Supplementary File S1

Simulated data

This document contains the required steps to generate the results for section **3.1 Methods performance in simulated scenarios**. Please follow the instructions in materials and methods within the main manuscript to download and process the experimental data. Pull a Docker image with all the required packages installed (docker pull ecvpaper2024/ecv_results). # Data preparation.

Load required packages.

```
library("eCV")
Loading required package: idr
Loading required package: mvtnorm
Loading required package: future
Loading required package: future.apply
Loading required package: MatrixGenerics
Loading required package: matrixStats
Attaching package: 'MatrixGenerics'
The following objects are masked from 'package:matrixStats':
    colAlls, colAnyNAs, colAnys, colAvgsPerRowSet, colCollapse,
    colCounts, colCummaxs, colCummins, colCumprods, colCumsums,
    colDiffs, colIQRDiffs, colIQRs, colLogSumExps, colMadDiffs,
    colMads, colMaxs, colMeans2, colMedians, colMins, colOrderStats,
    colProds, colQuantiles, colRanges, colRanks, colSdDiffs, colSds,
    colSums2, colTabulates, colVarDiffs, colVars, colWeightedMads,
    colWeightedMeans, colWeightedMedians, colWeightedSds,
    colWeightedVars, rowAlls, rowAnyNAs, rowAnys, rowAvgsPerColSet,
    rowCollapse, rowCounts, rowCummaxs, rowCummins, rowCumprods,
    rowCumsums, rowDiffs, rowIQRDiffs, rowIQRs, rowLogSumExps,
    rowMadDiffs, rowMads, rowMeans2, rowMedians, rowMins,
    rowOrderStats, rowProds, rowQuantiles, rowRanges, rowRanks,
    rowSdDiffs, rowSds, rowSums2, rowTabulates, rowVarDiffs, rowVars,
    rowWeightedMads, rowWeightedMeans, rowWeightedMedians,
    rowWeightedSds, rowWeightedVars
                --- / --- \ / /
/ _ \ | | \ \ \ / /
| __/ | | ___ \ V /
\__| \ \__| \__/
        Enhanced Coefficient of Variation and IDR Extensions
                  for Reproducibility Assessment
```

```
This package provides extensions and alternative methods to IDR to
measure the reproducibility of omic data with an arbitrary number of
replicates. It introduces an enhanced Coefficient of Variation (eCV)
metric to assess the likelihood of omic features being reproducible.
require("tidyverse")
Loading required package: tidyverse
-- Attaching packages ----- tidyverse 1.3.2 --
v ggplot2 3.4.3 v purrr 1.0.2
v tibble 3.2.1 v dplyr 1.1.2
v tidyr 1.3.0 v stringr 1.5.0
v readr 2.1.2 v forcats 0.5.1
-- Conflicts ----- tidyverse_conflicts() --
x dplyr::count() masks matrixStats::count()
x dplyr::filter() masks stats::filter()
x dplyr::lag()
                 masks stats::lag()
library("pROC")
Type 'citation("pROC")' for a citation.
Attaching package: 'pROC'
The following objects are masked from 'package:stats':
    cov, smooth, var
library("future")
library("future.apply")
library("reshape")
Attaching package: 'reshape'
The following object is masked from 'package:dplyr':
    rename
The following objects are masked from 'package:tidyr':
    expand, smiths
library("idr2d")
Set color palette.
res_colors <-
  c(IDR="tan",
    gIDR="#30B7BC", # bright teal
eCV="#AF275F", # light magenta
mIDR="#DE653A") #bright orange
```

Simulation study.

Run simulations. The three simulation scenarios were created with function eCV::simulate_data, following Table 1 from Li et al. (2011).

Create a table with all combinations of simulation parameters.

Set parameters for each model.

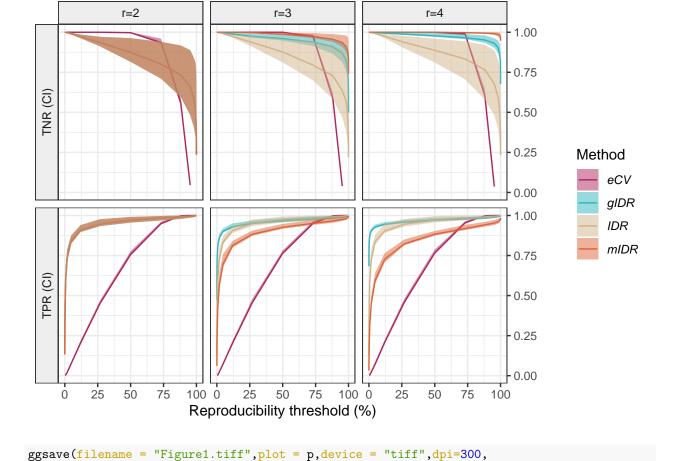
```
# Set parameters for each model.
methods_params <- list(</pre>
 eCV = list(max.ite = 1e4),
 IDR = list(
    mu = 2,
    sigma = 1,
    rho = 0.85,
    p = 0.5,
   eps = 1e-3,
   max.ite = 50
 ),
 mIDR = list(
    mu = 2,
    sigma = 1,
    rho = 0.85,
    p = 0.5
   eps = 1e-3,
   max.ite = 50
  ),
 gIDR = list(
    mu = 2,
    sigma = 1,
    rho = 0.85,
    p = 0.5,
    eps = 1e-3,
    max.ite = 50
  )
)
```

```
if(method != "eCV") {
        X <- preprocess(sim_data$sim_data,</pre>
                           value_transformation = "identity",
                           jitter_factor=1e-4)
      } else X <- sim_data$sim_data</pre>
     rep_index <- mrep_assessment(</pre>
          x = X,
          method = method,
          param = methods_params[[method]],
          n_{threads} = 1
        )$rep_index
    perf <- roc(sim_data$sim_params$feature_group == 2, rep_index, quiet = TRUE)</pre>
    perf_thr <- coords(perf,</pre>
                            x = 1/(1 + \exp(-c(1:20) + 10)),
                            ret=c("threshold","tpr","tnr"))
    perf_thr <- rbind(perf_thr$tpr %>%
    as.data.frame() %>%
    mutate(threshold=1/(1 + exp(-c(1:20) + 10)),
           perf="TPR (CI)") ,
        perf_thr$tnr %>%
    as.data.frame() %>%
    mutate(threshold=1/(1 + exp(-c(1:20) + 10)),
           perf="TNR (CI)"))
    perf_thr$n_rep <- n_rep</pre>
    perf_thr$method <- method</pre>
    perf_thr$scenario <- scenario</pre>
    perf_thr$sim <- sim</pre>
    return(perf_thr)
  ,n_{rep} = n_{rep}
  scenario = scenario,
  method=method,
  future.seed = NULL) %>% do.call(what = rbind)
  perf_res <- rbind(perf_res, tmp)</pre>
future::plan(sequential)
# Save results.
colnames(perf res)[1] <- "value"</pre>
saveRDS(perf_res,file="perf_resSim.rds")
```

Arrange results for figure creation.

Scenario 1.

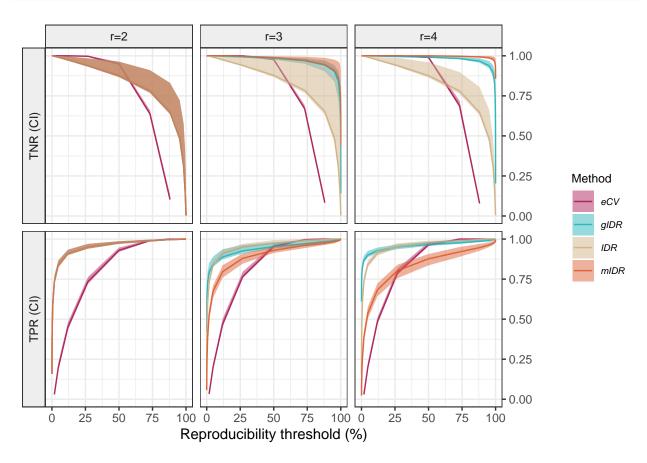
```
p <- perf_plot_df %>%
  filter(scenario==1) %>%
  ggplot(aes(x=threshold,y=`50%`,color=method)) +
  facet_grid(perf ~ n_rep,switch ="y") +
  geom_line() +
  geom_ribbon(color = NA,
    aes(x = threshold, ymin = 5\%, ymax = 95%, fill=method),
     alpha = 0.5) +
  scale_x_continuous(breaks=c(0,0.25,0.5,0.75,1),
                    labels = c("0","25","50","75","100"))+
  scale color manual(values=alpha(res colors,1)) +
  scale_fill_manual(values=alpha(res_colors,0.6)) +
    scale_y_continuous(position = "right")+
   theme_bw() + theme(
                     legend.text = element_text(face = "italic"),
                     strip.background = element_rect(fill=alpha("gray", 0.25)),
                  legend.background = element_rect(fill = alpha("white",0.1))) +
   guides(color=guide_legend(ncol=1, override.aes = list(size = 2))) +
   labs(y="",
       fill="Method",
       color="Method",
       x="Reproducibility threshold (%)")
p
```



units = "in", width = 7, height = 5, scale = 0.85)

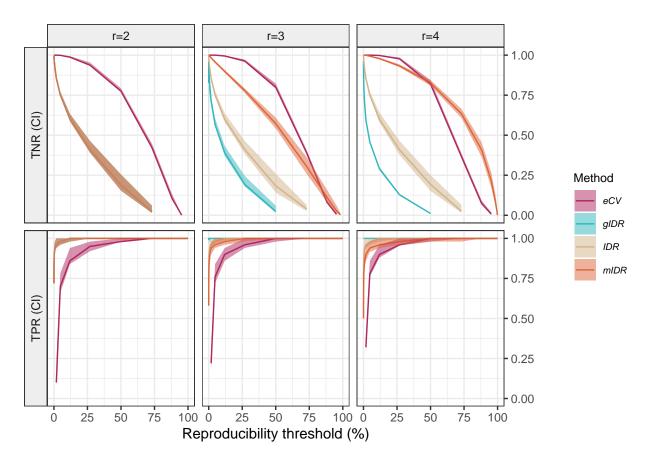
Scenario 2.

```
p <- perf_plot_df %>%
  filter(scenario==2) %>%
  ggplot(aes(x=threshold,y=`50%`,color=method)) +
  facet_grid(perf ~ n_rep,switch ="y") +
  geom_line() +
   geom_ribbon(color = NA,
    aes(x = threshold, ymin = 5\%, ymax = 95%, fill=method),
     alpha = 0.5) +
  scale_color_manual(values=alpha(res_colors,1)) +
    scale x continuous(breaks=c(0,0.25,0.5,0.75,1),
                     labels = c("0","25","50","75","100"))+
  scale_fill_manual(values=alpha(res_colors,0.6)) +
    scale_y_continuous(position = "right")+
   theme_bw() + theme(
                     legend.title = element_text(size=9),
                     legend.text = element_text(size=7,face = "italic"),
                     strip.background = element_rect(fill=alpha("gray",0.25)),
                  legend.background = element_rect(fill = alpha("white",0.1))) +
```



```
ggsave(filename = "Figure2.tiff",plot = p,device = "tiff",dpi=300,units = "in",
    width = 7,height = 5,scale = 0.85)
```

Scenario 3.



```
ggsave(filename = "Figure3.tiff",plot = p,device = "tiff",dpi=300,
    units = "in",width = 7,height = 5,scale = 0.85)
```