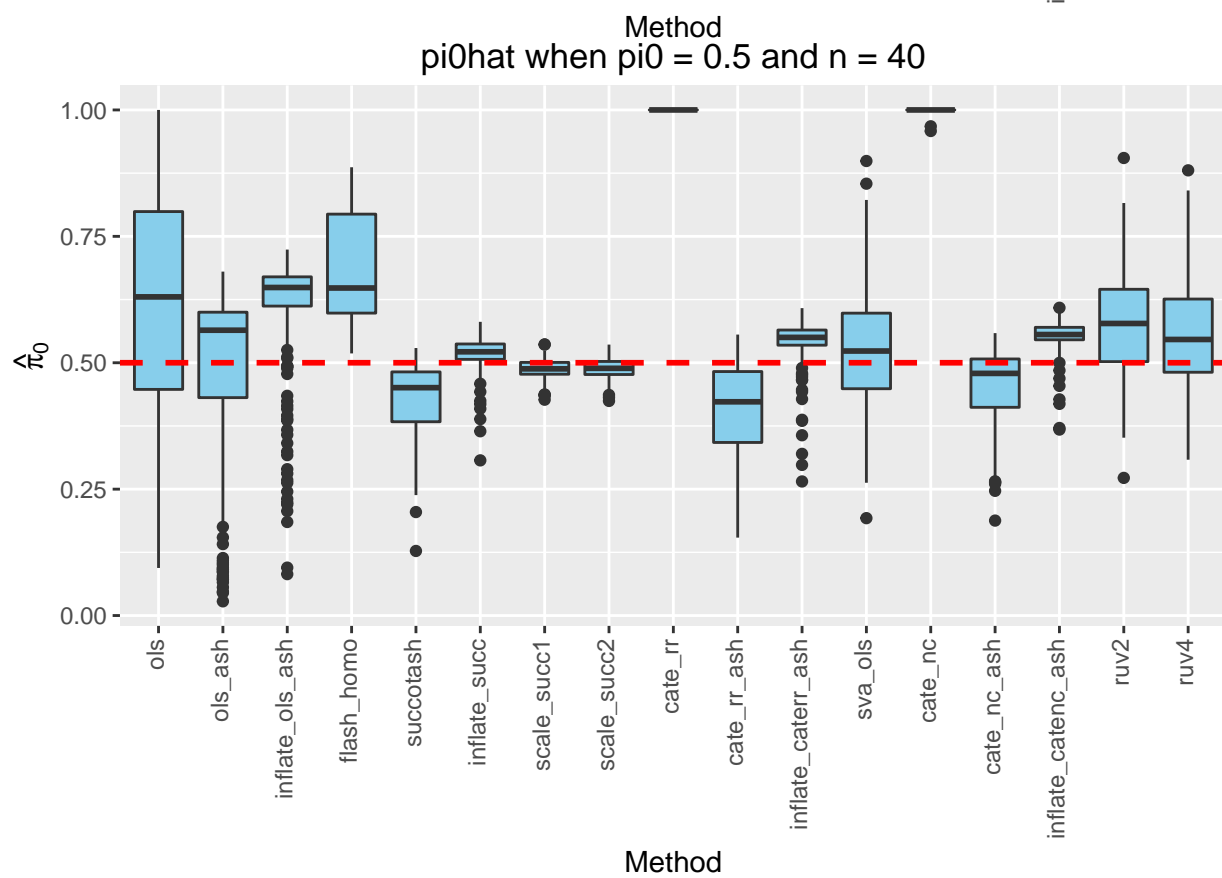
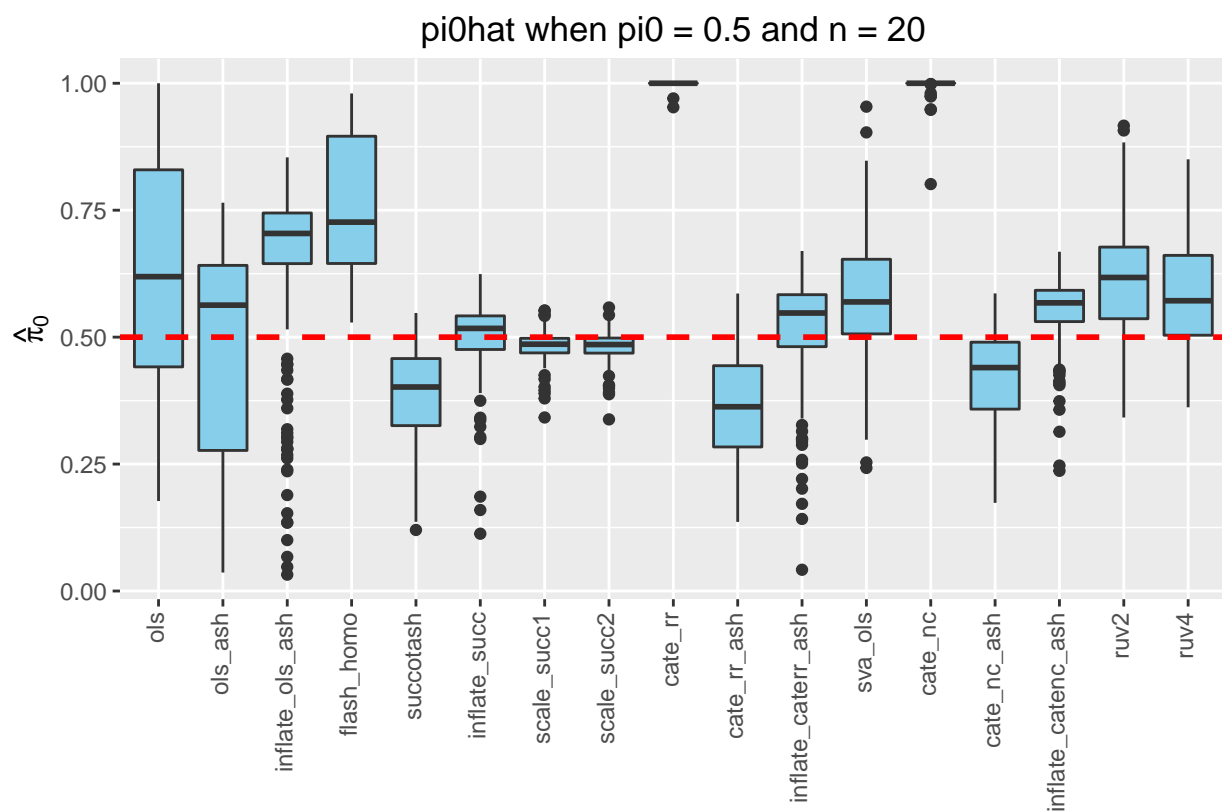
    subdf <- select(
      filter(
        reg_pi0, nullpi == current_pi & nsamp == current_nsamp),
      -c(nsamp, nullpi)
    )

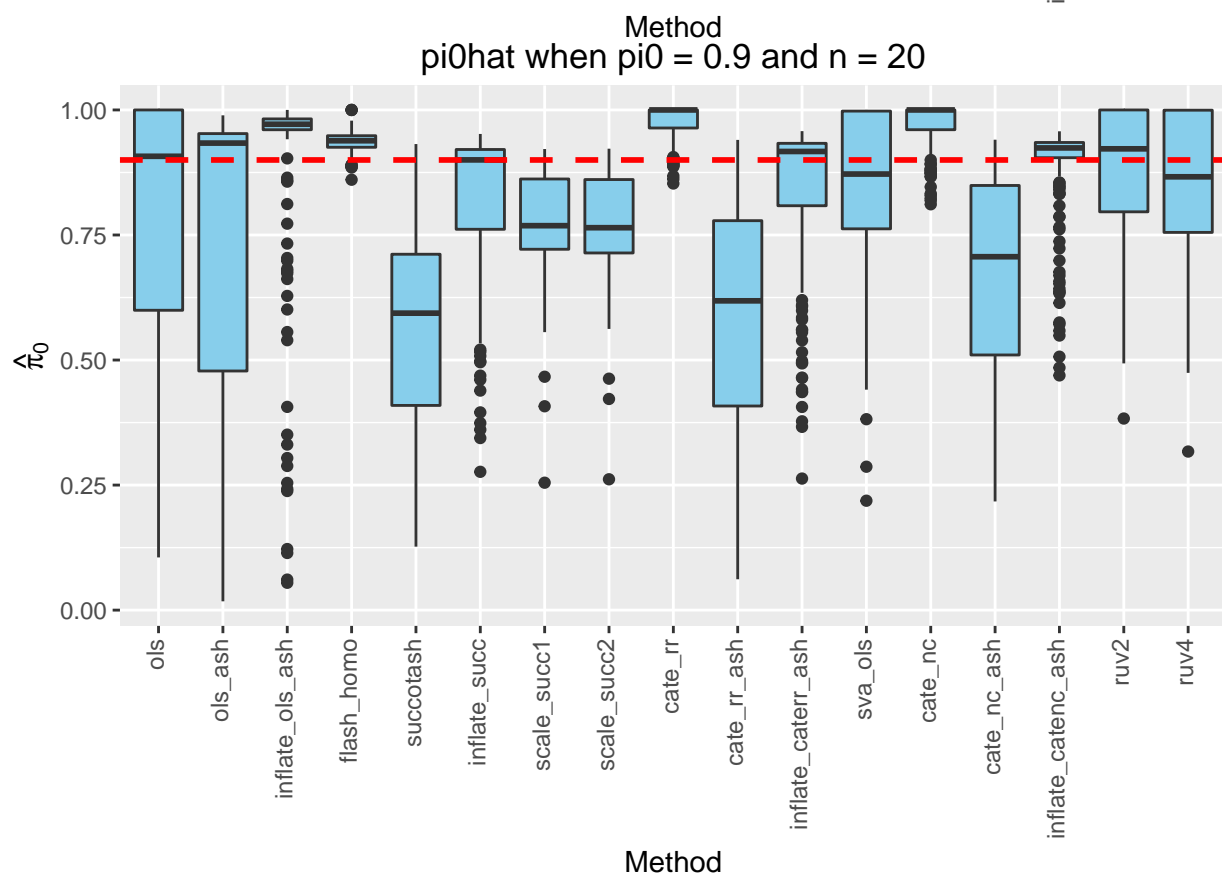
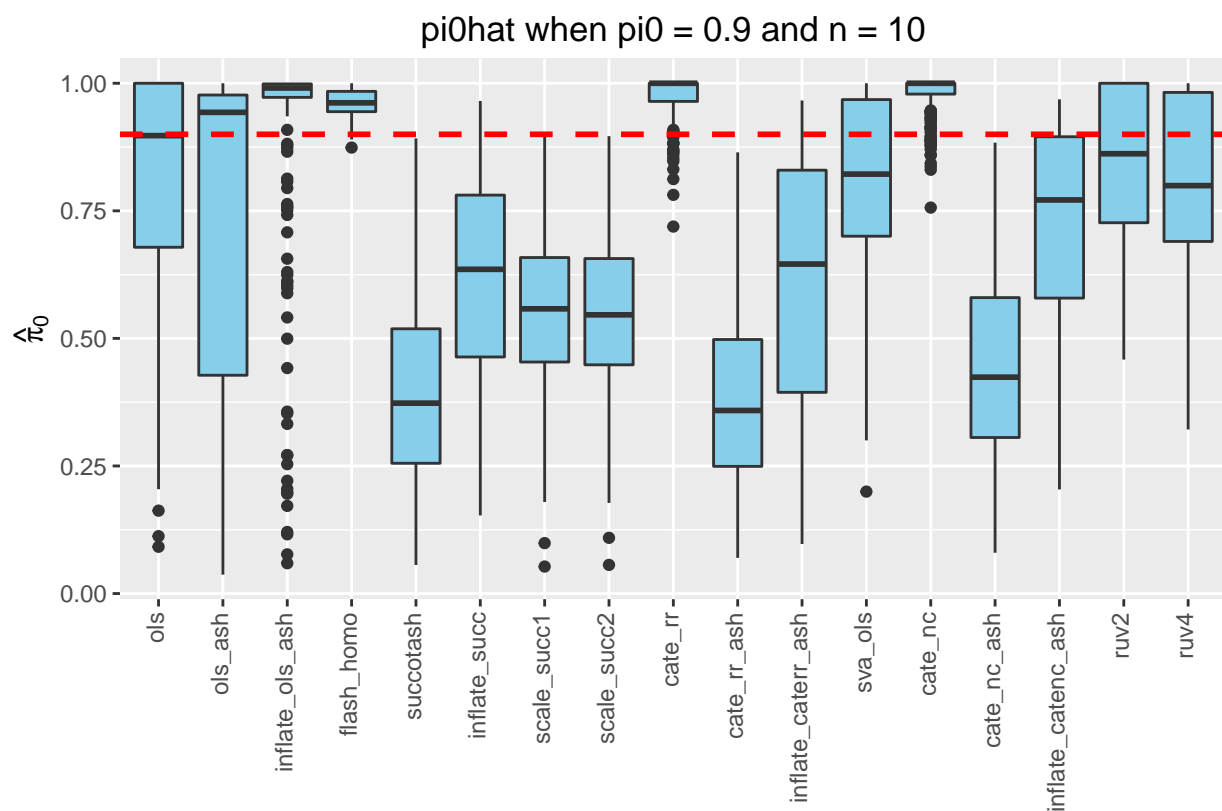
    melted_df <- melt(subdf, id.vars = NULL)

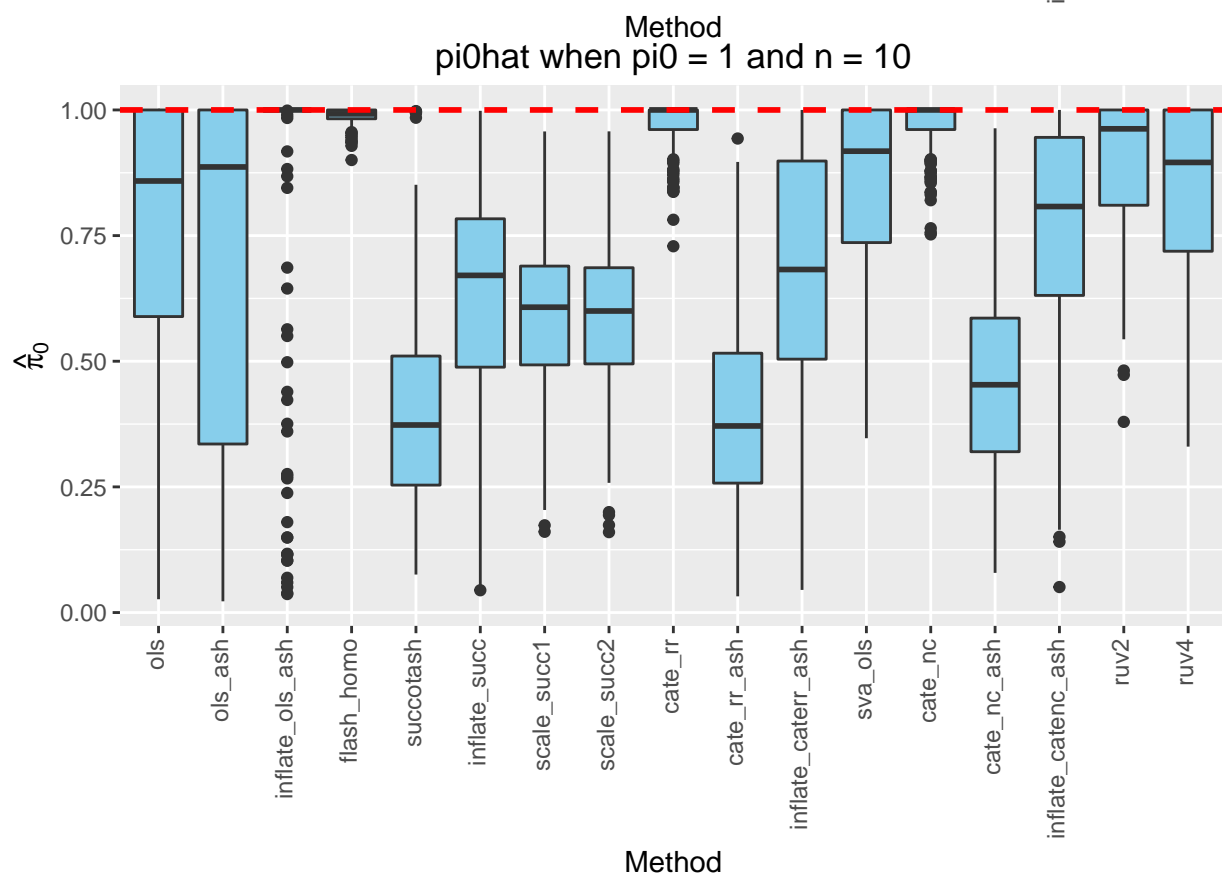
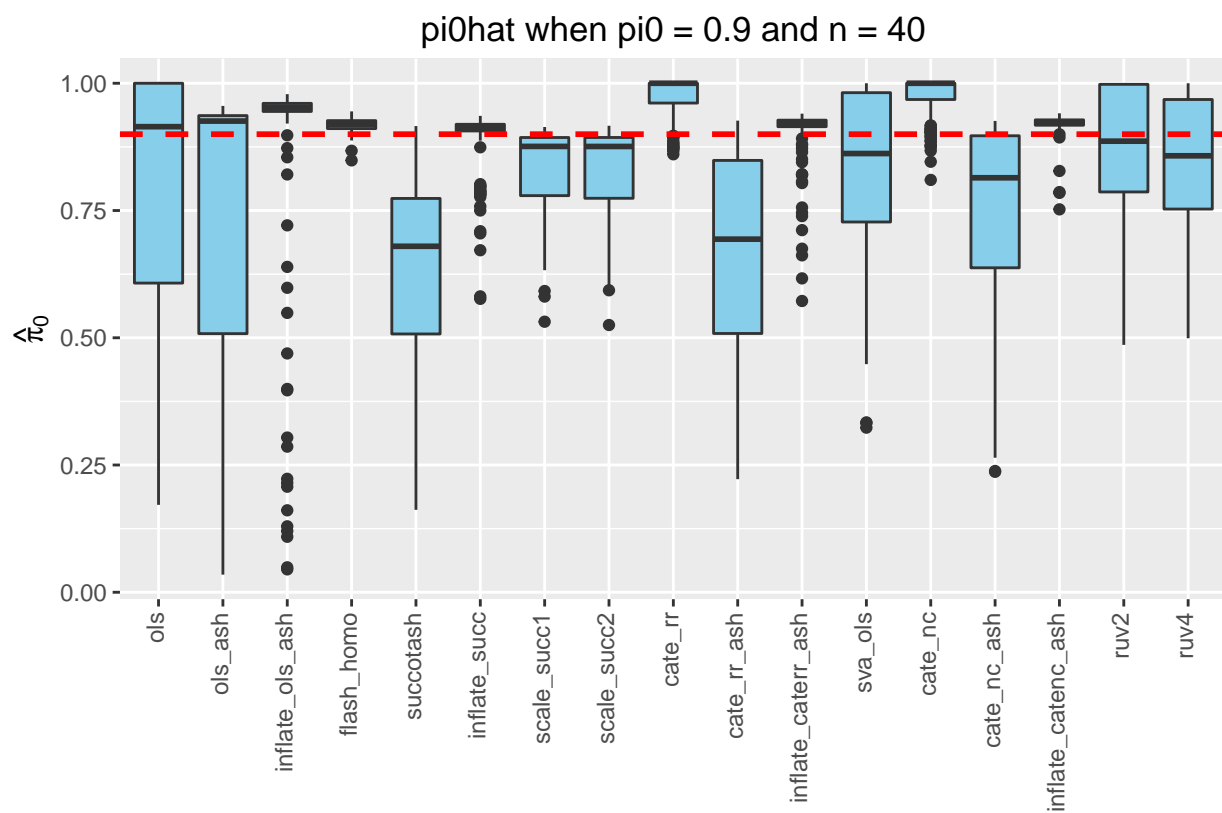
    p <- ggplot(data = melted_df, mapping = aes(x = variable, y = value)) +
      geom_boxplot(fill = I("skyblue")) +
      xlab(label = "Method") + ylab(label = expression(hat(pi)[0])) +
      geom_hline(yintercept = current_pi, color = I("red"), lty = 2, lwd = 1) +
      ggtitle(paste("pi0hat when pi0 =", current_pi, "and n =", current_nsamp * 2)) +
      theme(axis.text.x = element_text(angle = 90, hjust = 1, vjust = 0.3))
    print(p)
  }
}

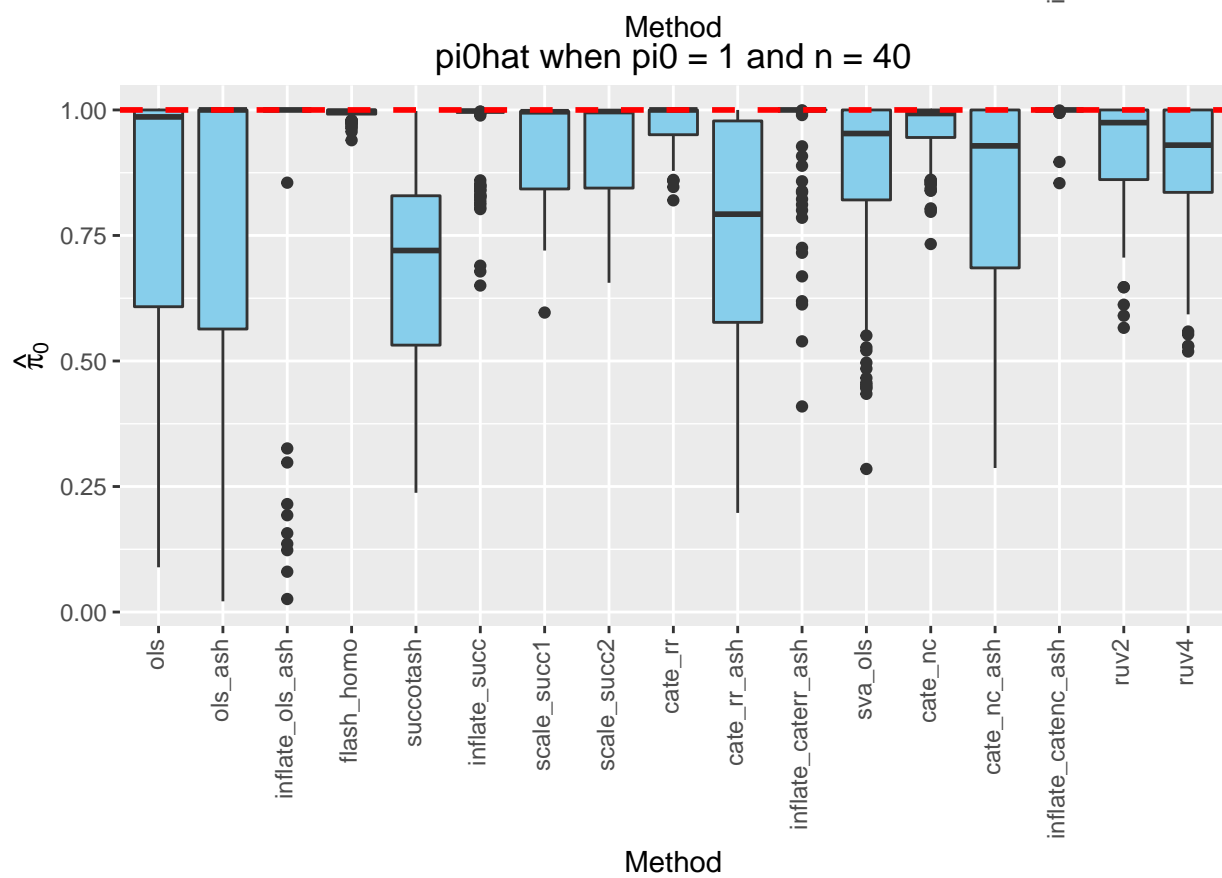
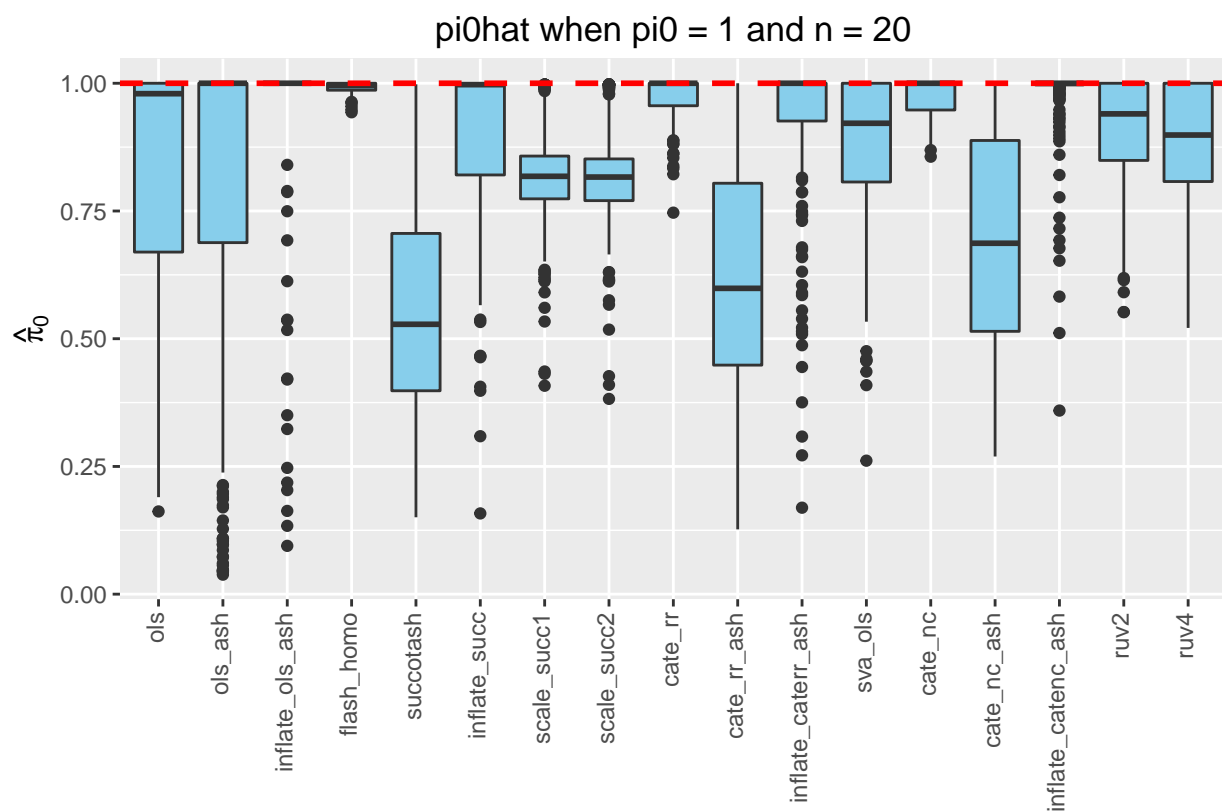
```











MSE Plots

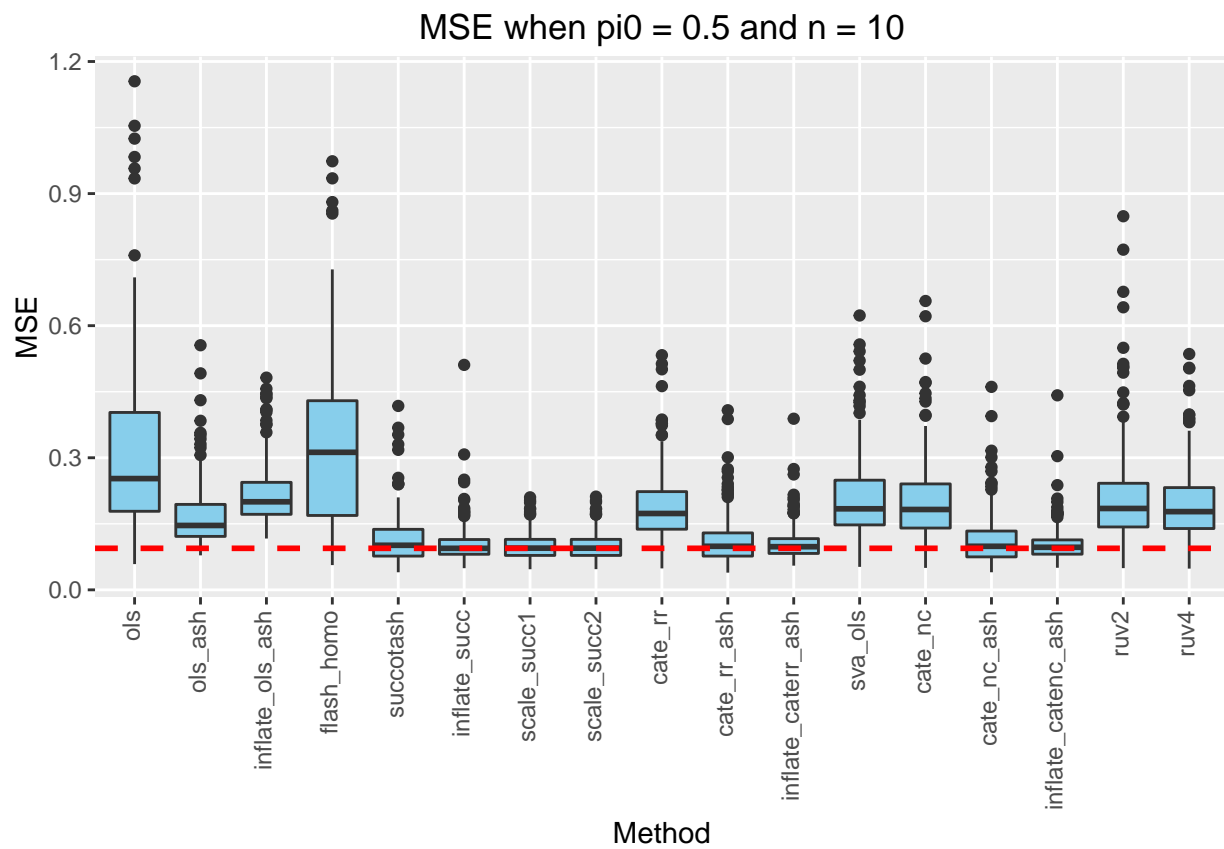
```
double_mse <- read.csv("../double_succ/mse_mat.csv")
reg_mse <- read.csv("../flash_v_rest_using_package/mse_mat.csv")
scale_mse <- read.csv("mse_ssuc.csv")
reg_mse$inflate_succ <- double_mse$succotash
reg_mse$inflate_caterr_ash <- double_mse$cate_rr_ash
reg_mse$inflate_catenc_ash <- double_mse$cate_nc_ash
reg_mse$inflate_ols_ash <- double_mse$ols_ash
reg_mse$scale_succ1 <- scale_mse$scale_suc1
reg_mse$scale_succ2 <- scale_mse$scale_suc2
reg_mse <- tbl_df(reg_mse)
reg_mse <- reg_mse[, c(1:2, 17, 3:4, 14, 18:19, 5:6, 15, 7:9, 16, 10:13)]
nsamp_seq <- unique(reg_mse$nsamp)
nullpi_seq <- unique(reg_mse$nullpi)
for (current_pi in nullpi_seq) {
  for (current_nsamp in nsamp_seq) {

    subdf <- select(
      filter(
        reg_mse, nullpi == current_pi & nsamp == current_nsamp),
      ~c(nsamp, nullpi)
    )

    hval <- min(apply(subdf, 2, median))

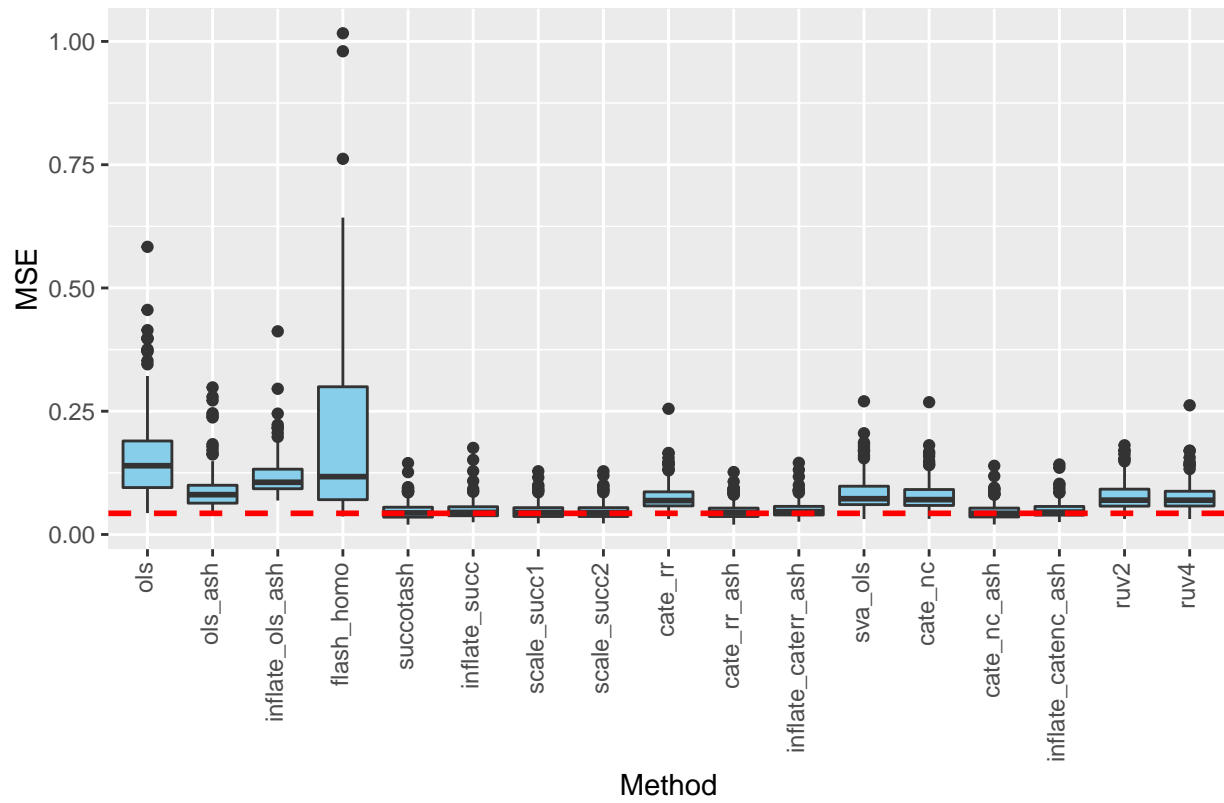
    melted_df <- melt(subdf, id.vars = NULL)

    p <- ggplot(data = melted_df, mapping = aes(x = variable, y = value)) +
      geom_boxplot(fill = I("skyblue")) +
      xlab(label = "Method") + ylab(label = "MSE") +
      geom_hline(yintercept = hval, color = I("red"), lty = 2, lwd = 1) +
      ggtitle(paste("MSE when pi0 =", current_pi, "and n =", current_nsamp * 2)) +
      theme(axis.text.x = element_text(angle = 90, hjust = 1, vjust = 0.3))
    print(p)
  }
}
```

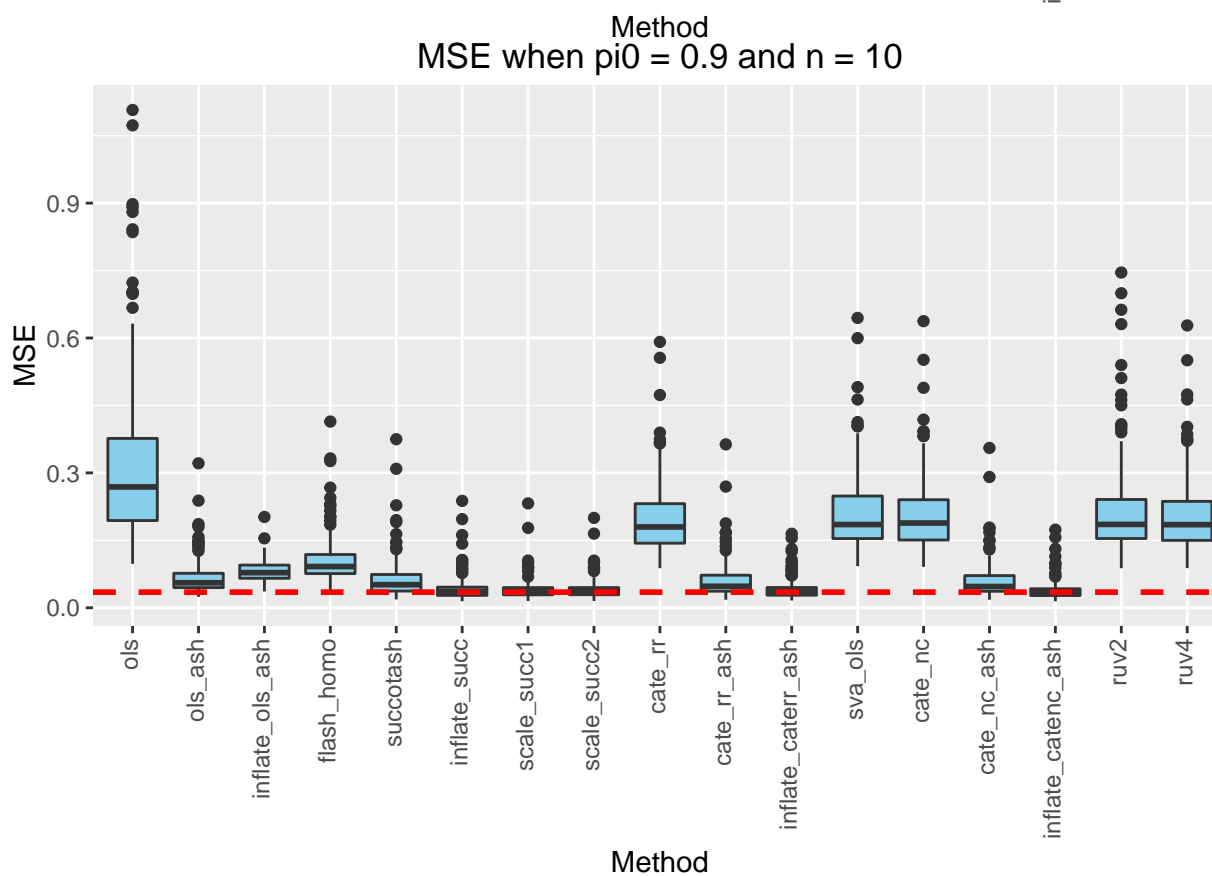
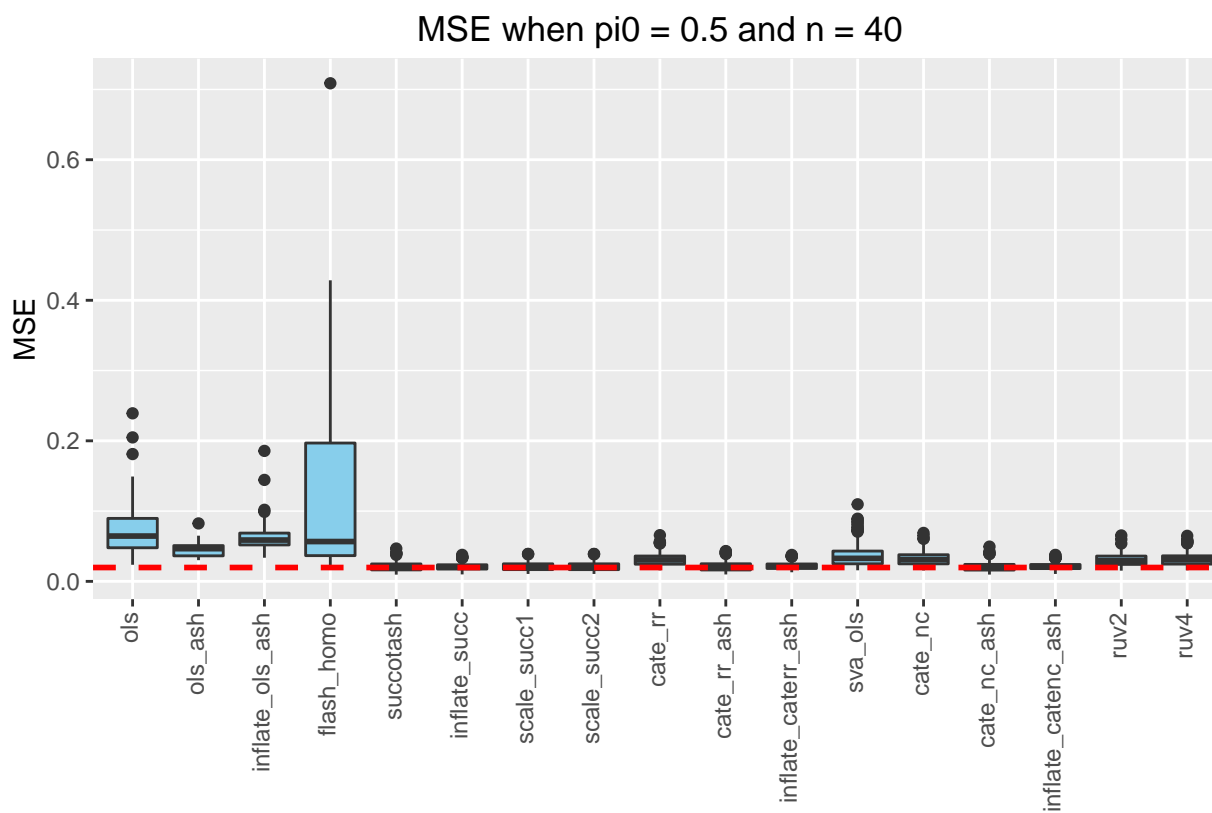



Warning: Removed 5 rows containing non-finite values (stat_boxplot).

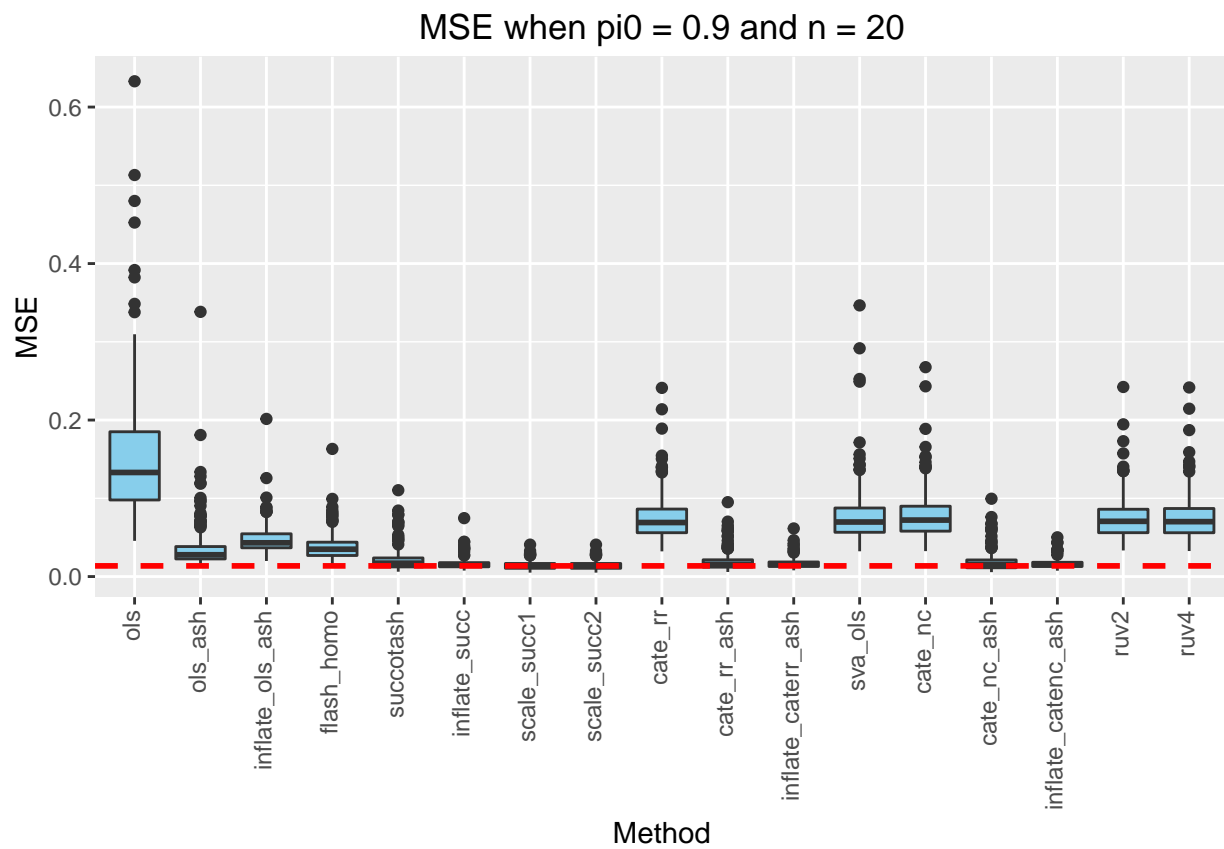
MSE when $\pi_0 = 0.5$ and $n = 20$



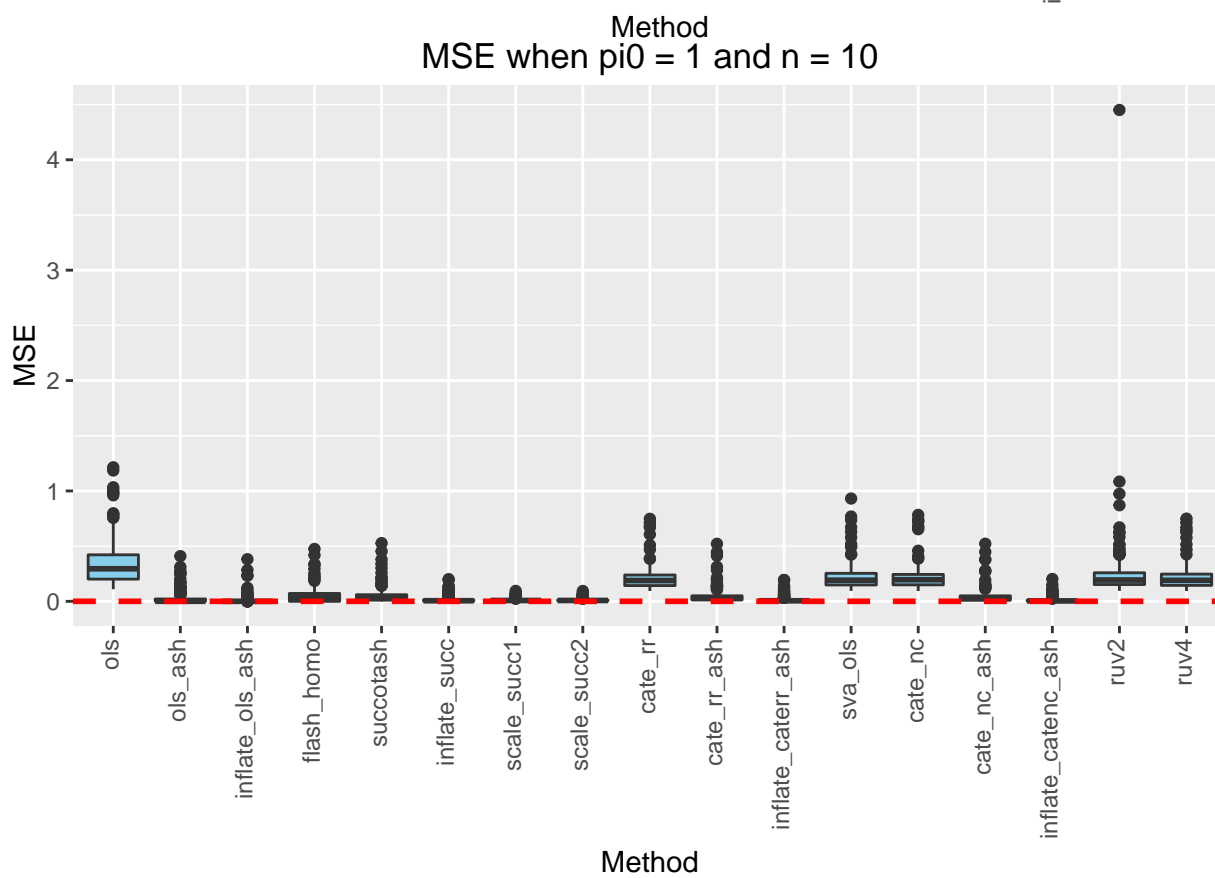
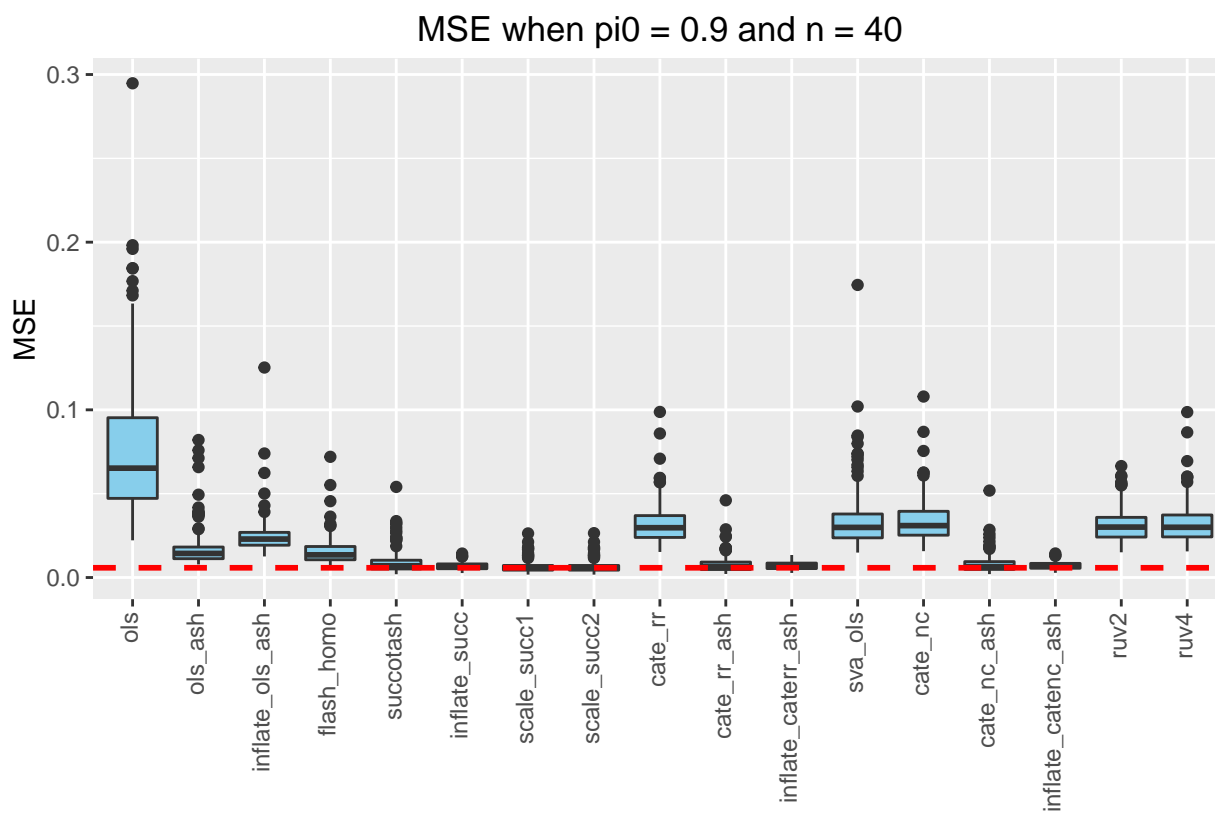
Warning: Removed 203 rows containing non-finite values (stat_boxplot).



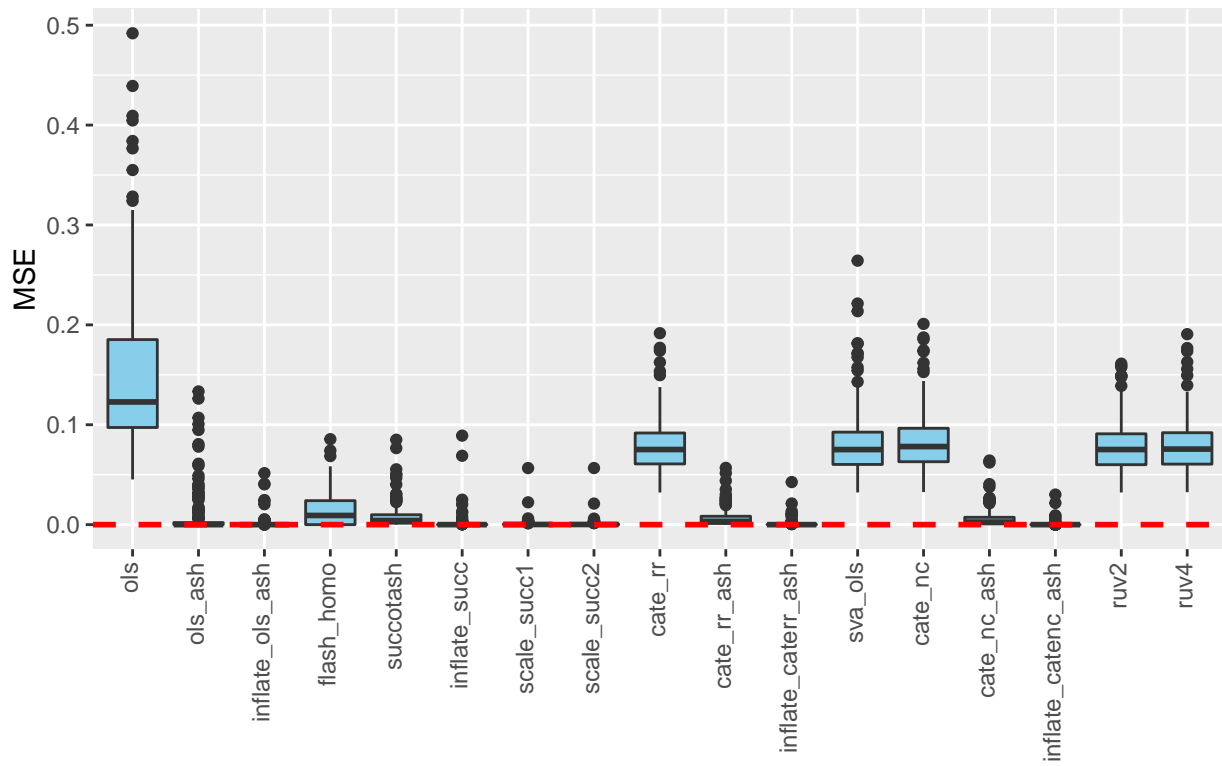
Warning: Removed 1 rows containing non-finite values (stat_boxplot).



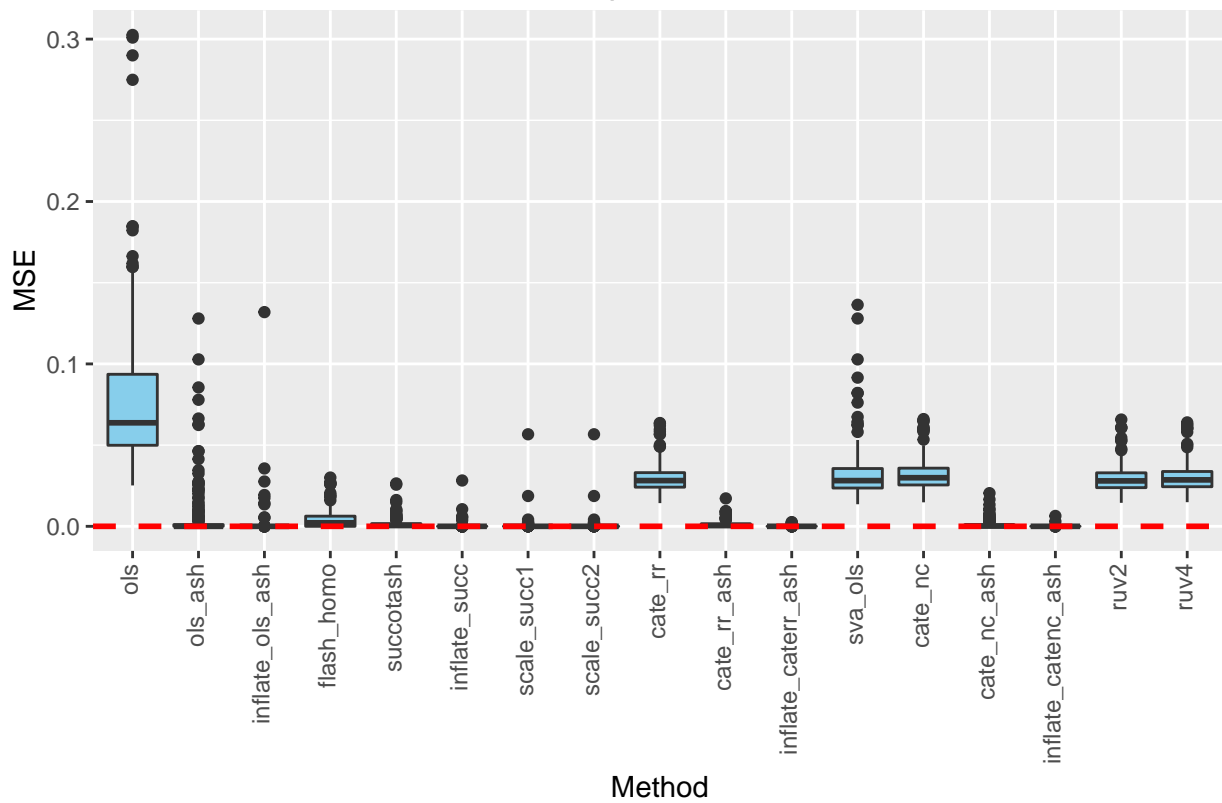
Warning: Removed 89 rows containing non-finite values (stat_boxplot).



MSE when $\pi_0 = 1$ and $n = 20$



MSE when $\pi_0 = 1$ and $n = 40$



AUC Plots

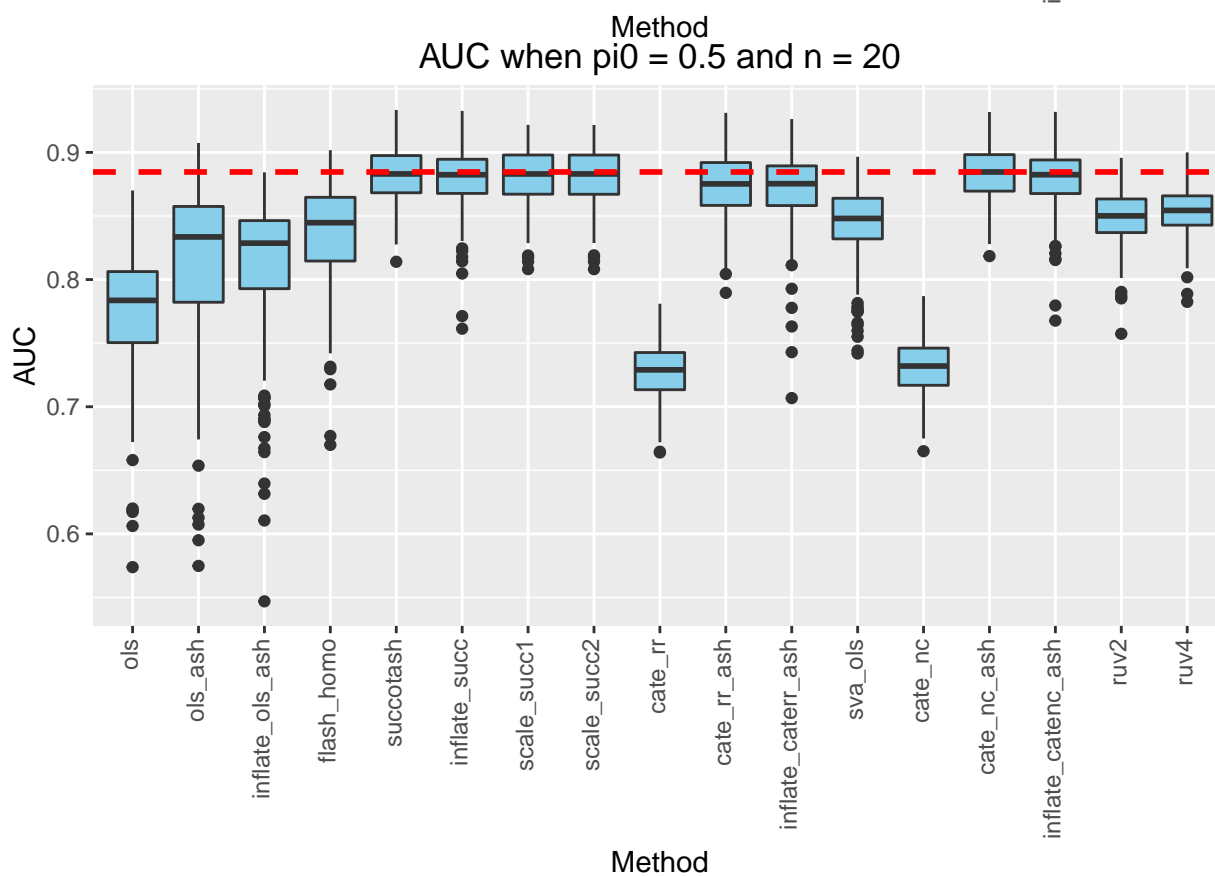
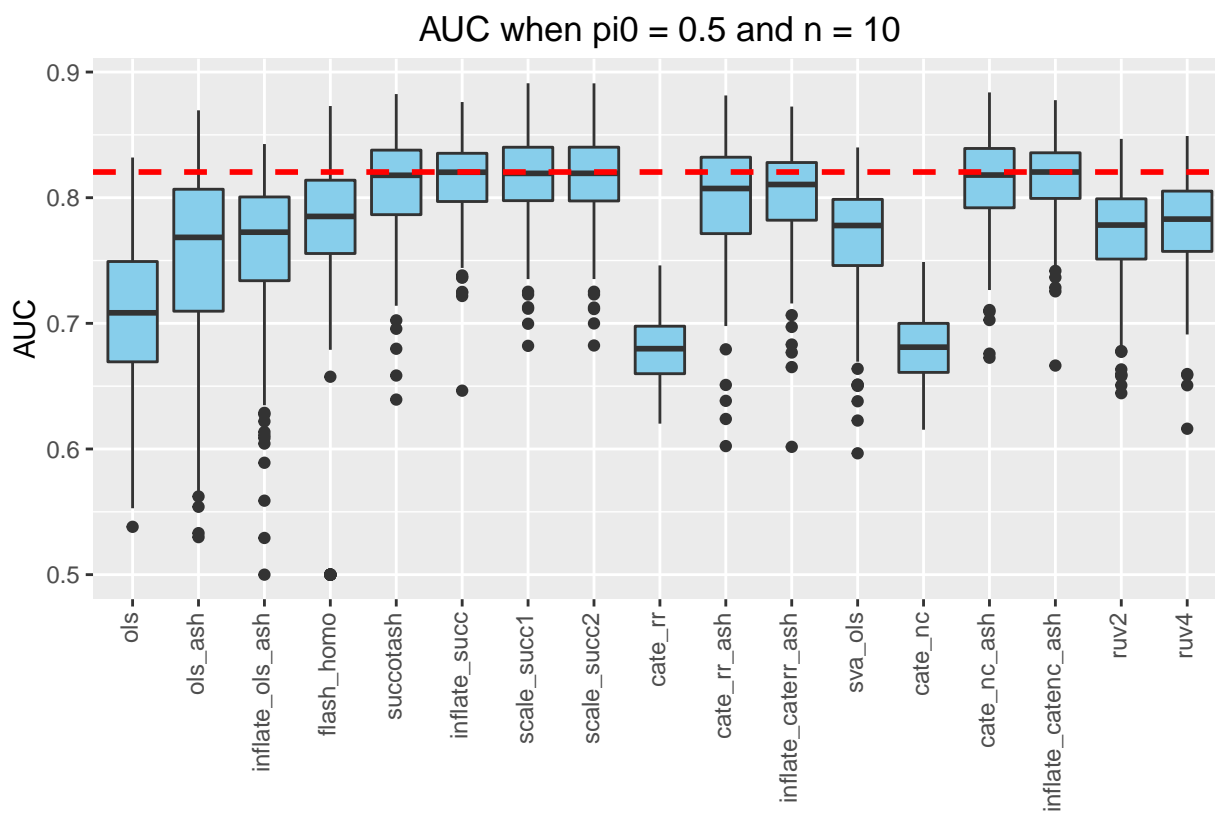
```
double_auc <- read.csv("../double_succ/auc_mat.csv")
reg_auc <- read.csv("../flash_v_rest_using_package/auc_mat.csv")
scale_auc <- read.csv("auc_ssuc.csv")
reg_auc$inflate_succ <- double_auc$succotash
reg_auc$inflate_caterr_ash <- double_auc$cate_rr_ash
reg_auc$inflate_catenc_ash <- double_auc$cate_nc_ash
reg_auc$inflate_ols_ash <- double_auc$ols_ash
reg_auc$scale_succ1 <- scale_auc$scale_suc1
reg_auc$scale_succ2 <- scale_auc$scale_suc2
reg_auc <- tbl_df(reg_auc)
reg_auc <- reg_auc[, c(1:2, 17, 3:4, 14, 18:19, 5:6, 15, 7:9, 16, 10:13)]
nsamp_seq <- unique(reg_auc$nsamp)
nullpi_seq <- unique(reg_auc$nullpi)
for (current_pi in nullpi_seq) {
  for (current_nsamp in nsamp_seq) {

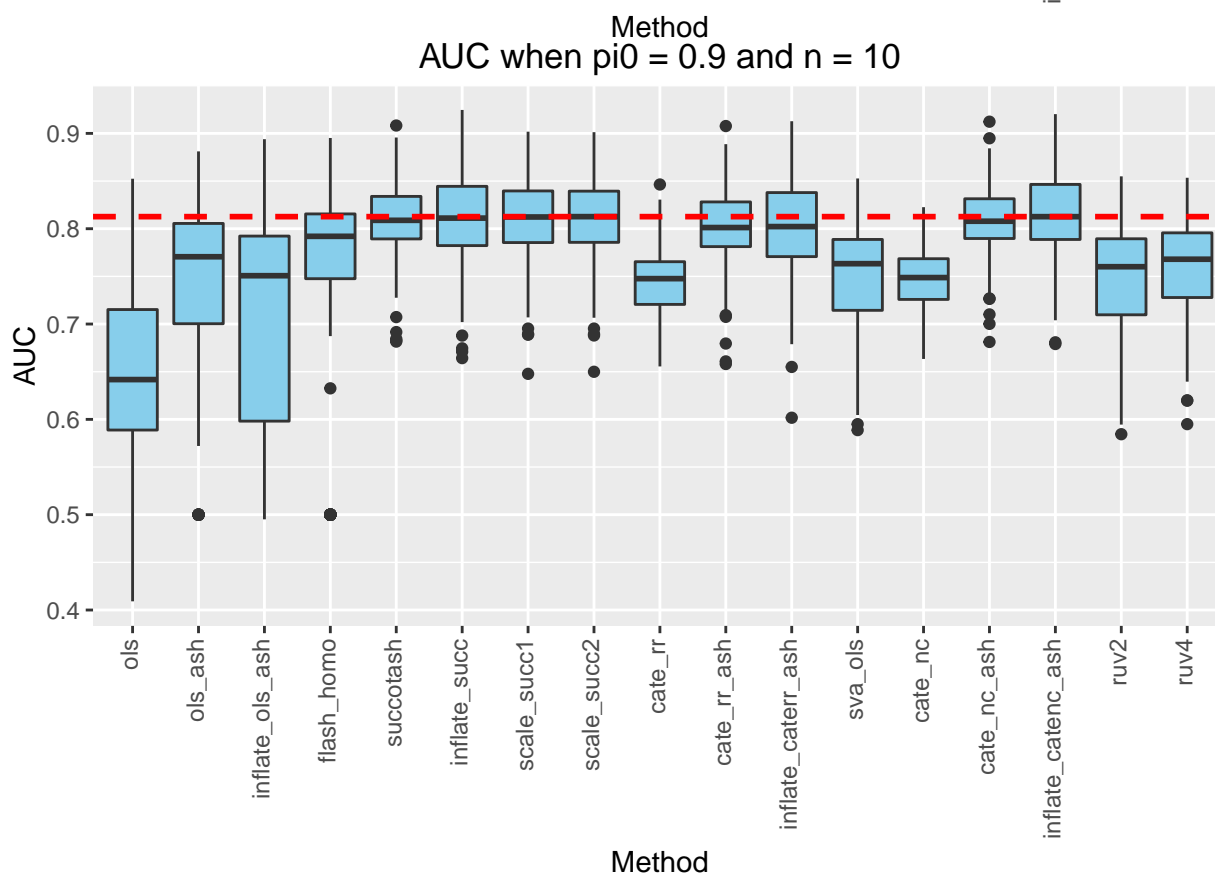
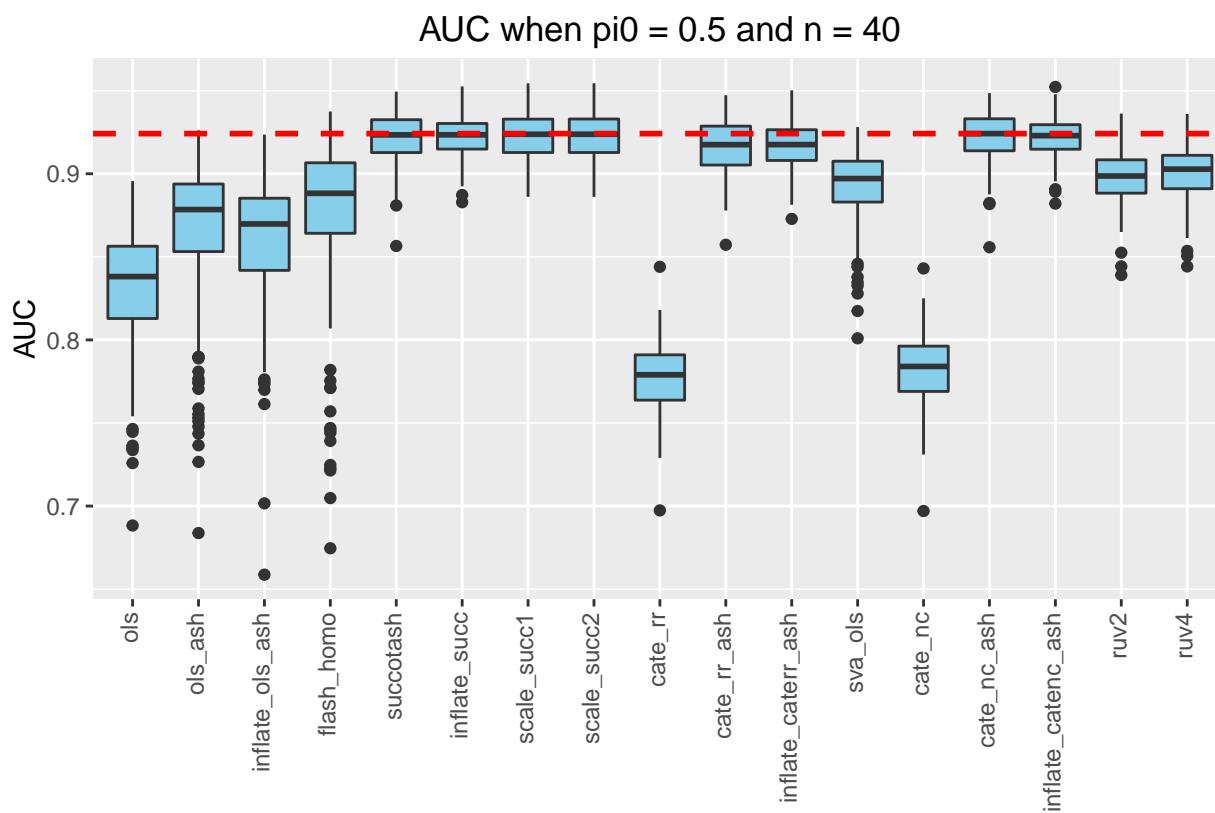
    subdf <- select(
      filter(
        reg_auc, nullpi == current_pi & nsamp == current_nsamp),
      -c(nsamp, nullpi)
    )

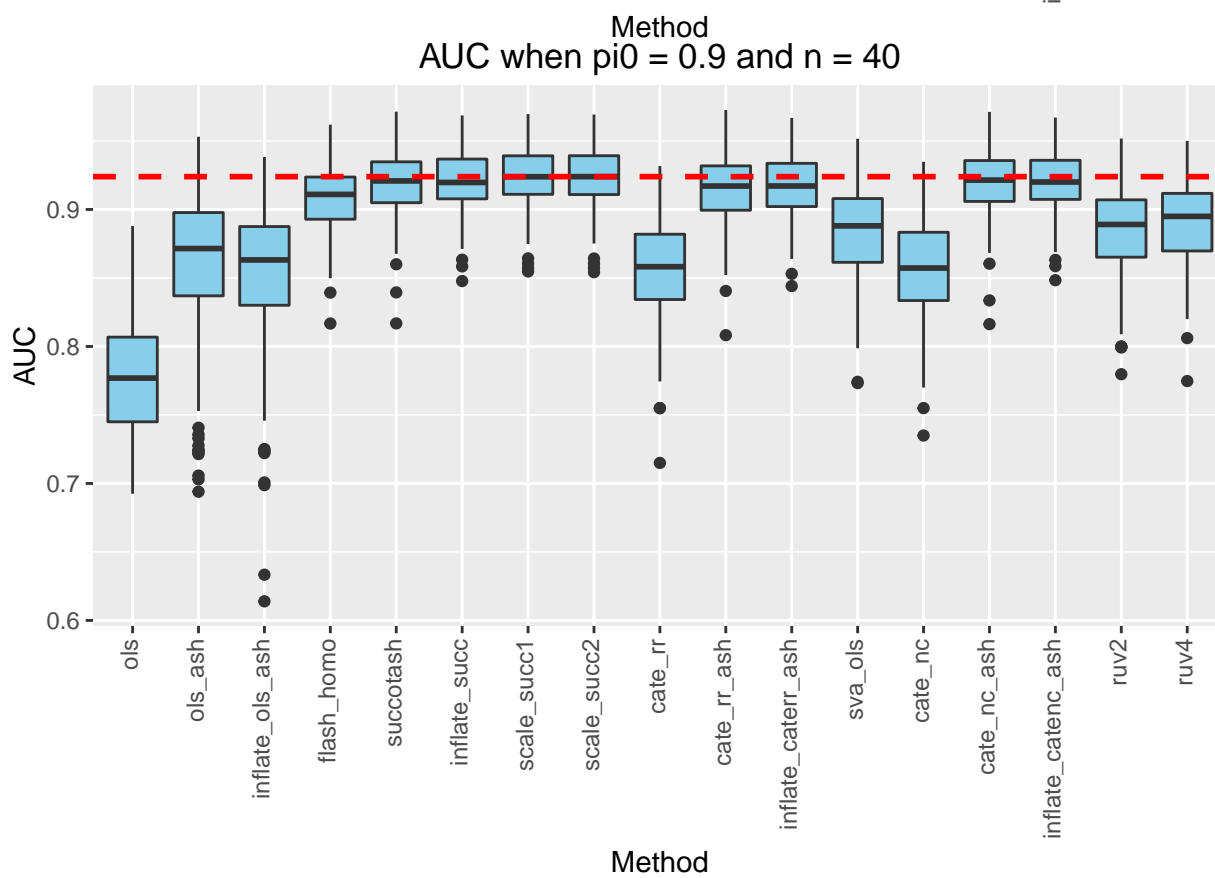
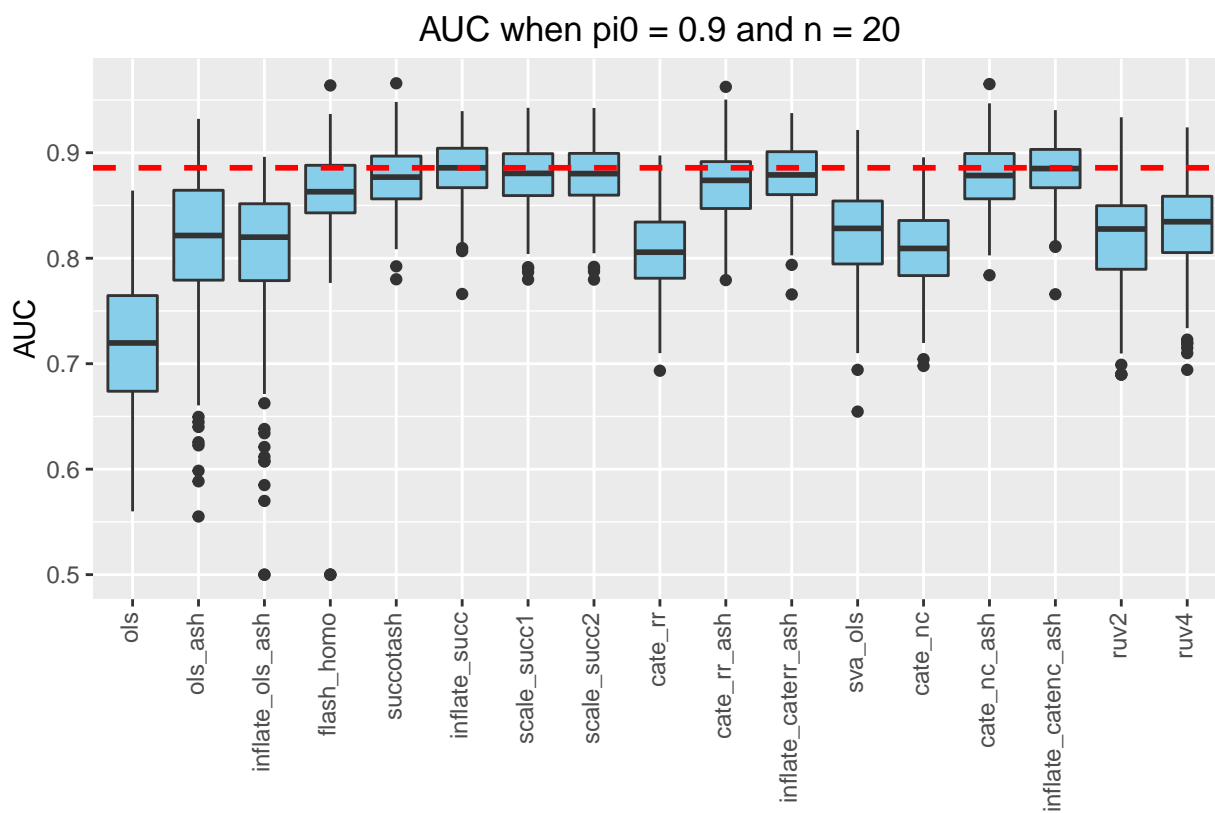
    hval <- max(apply(subdf, 2, median))

    melted_df <- melt(subdf, id.vars = NULL)

    p <- ggplot(data = melted_df, mapping = aes(x = variable, y = value)) +
      geom_boxplot(fill = I("skyblue")) +
      xlab(label = "Method") + ylab(label = "AUC") +
      geom_hline(yintercept = hval, color = I("red"), lty = 2, lwd = 1) +
      ggtitle(paste("AUC when pi0 =", current_pi, "and n =", current_nsamp * 2)) +
      theme(axis.text.x = element_text(angle = 90, hjust = 1, vjust = 0.3))
    print(p)
  }
}
```







Log(MSE + 0.1) Plots

```

nsamp_seq <- unique(reg_mse$nsamp)
nullpi_seq <- unique(reg_mse$nullpi)
for (current_pi in nullpi_seq) {
  for (current_nsamp in nsamp_seq) {

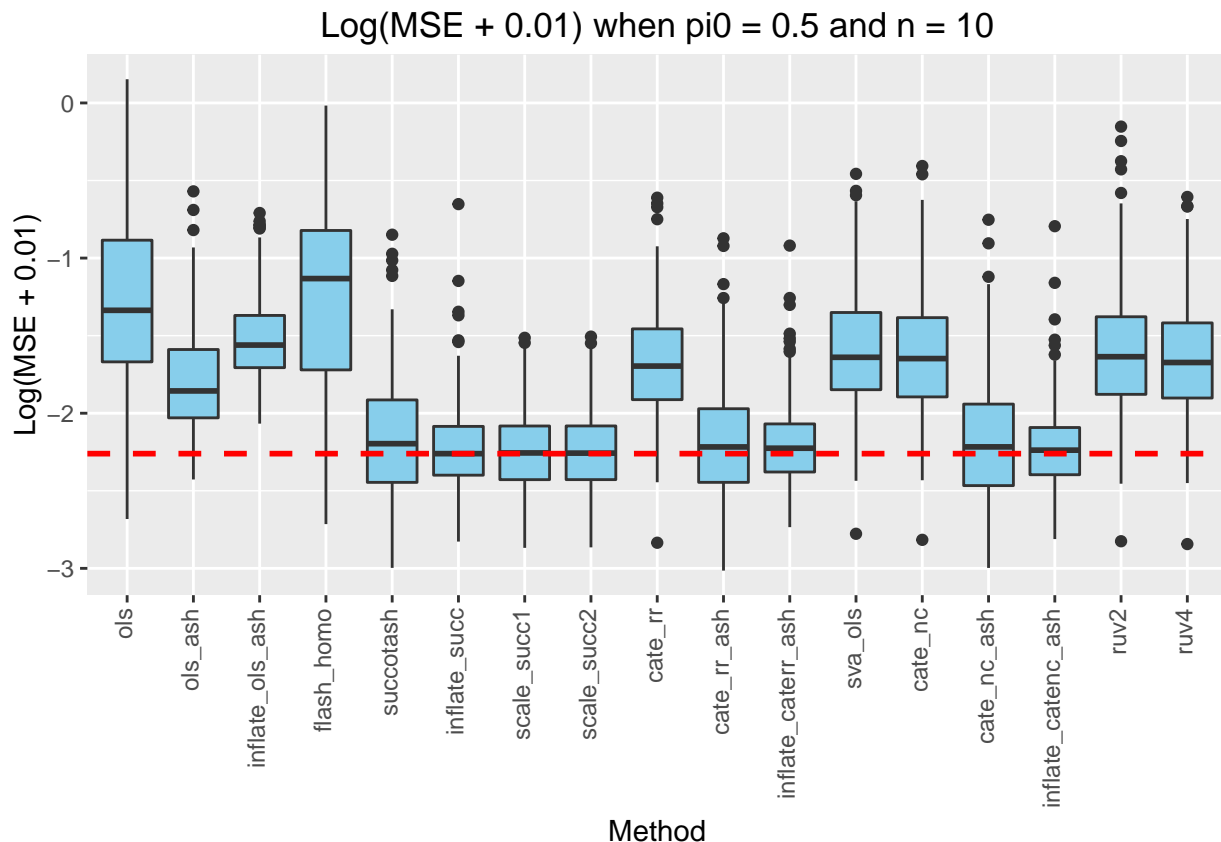
    subdf <- select(
      filter(
        reg_mse, nullpi == current_pi & nsamp == current_nsamp),
      -c(nsamp, nullpi)
    )

    hval <- min(apply(log(subdf + 0.01), 2, median))

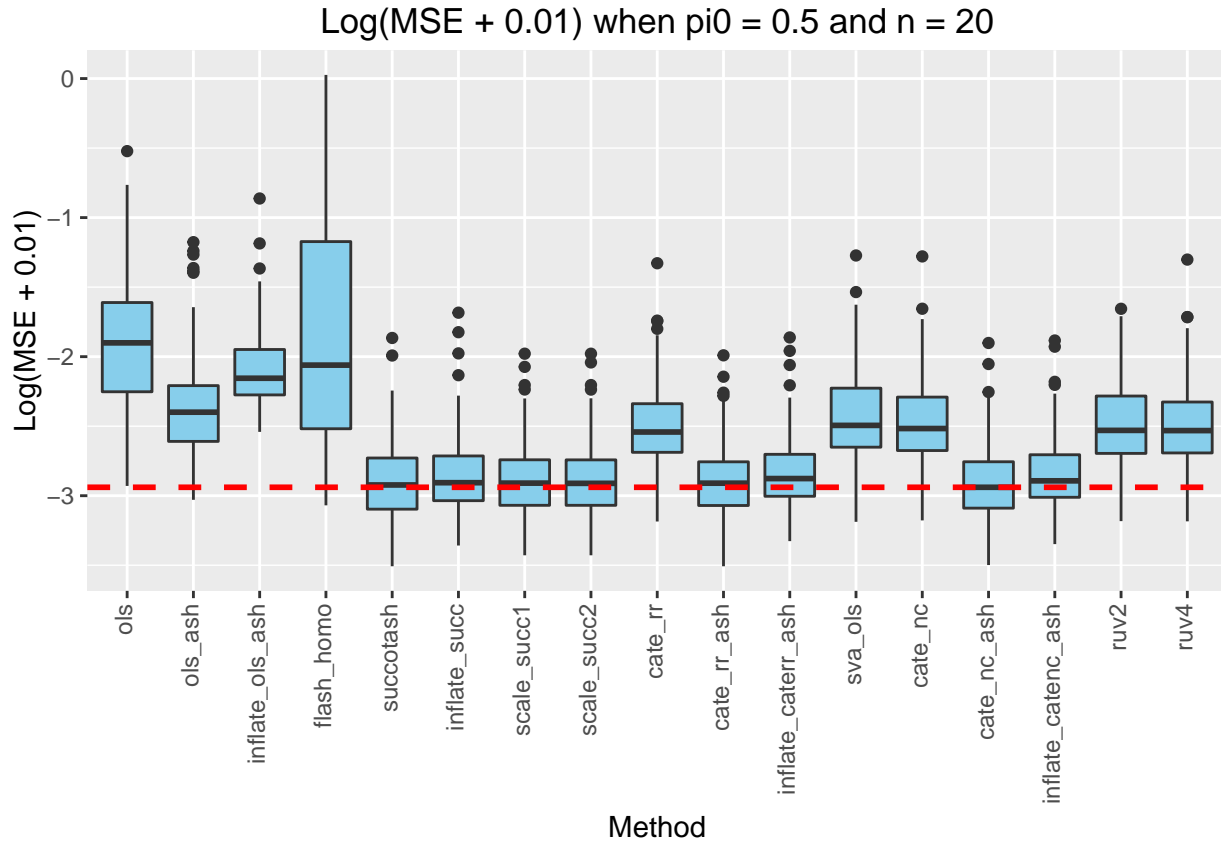
    melted_df <- melt(subdf, id.vars = NULL)

    p <- ggplot(data = melted_df, mapping = aes(x = variable, y = log(value + 0.01))) +
      geom_boxplot(fill = I("skyblue")) +
      xlab(label = "Method") + ylab(label = "Log(MSE + 0.01)") +
      geom_hline(yintercept = hval, color = I("red"), lty = 2, lwd = 1) +
      ggtitle(paste("Log(MSE + 0.01) when pi0 =", current_pi, "and n =", current_nsamp * 2)) +
      theme(axis.text.x = element_text(angle = 90, hjust = 1, vjust = 0.3))
    print(p)
  }
}

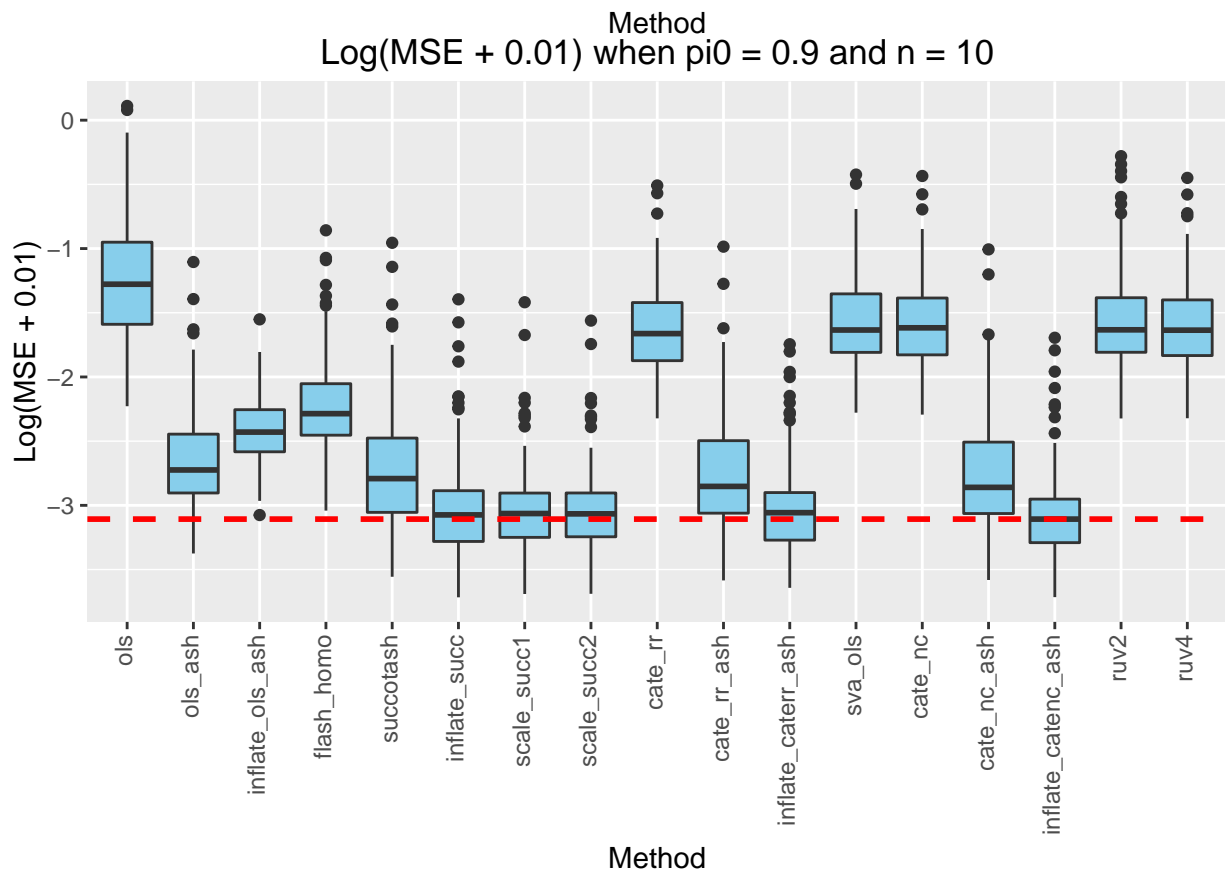
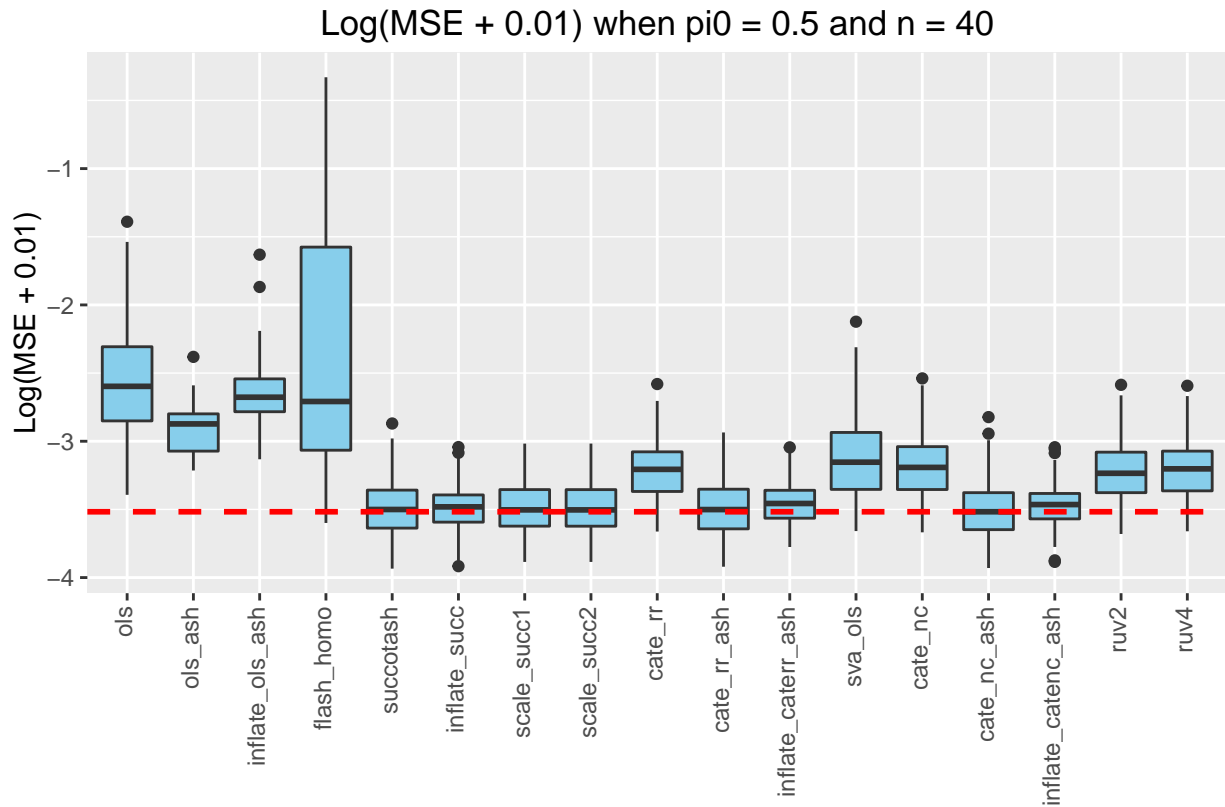
```



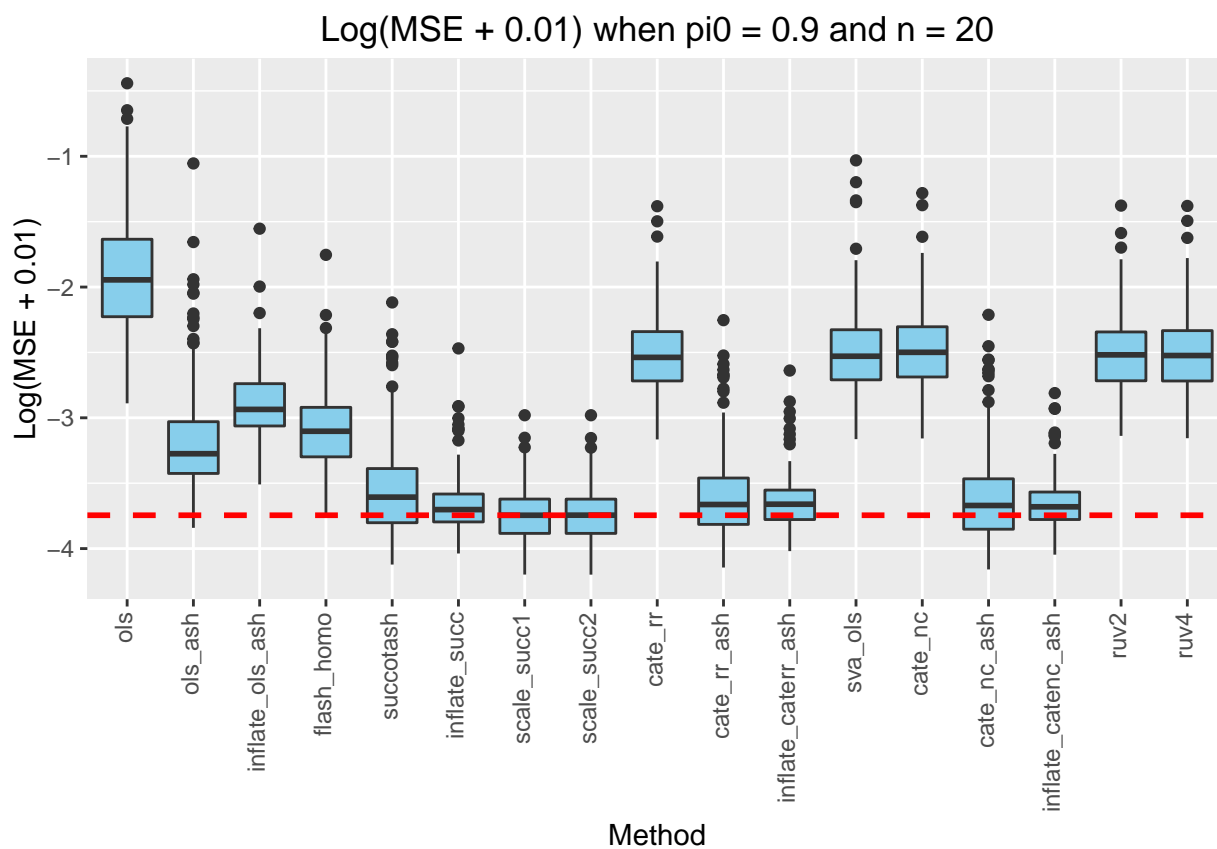
Warning: Removed 5 rows containing non-finite values (stat_boxplot).



Warning: Removed 203 rows containing non-finite values (stat_boxplot).

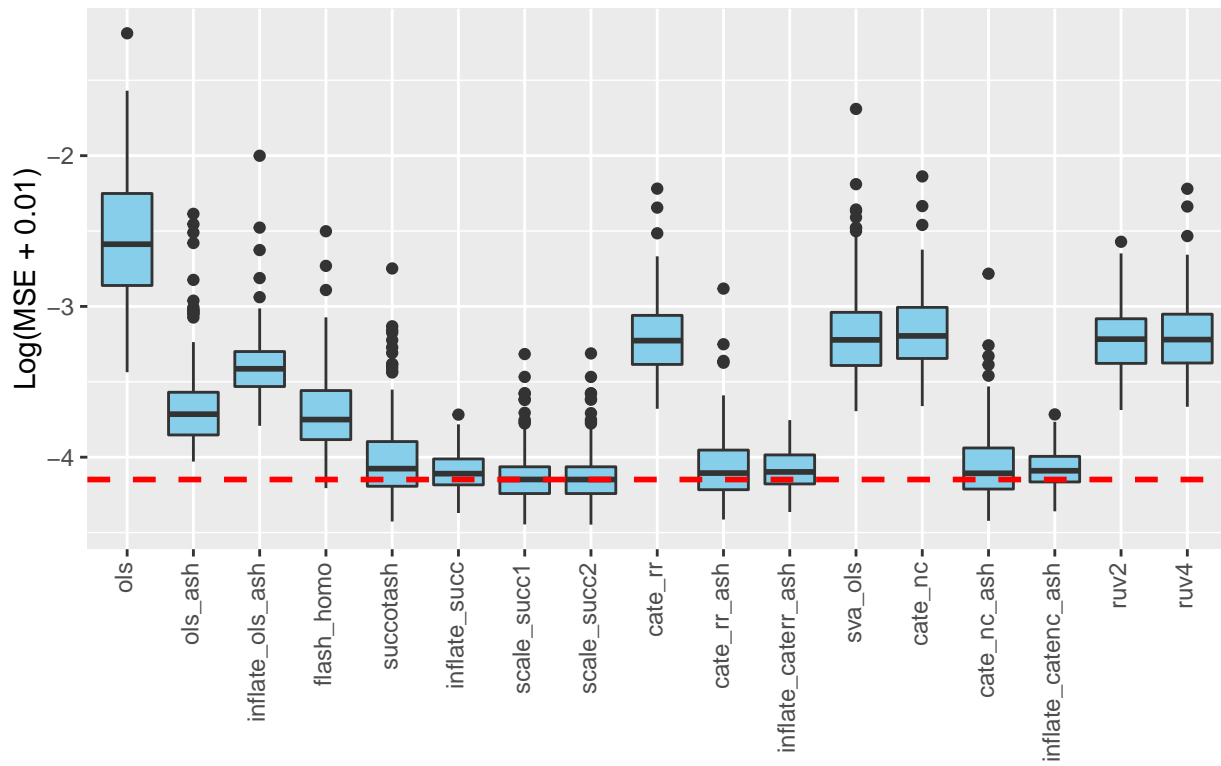


Warning: Removed 1 rows containing non-finite values (stat_boxplot).

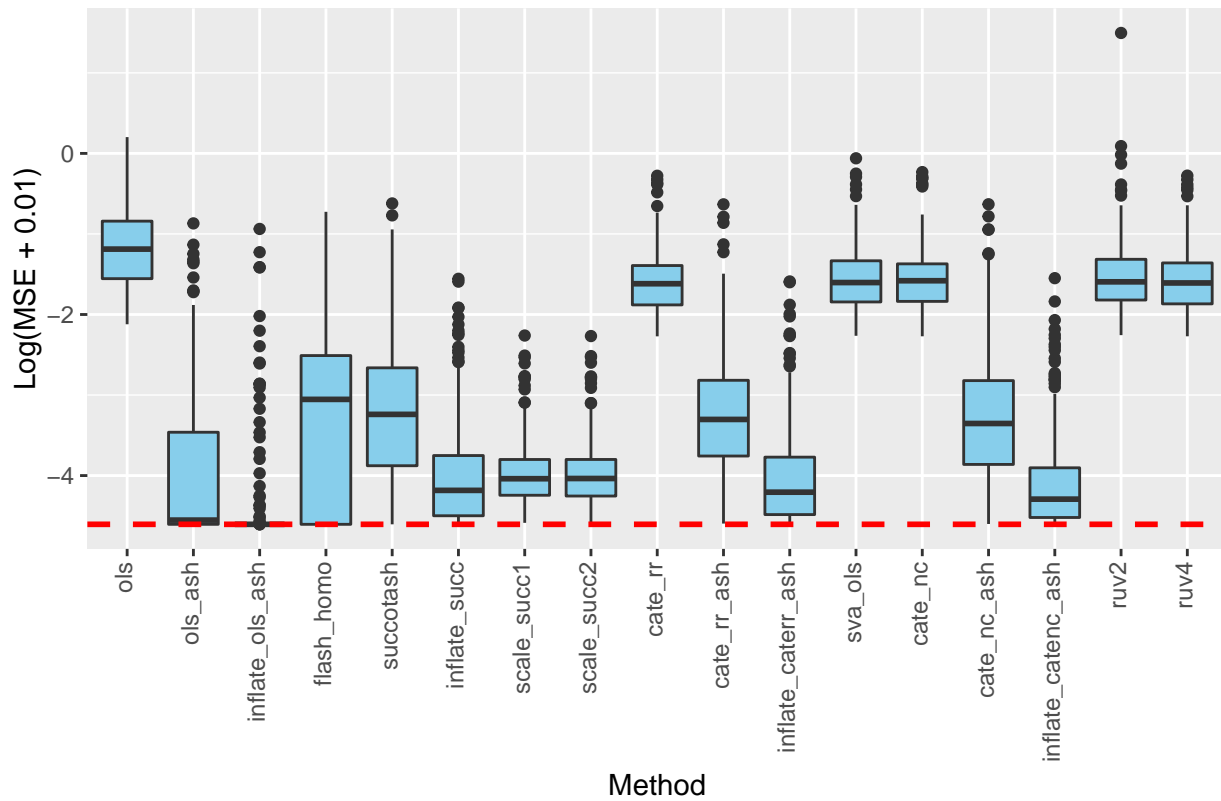


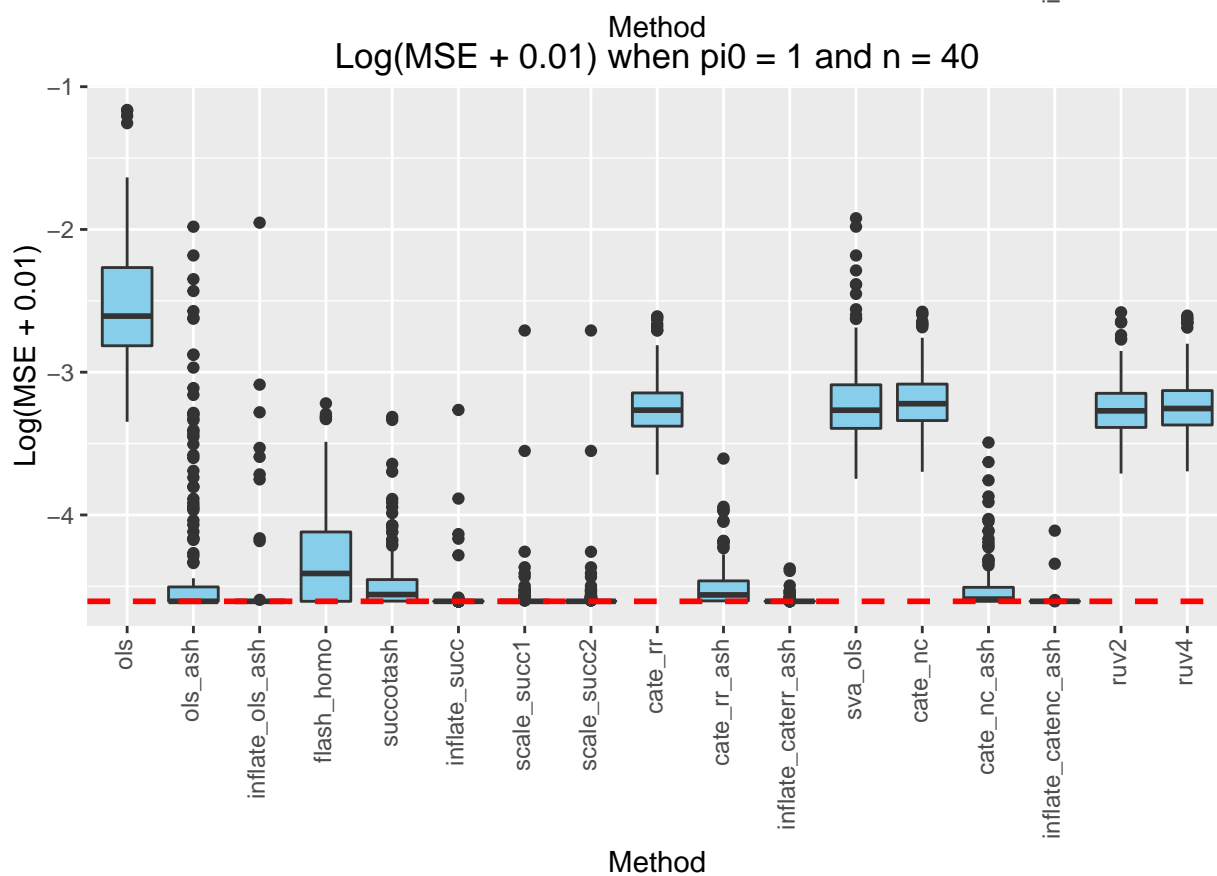
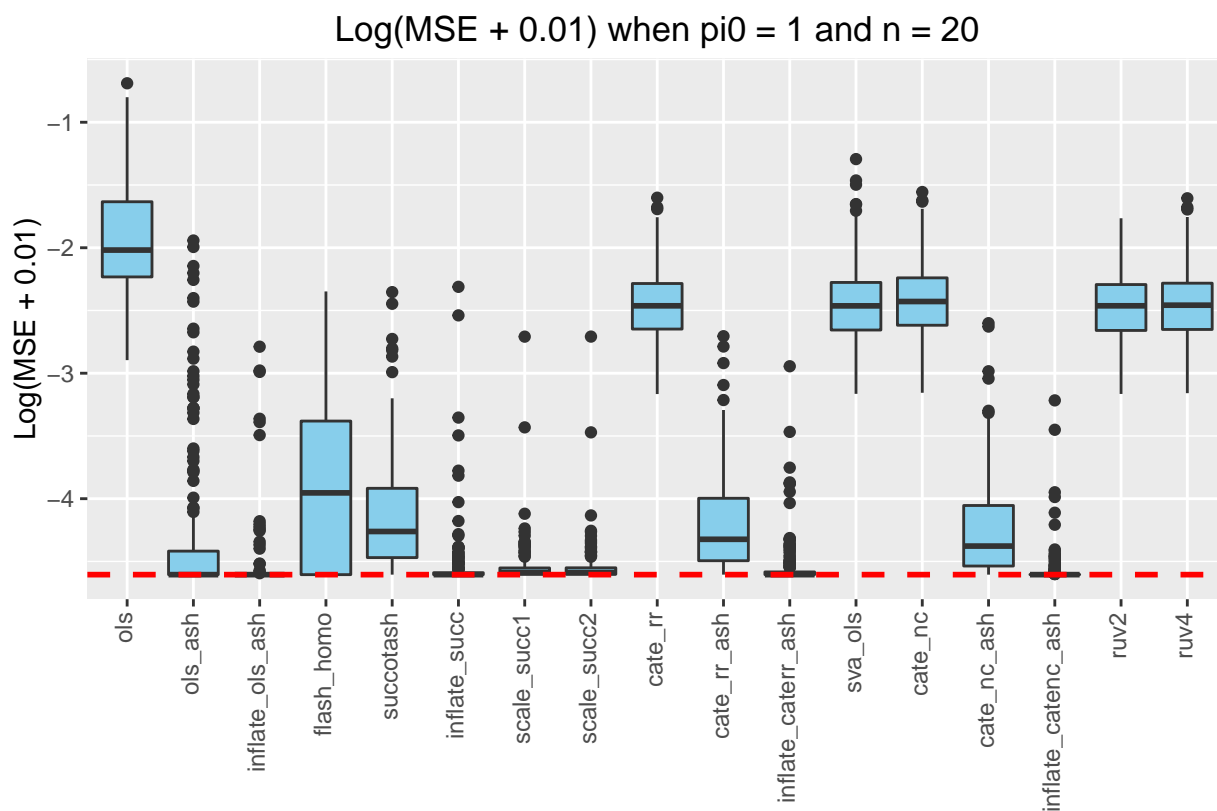
Warning: Removed 89 rows containing non-finite values (stat_boxplot).

Log(MSE + 0.01) when $\pi_0 = 0.9$ and $n = 40$



Method
Log(MSE + 0.01) when $\pi_0 = 1$ and $n = 10$






```
sessionInfo()
```

```
## R version 3.2.5 (2016-04-14)
## Platform: x86_64-pc-linux-gnu (64-bit)
## Running under: Ubuntu 14.04.4 LTS
##
## locale:
##  [1] LC_CTYPE=en_US.UTF-8      LC_NUMERIC=C
##  [3] LC_TIME=en_US.UTF-8      LC_COLLATE=en_US.UTF-8
##  [5] LC_MONETARY=en_US.UTF-8  LC_MESSAGES=en_US.UTF-8
##  [7] LC_PAPER=en_US.UTF-8     LC_NAME=C
##  [9] LC_ADDRESS=C             LC_TELEPHONE=C
## [11] LC_MEASUREMENT=en_US.UTF-8 LC_IDENTIFICATION=C
##
## attached base packages:
## [1] stats      graphics  grDevices  utils      datasets  methods    base
##
## other attached packages:
## [1] ggplot2_2.1.0  reshape2_1.4.1 dplyr_0.4.3   xtable_1.8-2
## [5] knitr_1.12.26
##
## loaded via a namespace (and not attached):
##  [1] Rcpp_0.12.4      magrittr_1.5      munsell_0.4.3
##  [4] colorspace_1.2-6 R6_2.1.1          highr_0.5.1
##  [7] stringr_1.0.0    plyr_1.8.3        tools_3.2.5
## [10] parallel_3.2.5   grid_3.2.5        gtable_0.2.0
## [13] DBI_0.3.1        htmltools_0.3.5   yaml_2.1.13
## [16] lazyeval_0.1.10  assertthat_0.1    digest_0.6.9
## [19] formatR_1.3      codetools_0.2-14  evaluate_0.8.3
## [22] rmarkdown_0.9.5.12 labeling_0.3       stringi_1.0-1
## [25] compiler_3.2.5   scales_0.4.0
```