Inflate Variance in Confounder Adjustment Methods

David Gerard 2016-04-26

Abstract

Here, I inflate the variance for SUCCOTASH, CATE + ASH, and OLS + ASH, and compare them to uninflated procedures. Inflating the variance for SUCCOTASH and CATE + ASH results in improved MSE and improved estimates of π_0 with no detriment to AUC. Inflating the variance for OLS + ASH does not seem to do much.

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.

I just multiplied the variance estimates by 2 (multiplied the standard deviation estimates by $\sqrt{2}$) for the inflated variance versions.

Inflating the variance in SUCCOTASH and CATE + ASH improve both the estimates of π_0 and the MSE. The AUC is almost the exact same for both the inflated and uninflated versions.

Interestingly, inflating the variance in OLS + ASH saw a much smaller improvement (if any). In particular, MSE got worse. The estimates of π_0 look a little better for each pair of π_0 and n, but the estimates have a very large left tail with about 10% of the values being deemed "outliers" in the boxplot for each combination.

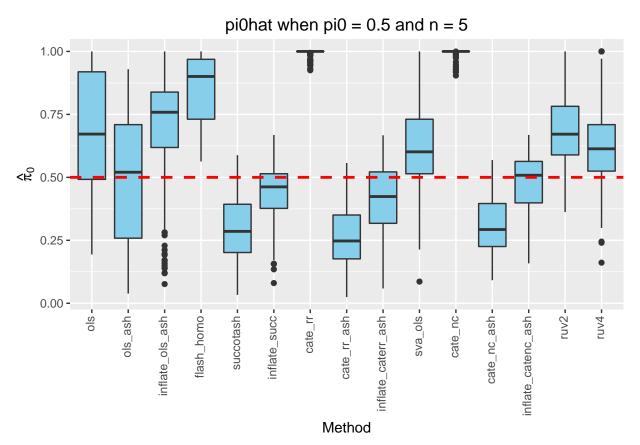
$\hat{\pi}_0$ Plots

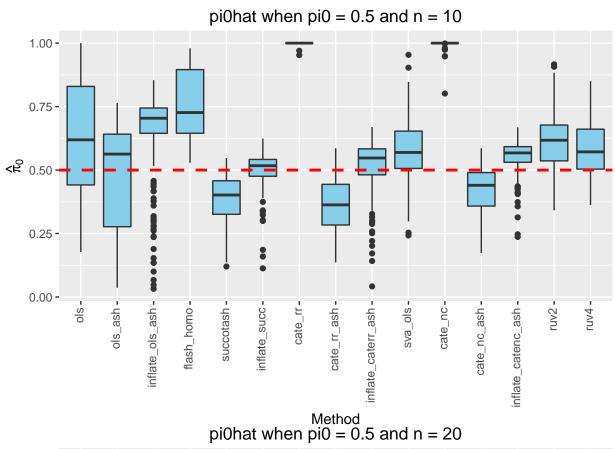
```
inflate_pi0 <- read.csv("pi0_mat.csv")
reg_pi0 <- read.csv("../flash_v_rest_using_package/pi0_mat.csv")
reg_pi0$inflate_succ <- inflate_pi0$succotash
reg_pi0$inflate_caterr_ash <- inflate_pi0$cate_rr_ash
reg_pi0$inflate_catenc_ash <- inflate_pi0$cate_nc_ash
reg_pi0$inflate_ols_ash <- inflate_pi0$ols_ash
reg_pi0 <- tbl_df(reg_pi0)
reg_pi0 <- reg_pi0[, c(1:2, 17, 3:4, 14, 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) {</pre>
```

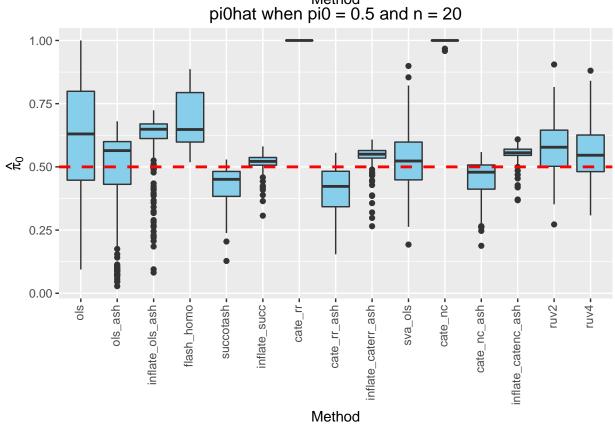
```
subdf <- select(
    filter(
        reg_pi0, nullpi == current_pi & nsamp == current_nsamp),
        -c(nsamp, nullpi)
)

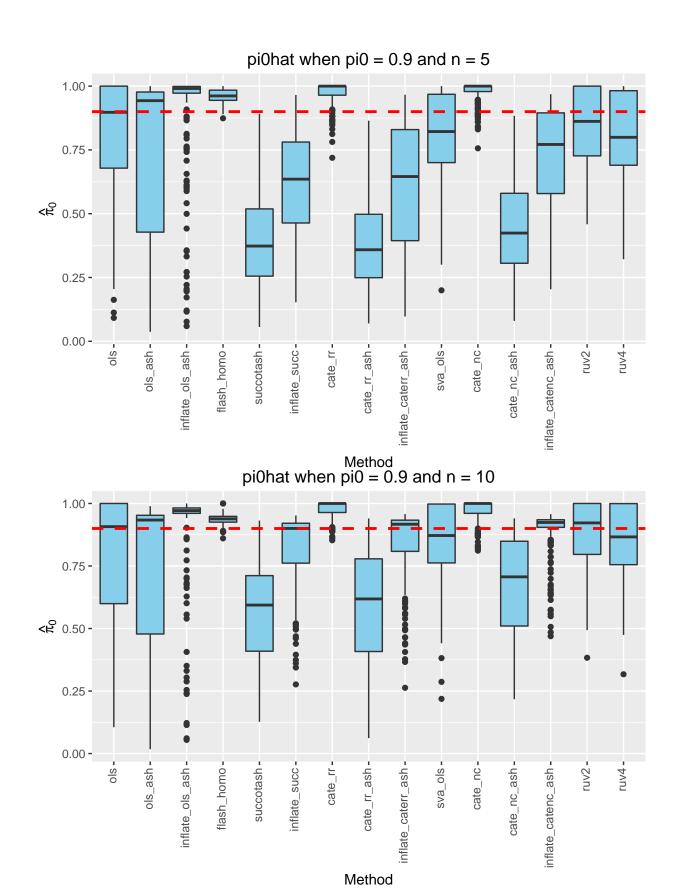
melted_df <- melt(subdf, id.vars = NULL)

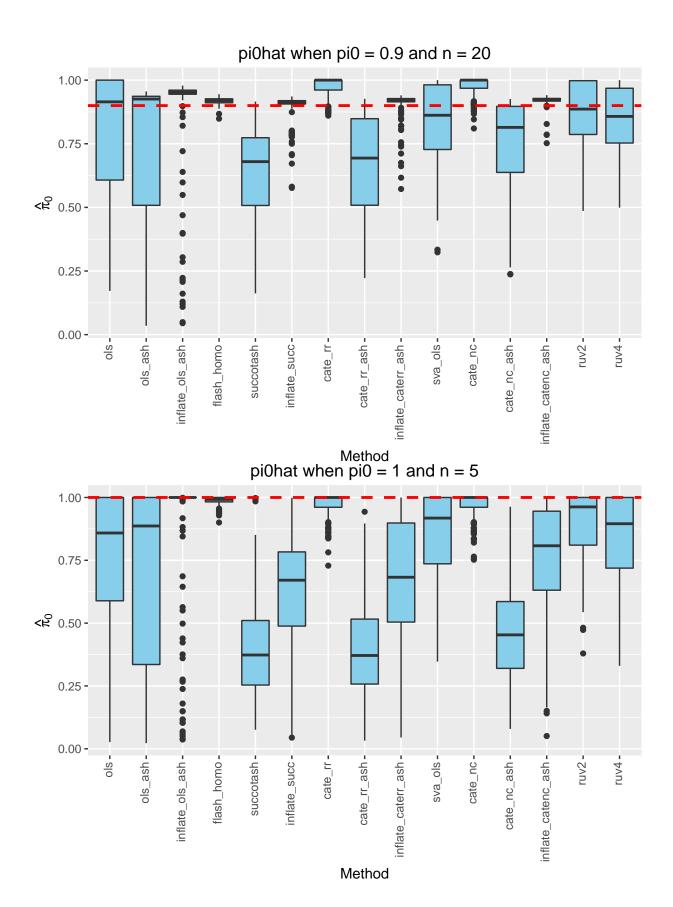
p <- ggplot(data = melted_df, mapping = aes(x = variable, y = value)) +
        geom_boxplot(fill = I("skyblue")) +
        xlab(label = "Method") + ylab(label = expression(hat(pi)[0])) +
        geom_hline(yintercept = current_pi, color = I("red"), lty = 2, lwd = 1) +
        ggtitle(paste("pi0hat when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

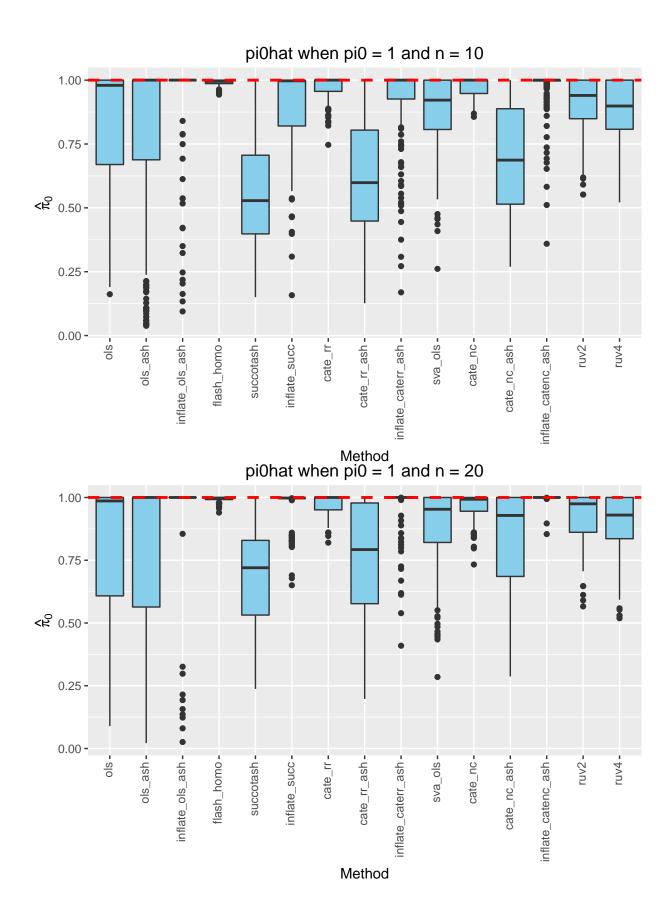






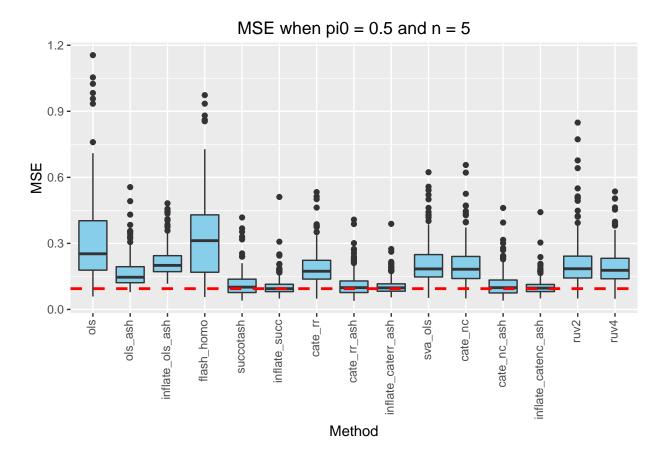




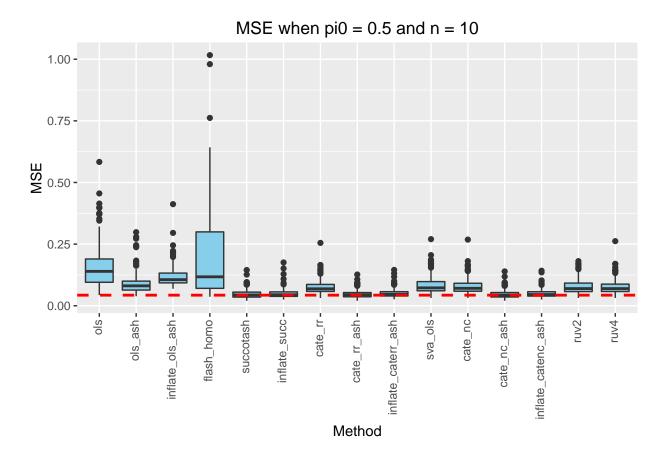


MSE Plots

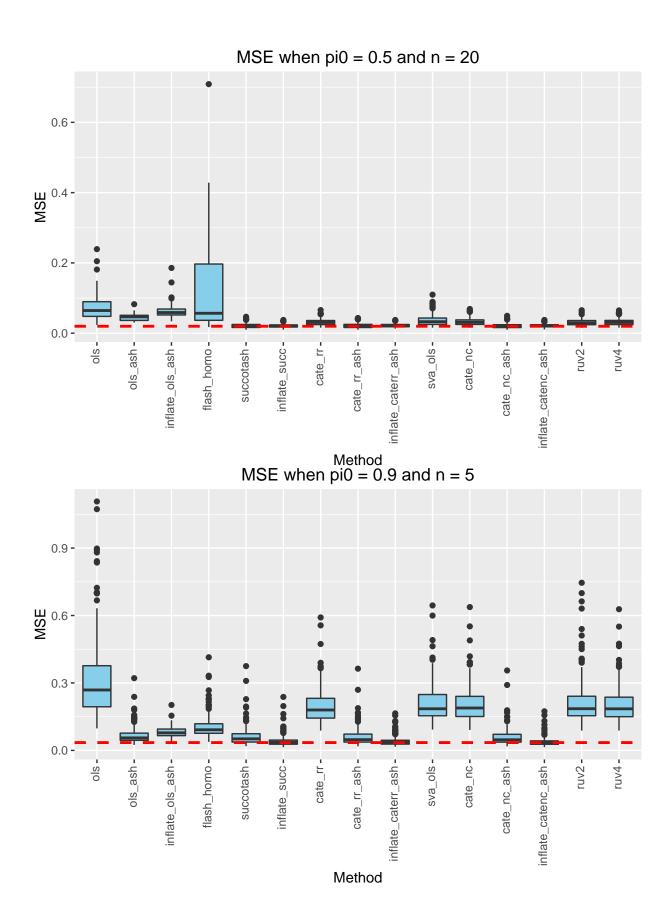
```
inflate mse <- read.csv("mse mat.csv")</pre>
reg_mse <- read.csv("../flash_v_rest_using_package/mse_mat.csv")</pre>
reg_mse$inflate_succ <- inflate_mse$succotash</pre>
reg_mse$inflate_caterr_ash <- inflate_mse$cate_rr_ash</pre>
reg mse$inflate catenc ash <- inflate mse$cate nc ash
reg_mse$inflate_ols_ash <- inflate_mse$ols_ash</pre>
reg_mse <- tbl_df(reg_mse)</pre>
reg_mse <- reg_mse[, c(1:2, 17, 3:4, 14, 5:6, 15, 7:9, 16, 10:13)]
nsamp_seq <- unique(reg_mse$nsamp)</pre>
nullpi_seq <- unique(reg_mse$nullpi)</pre>
for (current_pi in nullpi_seq) {
    for (current_nsamp in nsamp_seq) {
        subdf <- select(</pre>
            filter(
                 reg_mse, nullpi == current_pi & nsamp == current_nsamp),
             -c(nsamp, nullpi)
        )
        hval <- min(apply(subdf, 2, median))</pre>
        melted df <- melt(subdf, id.vars = NULL)</pre>
        p <- ggplot(data = melted_df, mapping = aes(x = variable, y = value)) +</pre>
             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)) +
             theme(axis.text.x = element_text(angle = 90, hjust = 1, vjust = 0.3))
        print(p)
    }
}
```



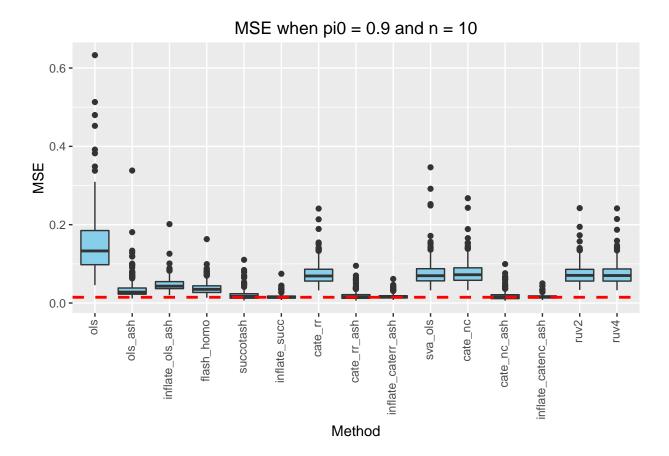
Warning: Removed 5 rows containing non-finite values (stat_boxplot).



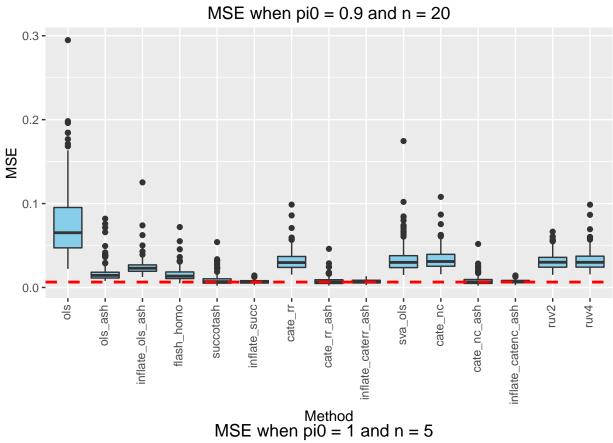
Warning: Removed 203 rows containing non-finite values (stat_boxplot).

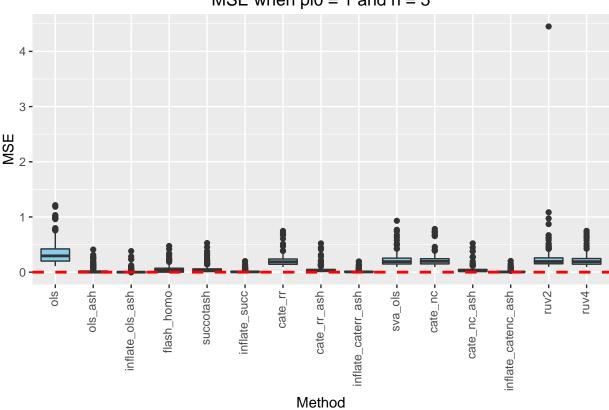


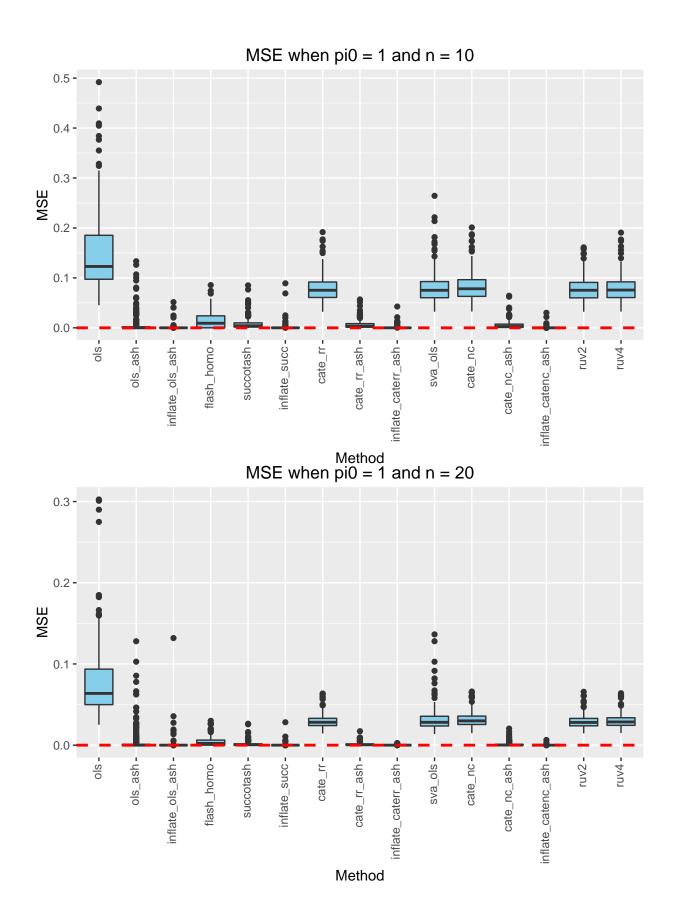
Warning: Removed 1 rows containing non-finite values (stat_boxplot).



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

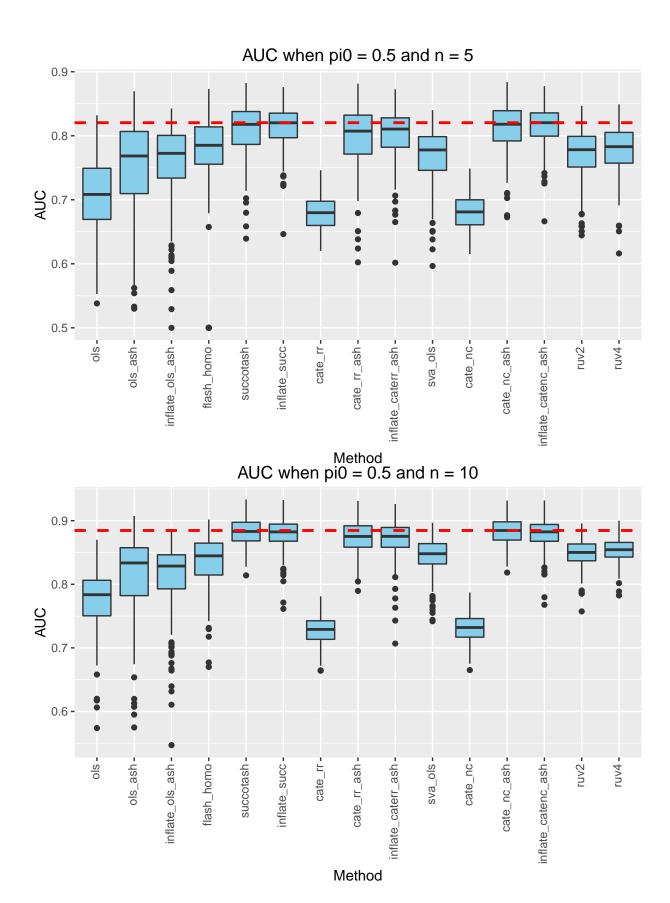


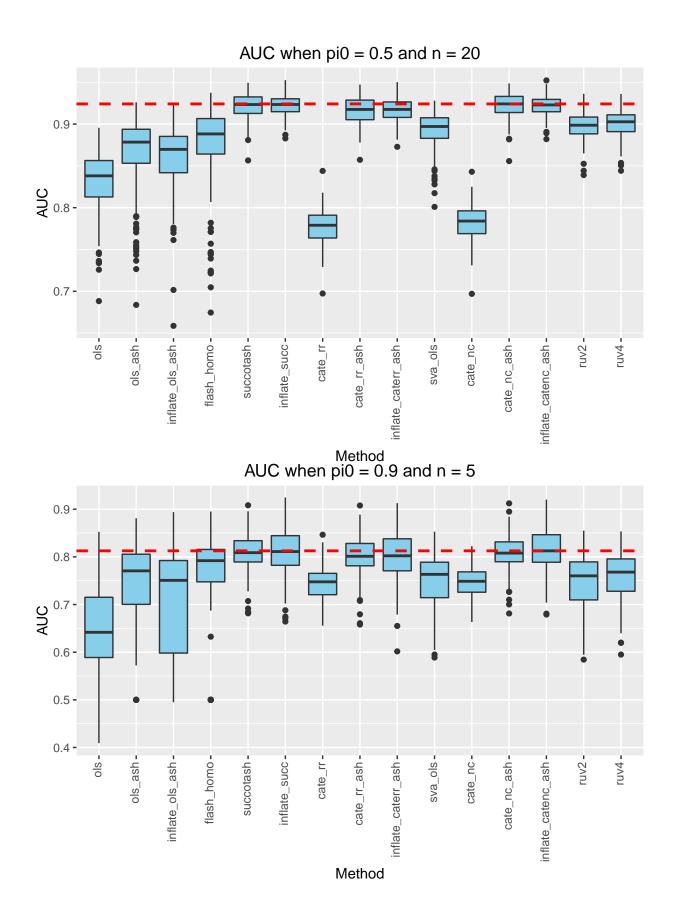


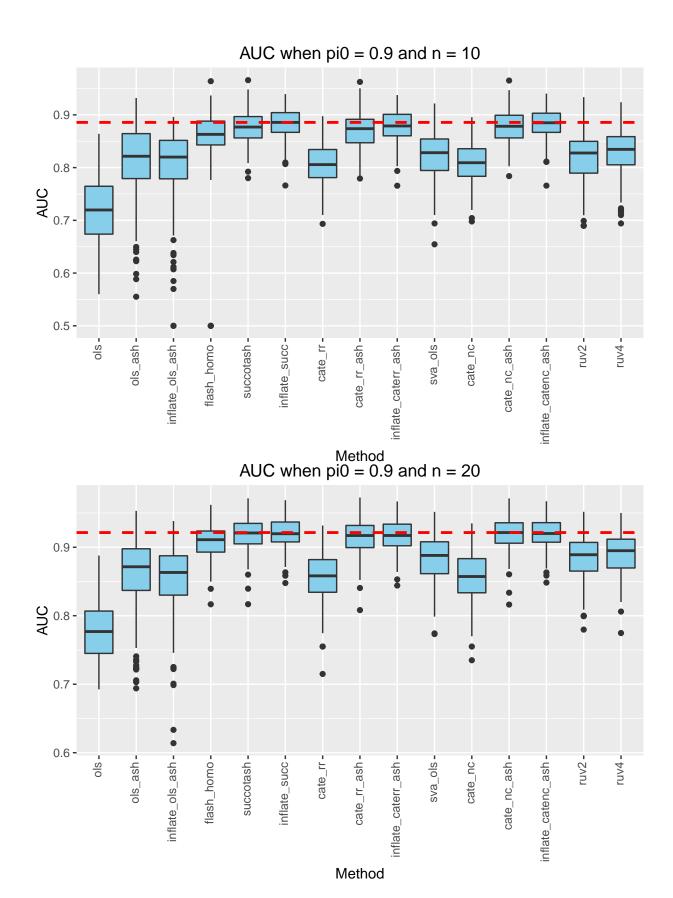


AUC Plots

```
inflate auc <- read.csv("auc mat.csv")</pre>
reg_auc <- read.csv("../flash_v_rest_using_package/auc_mat.csv")</pre>
reg_auc$inflate_succ <- inflate_auc$succotash</pre>
reg_auc$inflate_caterr_ash <- inflate_auc$cate_rr_ash</pre>
reg auc$inflate catenc ash <- inflate auc$cate nc ash
reg_auc$inflate_ols_ash <- inflate_auc$ols_ash</pre>
reg_auc <- tbl_df(reg_auc)</pre>
reg_auc <- reg_auc[, c(1:2, 17, 3:4, 14, 5:6, 15, 7:9, 16, 10:13)]
nsamp_seq <- unique(reg_auc$nsamp)</pre>
nullpi_seq <- unique(reg_auc$nullpi)</pre>
for (current_pi in nullpi_seq) {
    for (current_nsamp in nsamp_seq) {
        subdf <- select(</pre>
            filter(
                 reg_auc, nullpi == current_pi & nsamp == current_nsamp),
             -c(nsamp, nullpi)
        )
        hval <- max(apply(subdf, 2, median))</pre>
        melted df <- melt(subdf, id.vars = NULL)</pre>
        p <- ggplot(data = melted_df, mapping = aes(x = variable, y = value)) +</pre>
             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)) +
             theme(axis.text.x = element_text(angle = 90, hjust = 1, vjust = 0.3))
        print(p)
    }
}
```

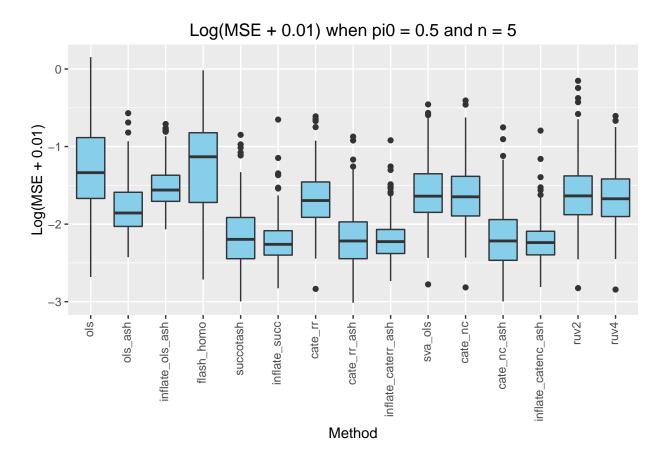




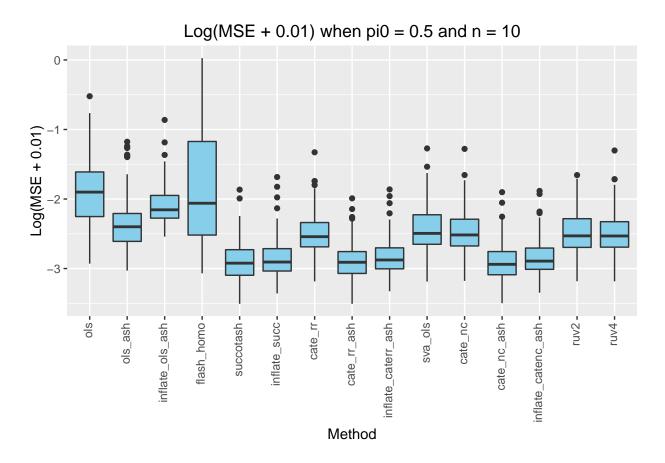


Log(MSE + 0.01) Plots

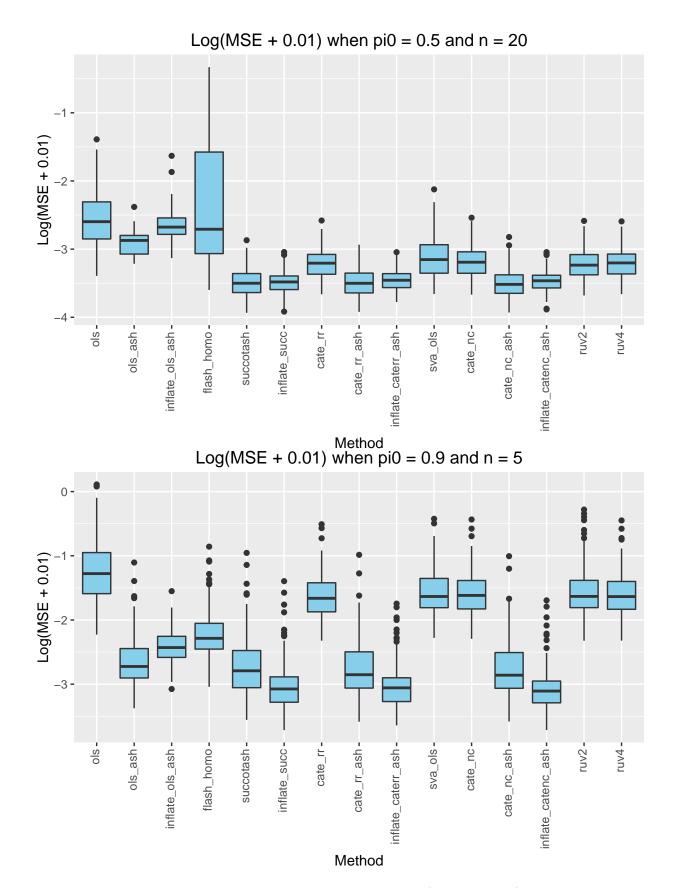
```
inflate_mse <- read.csv("mse_mat.csv")</pre>
reg_mse <- read.csv("../flash_v_rest_using_package/mse_mat.csv")</pre>
reg_mse$inflate_succ <- inflate_mse$succotash</pre>
reg mse$inflate caterr ash <- inflate mse$cate rr ash</pre>
reg_mse$inflate_catenc_ash <- inflate_mse$cate_nc_ash</pre>
reg_mse$inflate_ols_ash <- inflate_mse$ols_ash</pre>
reg_mse <- tbl_df(reg_mse)</pre>
reg_mse <- reg_mse[, c(1:2, 17, 3:4, 14, 5:6, 15, 7:9, 16, 10:13)]
nsamp_seq <- unique(reg_mse$nsamp)</pre>
nullpi_seq <- unique(reg_mse$nullpi)</pre>
for (current_pi in nullpi_seq) {
    for (current_nsamp in nsamp_seq) {
        subdf <- select(</pre>
             filter(
                 reg_mse, nullpi == current_pi & nsamp == current_nsamp),
             -c(nsamp, nullpi)
        )
        melted_df <- melt(subdf, id.vars = NULL)</pre>
        melted_df$value <- log(melted_df$value + 0.01)</pre>
        p <- ggplot(data = melted_df, mapping = aes(x = variable, y = value)) +</pre>
             geom_boxplot(fill = I("skyblue")) +
             xlab(label = "Method") + ylab(label = "Log(MSE + 0.01)") +
             ggtitle(paste("Log(MSE + 0.01) when pi0 =", current_pi, "and n =", current_nsamp)) +
             theme(axis.text.x = element_text(angle = 90, hjust = 1, vjust = 0.3))
        print(p)
}
```



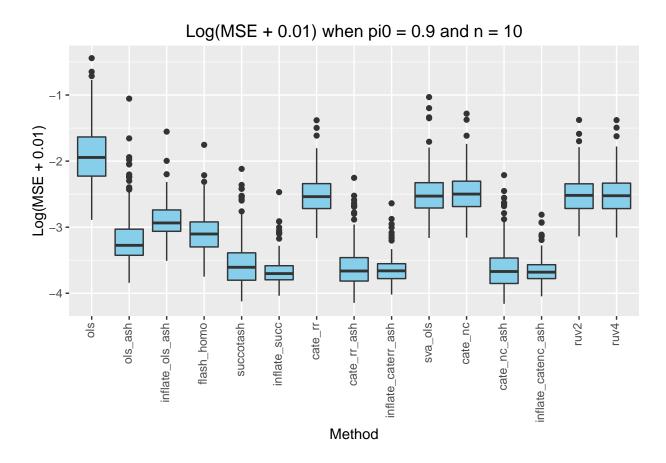
Warning: Removed 5 rows containing non-finite values (stat_boxplot).



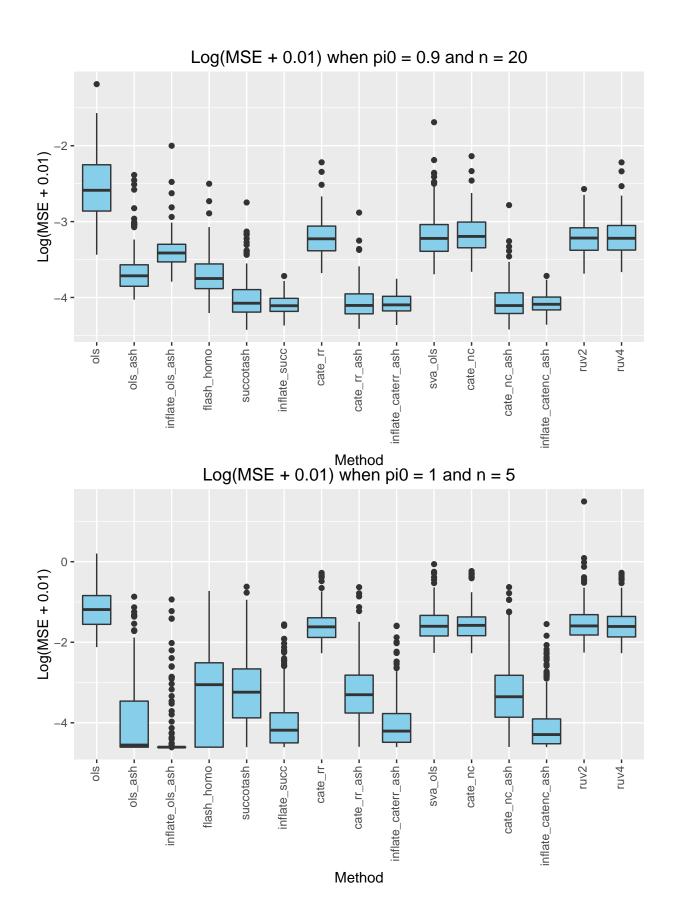
Warning: Removed 203 rows containing non-finite values (stat_boxplot).

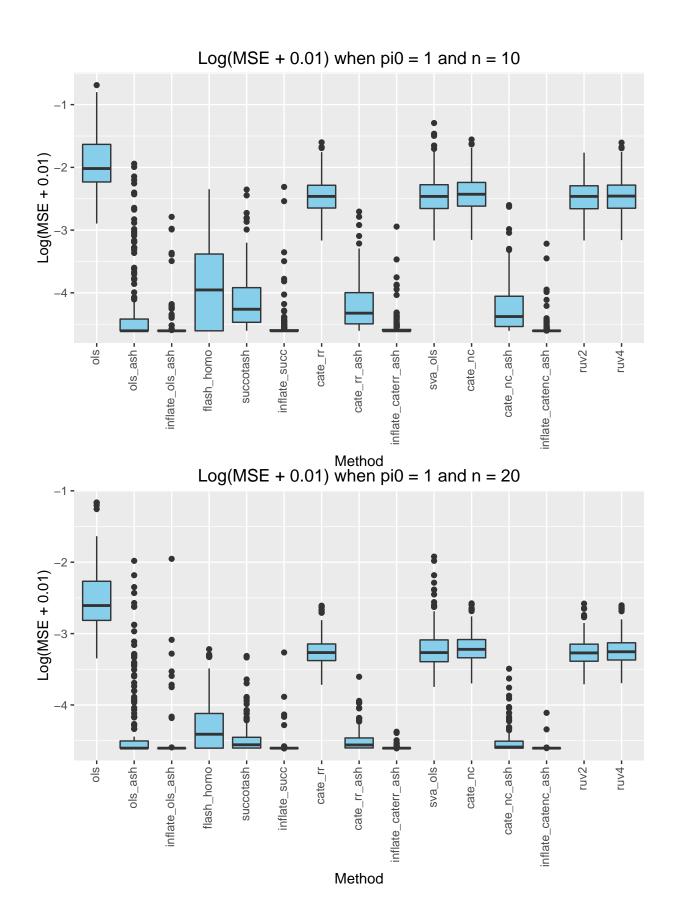


Warning: Removed 1 rows containing non-finite values (stat_boxplot).



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





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:
                    knitr_1.12.24 ggplot2_2.1.0 reshape2_1.4.1
## [1] xtable 1.8-2
## [5] dplyr_0.4.3
##
## loaded via a namespace (and not attached):
## [1] Rcpp_0.12.2
                         magrittr_1.5
                                           munsell_0.4.3
## [4] colorspace_1.2-6 R6_2.1.1
                                            stringr_1.0.0
## [7] plyr_1.8.3
                         tools_3.2.5
                                           parallel_3.2.5
## [10] grid_3.2.5
                         gtable_0.2.0
                                           DBI_0.3.1
## [13] htmltools_0.3.5
                         yaml_2.1.13
                                            lazyeval_0.1.10
## [16] assertthat_0.1
                         digest_0.6.9
                                            formatR_1.3
## [19] codetools_0.2-14 evaluate_0.8.3
                                           rmarkdown_0.9.5.9
                         stringi_1.0-1
                                            compiler_3.2.5
## [22] labeling_0.3
## [25] scales_0.4.0
```