

The battle of total-order sensitivity estimators

R code

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```

# PRELIMINARY FUNCTIONS -----
# Function to read in all required packages in one go:
loadPackages <- function(x) {
  for(i in x) {
    if(!require(i, character.only = TRUE)) {
      install.packages(i, dependencies = TRUE)
      library(i, character.only = TRUE)
    }
  }
}

# Install development version of sensobol
remotes::install_github("arnaldpuy/sensobol")

# Load the packages
loadPackages(c("Rcpp", "RcppArmadillo", "tidyverse", "parallel", "foreach",
              "doParallel", "Rfast", "data.table", "scales", "cowplot",
              "benchmarkme", "logitnorm", "sensobol", "ggrepel"))

# Create custom theme
theme_AP <- function() {
  theme_bw() +
  theme(panel.grid.major = element_blank(),
        panel.grid.minor = element_blank(),
        legend.background = element_rect(fill = "transparent",
                                           color = NA),
        legend.key = element_rect(fill = "transparent",
                                  color = NA))
}

# Set checkpoint

dir.create(".checkpoint")
library("checkpoint")

checkpoint("2020-11-05",
           R.version ="4.0.3",
           checkpointLocation = getwd())

```

1 Functions

1.1 Savage scores

```
# SAVAGE SCORES -----  
  
savage_scores <- function(x) {  
  true.ranks <- rank(-x)  
  p <- sort(1 / true.ranks)  
  mat <- matrix(rep(p, length(p)), nrow = length(p), byrow = TRUE)  
  mat[upper.tri(mat)] <- 0  
  out <- sort(rowSums(mat), decreasing = TRUE)[true.ranks]  
  return(out)  
}
```

1.2 Sensitivity indices

```
# VARS FUNCTIONS -----  
  
vars_matrices <- function(star.centers, params, h, method = "QRN") {  
  out <- center <- sections <- A <- B <- AB <- X <- out <- list()  
  if(method == "QRN") {  
    mat <- randtoolbox::sobol(n = star.centers, dim = length(params))  
  } else if(method == "R") {  
    mat <- replicate(length(params), stats::runif(star.centers))  
  } else if(method == "LHS") {  
    mat <- lhs::randomLHS(star.centers, length(params))  
  } else {  
    stop("method should be either QRN, R or LHS")  
  }  
  for(i in 1:nrow(mat)) {  
    center[[i]] <- mat[i, ]  
    sections[[i]] <- sapply(center[[i]], function(x) {  
      all <- seq(x %% h, 1, h)  
      non.zeros <- all[all != 0] # Remove zeroes  
    })  
    B[[i]] <- sapply(1:ncol(mat), function(x)  
      sections[[i]][, x][!sections[[i]][, x] %in% center[[i]][x]])  
    A[[i]] <- matrix(center[[i]], nrow = nrow(B[[i]]),  
      ncol = length(center[[i]]), byrow = TRUE)  
    X[[i]] <- rbind(A[[i]], B[[i]])  
    for(j in 1:ncol(A[[i]])) {  
      AB[[i]] <- A[[i]]  
      AB[[i]][, j] <- B[[i]][, j]  
      X[[i]] <- rbind(X[[i]], AB[[i]])  
    }  
    AB[[i]] <- X[[i]][(2 * nrow(B[[i]])) + 1]:nrow(X[[i]]), ]  
    out[[i]] <- rbind(unname(center[[i]]), AB[[i]])  
  }
```

```

    }
    return(do.call(rbind, out))
}

# Function to cut by size
CutBySize <- function(m, block.size, nb = ceiling(m / block.size)) {
  int <- m / nb
  upper <- round(1:nb * int)
  lower <- c(1, upper[-nb] + 1)
  size <- c(upper[1], diff(upper))
  cbind(lower, upper, size)
}

# VARS-TO algorithm
vars_ti <- function(Y, star.centers, params, h, method = "one.step") {
  n.cross.points <- length(params) * ((1 / h) - 1) + 1
  index.centers <- seq(1, length(Y), n.cross.points)
  mat <- matrix(Y[-index.centers], ncol = star.centers)
  indices <- CutBySize(nrow(mat), nb = length(params))
  out <- list()
  for(i in 1:nrow(indices)) {
    out[[i]] <- mat[indices[i], "lower":indices[i, "upper"], ]
  }
  if(method == "one.step") {
    d <- lapply(1:length(params), function(x)
      lapply(1:ncol(out[[x]]), function(j) {
        da <- c(out[[x]][, j][1],
          rep(out[[x]][, j][-c(1, length(out[[x]][, j]))], each = 2),
          out[[x]][, j][length(out[[x]][, j])])
      }))
  } else if(method == "all.step") {
    d <- lapply(1:length(params), function(x)
      lapply(1:ncol(out[[x]]), function(j) {
        da <- c(combn(out[[x]][, j], 2)
      }))
  } else {
    stop("method should be either one.step or all.step")
  }
  out <- lapply(d, function(x)
    lapply(x, function(y) matrix(y, nrow = length(y) / 2, byrow = TRUE)))
  variogr <- unlist(lapply(out, function(x) lapply(x, function(y)
    mean(0.5 * (y[, 1] - y[, 2]) ^ 2))) %>%
    lapply(., function(x) do.call(rbind, x)) %>%
    lapply(., mean))
  covariogr <- unlist(lapply(out, function(x)
    lapply(x, function(y) cov(y[, 1], y[, 2]))) %>%
    lapply(., function(x) Rfast::colmeans(do.call(rbind, x))))
}

```

```

VY <- var(Y[index.centers])
Ti <- (variogr + covariogr) / VY
output <- data.table::data.table(Ti)
output[, `:=` (parameters = params)]
return(output)
}

# COMPUTATION OF SOBOL' Ti INDICES ----

sobel_Ti <- function(d, N, params, total) {
  m <- matrix(d, nrow = N)
  k <- length(params)
  if(total == "jansen" | total == "homma" | total == "monod" | total == "glen") {
    Y_A <- m[, 1]
    Y_AB <- m[, -1]
    f0 <- (1 / length(Y_A)) * sum(Y_A)
    VY <- 1 / length(Y_A) * sum((Y_A - f0) ^ 2)
    # VY <- 1 / length(Y_A) * (sum(Y_A ^ 2) -
    # (1 / N * sum(Y_A ^ 2))) ((Variance used by Becker))
  }
  if(total == "jansen") {
    Ti <- (1 / (2 * N) * Rfast:::colsums((Y_A - Y_AB) ^ 2)) / VY
  } else if(total == "homma") {
    Ti <- (VY - (1 / N) * Rfast:::colsums(Y_A * Y_AB) + f0 ^ 2) / VY
  } else if(total == "monod") {
    Ti <- 1 - (1 / N * Rfast:::colsums(Y_A * Y_AB) -
      (1 / N * Rfast:::colsums((Y_A + Y_AB) / 2)) ^ 2) /
      (1 / N * Rfast:::colsums((Y_A ^ 2 + Y_AB ^ 2) / 2) -
      (1 / N * Rfast:::colsums((Y_A + Y_AB) / 2)) ^ 2)
  } else if(total == "glen") {
    Ti <- 1 - (1 / (N - 1) *
      Rfast:::colsums(((Y_A - mean(Y_A)) * (Y_AB - Rfast:::colmeans(Y_AB))) /
      sqrt(var(Y_A) * Rfast:::colVars(Y_AB))))
  }
  if(total == "saltelli") {
    Y_A <- m[, 1]
    Y_B <- m[, 2]
    Y_BA <- m[, -c(1, 2)]
    f02 <- (1 / N * sum(Y_A)) ^ 2
    VY <- 1 / N * sum(Y_A ^ 2) - f02
    Ti <- 1 - ((1 / N * Rfast:::colsums(Y_B * Y_BA - f02)) / VY)
  }
  if(total == "azzini" | total == "lamboni") {
    Y_A <- m[, 1]
    Y_B <- m[, 2]
    Y_AB <- m[, 3:(k - 1)]
    Y_BA <- m[, (ncol(m) - k + 1):ncol(m)]
  }
}

```

```

f0 <- 1 / (2 * N) * sum(Y_A + Y_B)
VY <- 1 / (2 * N - 1) * sum((Y_A - f0) ^ 2 + (Y_B - f0) ^ 2)
}
if(total == "azzini") {
  Ti <- Rfast:::colsums((Y_B - Y_BA) ^ 2 + (Y_A - Y_AB) ^ 2) /
    Rfast:::colsums((Y_A - Y_B) ^ 2 + (Y_BA - Y_AB) ^ 2)
} else if(total == "lamboni") {
  Ti <- (1 / (4 * N) * colSums((Y_A - Y_AB) ^ 2 + (Y_B - Y_BA) ^ 2, na.rm = TRUE)) / VY
}
if(total == "owen") {
  Y_A <- m[, 1]
  Y_B <- m[, 2]
  Y_BA <- m[, 3:(3 + k - 1)]
  Y_CB <- m[, (ncol(m) - k + 1):ncol(m)]
  VY <- sapply(1:k, function(j)
    mean(Rfast:::rowmeans(m[,c(1, 2, 2 + j, 2 + j + k)] ^ 2)) -
    mean(Rfast:::rowmeans(m[,c(1, 2, 2 + j, 2 + j + k)]))) ^ 2)
  Ti <- (VY - (1 / N * Rfast:::colsums((Y_B - Y_CB) * (Y_BA - Y_A)))) / VY
}
output <- data.table(Ti)
output[, `:=` (parameters = paste("X", 1:k, sep = ""))]
return(output)
}

# CHECK THAT ALL TI ESTIMATORS WORK -----
# Settings
estimators <- c("jansen", "homma", "azzini", "monod", "glen", "owen", "saltelli")
test_functions <- c("Ishigami", "Sobol'G", "Morris")
N <- 2 ^ 11

# Run model
ind <- Y <- mt <- list()
for(i in estimators) {
  for(j in test_functions) {
    if(i == "jansen" | i == "homma" | i == "monod" | i == "glen") {
      matrices <- c("A", "AB")
    } else if(i == "azzini"){
      matrices <- c("A", "B", "AB", "BA")
    } else if(i == "owen") {
      matrices <- c("A", "B", "BA", "CB")
    } else if(i == "saltelli") {
      matrices <- c("A", "B", "BA")
    }
    if(j == "Ishigami") {
      k <- 3
      modelRun <- sensobol::ishigami_Fun
    }
  }
}

```

```

} else if(j == "Sobol'G") {
  k <- 8
  modelRun <- sensobol::sobol_Fun
} else if(j == "Morris") {
  k <- 20
  modelRun <- sensitivity::morris.fun
}
mt[[i]][[j]] <- sobol_matrices(N = N, params = paste("X", 1:k, sep = ""), matrices = matrix)
Y[[i]][[j]] <- modelRun(mt[[i]][[j]])
ind[[i]][[j]] <- sobol_Ti(d = Y[[i]][[j]], params = paste("X", 1:k, sep = ""),
                           N = N, total = i)
}
}

## Registered S3 method overwritten by 'sensitivity':
##   method      from
##   print.srct  dplyr

# Run model for VARS
star.centers <- 200
h <- 0.2
vars.ind <- Y <- list()
for(j in test_functions) {
  if(j == "Ishigami") {
    k <- 3
    modelRun <- sensobol::ishigami_Fun
  } else if(j == "Sobol'G") {
    k <- 8
    modelRun <- sensobol::sobol_Fun
  } else if(j == "Morris") {
    k <- 20
    modelRun <- sensitivity::morris.fun
  }
  mt[[j]] <- vars_matrices(star.centers = star.centers,
                            params = paste("X", 1:k, sep = ""),
                            h = h)
  Y[[j]] <- modelRun(mt[[j]])
  vars.ind[[j]] <- vars_ti(Y = Y[[j]], star.centers = star.centers,
                            params = paste("X", 1:k, sep = ""), h = h)
}

vars.ind <- rbindlist(vars.ind, idcol = "Function")[
  , estimator := "vars"] %>%
  setcolororder(., c("estimator", "Function", "Ti", "parameters"))

# PLOT SENSITIVITY INDICES -----

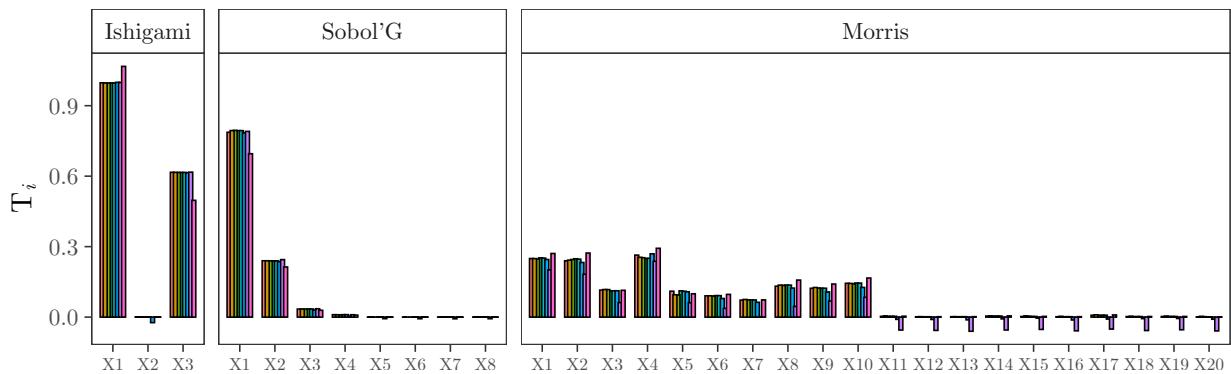
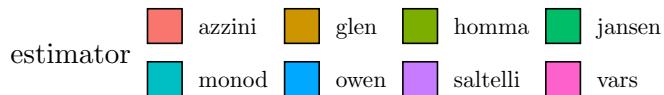
lapply(ind, function(x) rbindlist(x, idcol = "Function")) %>%

```

```

rbindlist(., idcol = "estimator") %>%
rbind(vars.ind) %>%
.[, parameters:= factor(parameters, levels = paste("X", 1:20, sep = ""))] %>%
.[, Function:= factor(Function, levels = test_functions)] %>%
ggplot(., aes(parameters, Ti, fill = estimator)) +
geom_bar(stat = "identity",
         position = position_dodge(0.7),
         color = "black") +
facet_grid(~Function,
           scales = "free_x",
           space = "free_x") +
labs(x = "",
     y = expression(T[italic(i)])) +
theme_AP() +
theme(axis.text.x = element_text(size = 6.5),
      legend.position = "top",
      strip.background = element_rect(fill = "white")) +
guides(fill = guide_legend(nrow = 2,
                           byrow = TRUE))

```



PLOT THE SAMPLING METHODS AVAILABLE -----

```

method <- c("R", "QRN")
matrices <- "A"
N <- 200
set.seed(666) # For the random sampling
params <- paste("X", 1:2, sep = "")
A <- lapply(method, function(x)
  sobol_matrices(N = N, params = params,
                  matrices = matrices, method = x))

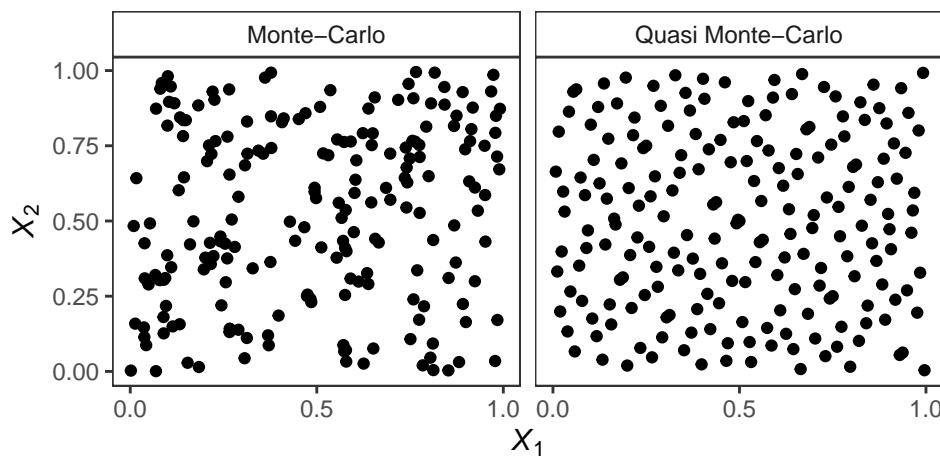
names(A) <- method
lapply(A, data.table) %>%

```

```

rbindlist(., idcol = "method") %>%
.[, method:= ifelse(method == "R", "Monte-Carlo", "Quasi Monte-Carlo")] %>%
.[, method:= factor(method, levels = c("Monte-Carlo", "Quasi Monte-Carlo"))] %>%
ggplot(., aes(X1, X2)) +
geom_point() +
scale_x_continuous(breaks = pretty_breaks(n = 3)) +
labs(x = expression(italic(X)[1]),
     y = expression(italic(X)[2])) +
facet_wrap(~method, ncol = 3) +
theme_AP() +
theme(strip.background = element_rect(fill = "white"))

```



2 The metafunction

```

# CREATE METAFUNCTION -----
function_list <- list(
  Linear = function(x) x,
  Quadratic = function(x) x ^ 2,
  Cubic = function(x) x ^ 3,
  Exponential = function(x) exp(1) ^ x / (exp(1) - 1),
  Periodic = function(x) sin(2 * pi * x) / 2,
  Discontinuous = function(x) ifelse(x > 0.5, 1, 0),
  Non.monotonic = function(x) 4 * (x - 0.5) ^ 2,
  Inverse = function(x) (10 - 1 / 1.1) ^ -1 * (x + 0.1) ^ -1,
  No.effect = function(x) x * 0,
  Trigonometric = function(x) cos(x)
)

# PLOT METAFUNCTION -----
a <- ggplot(data.frame(x = runif(100)), aes(x)) +
  map(1:length(function_list), function(nn) {

```

```

stat_function(fun = function_list[[nn]],
              geom = "line",
              aes_(color = factor(names(function_list[nn]))))

}) +
  labs(color= "Function", linetype = "Function",
       x = expression(italic(x)),
       y = expression(italic(y))) +
  scale_color_discrete(labels = c(expression(f[1](x) == x ^ 3),
                                   expression(paste(f[2](x) == 1, " if ", x > 1/2, ", otherwise "),
                                   expression(f[3](x) == (e ^ x - 1) / (e - 1)),
                                   expression(f[4](x) == (10 - 1 / 1.1) ^ -1 * (x + 0.1) ^ -1),
                                   expression(f[5](x) == x),
                                   expression(f[6](x) == 0),
                                   expression(f[7](x) == 4^(x - 0.5)^2),
                                   expression(f[8](x) == sin(2pi x) / 2),
                                   expression(f[9](x) == x ^ 2),
                                   expression(f[10](x) == cos(x)))) +
  theme_AP() +
  theme(legend.text.align = 0)

```

a

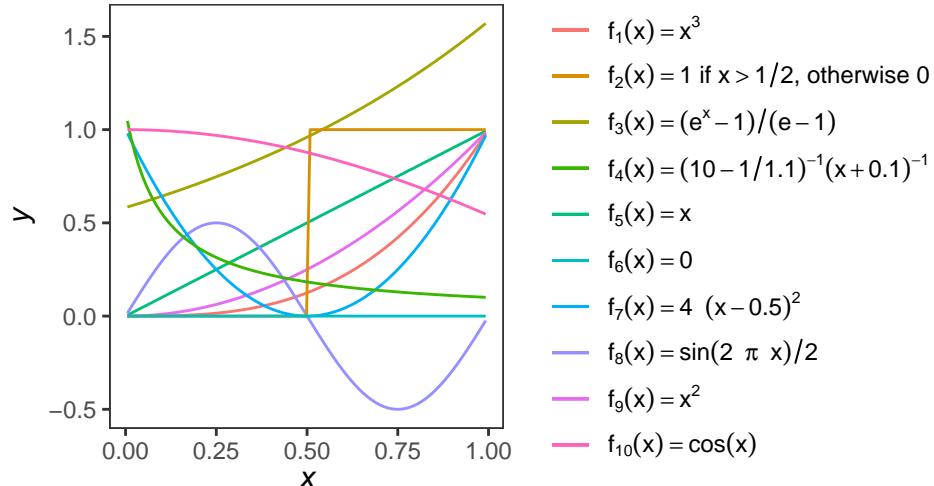


Figure 1: Functions used in the metafunction of Becker (2020).

CREATE FUNCTION FOR RANDOM DISTRIBUTIONS -----

```

sample_distributions <- list(
  "uniform" = function(x) x,
  "normal" = function(x) qnorm(x, 0.5, 0.2),
  "beta" = function(x) qbeta(x, 8, 2),
  "beta2" = function(x) qbeta(x, 2, 8),
  "beta3" = function(x) qbeta(x, 2, 0.5),
  "beta4" = function(x) qbeta(x, 0.5, 2),

```

```

"logitnormal" = function(x) qlogitnorm(x, 0, 3.16)
# Logit-normal, Bates too?
)

random_distributions <- function(X, phi) {
  names_ff <- names(sample_distributions)
  if(!phi == length(names_ff) + 1) {
    out <- sample_distributions[[names_ff[phi]]](X)
  } else {
    temp <- sample(names_ff, ncol(X), replace = TRUE)
    out <- sapply(seq_along(temp), function(x) sample_distributions[[temp[x]]](X[, x]))
  }
  return(out)
}

# PLOT DISTRIBUTIONS -----
names_ff <- names(sample_distributions)
prove <- randtoolbox::sobol(n = 1000, dim = length(names_ff))

out <- data.table(sapply(seq_along(names_ff), function(x)
  sample_distributions[[names_ff[x]]])(prove[, x]))

b <- data.table::melt(out) %>%
  ggplot(., aes(value, group = variable, colour = variable)) +
  geom_density() +
  scale_color_discrete(labels = c("U(0, 1)",
                                    "N(0.5, 0.2)",
                                    "Beta(8, 2)",
                                    "Beta(2, 8)",
                                    "Beta(2, 0.5)",
                                    "Beta(0.5, 2)",
                                    "Logitnormal(0, 3.16)"),
                        name = "Distribution") +
  labs(x = expression(italic(x)),
       y = "Density") +
  theme_AP() +
  theme(legend.text.align = 0)

## Warning in melt.data.table(out): id.vars and measure.vars are internally
## guessed when both are 'NULL'. All non-numeric/integer/logical type columns are
## considered id.vars, which in this case are columns []. Consider providing at
## least one of 'id' or 'measure' vars in future.

b

# MERGE METAFUNCTION PLOT AND DISTRIBUTIONS PLOT -----
plot_grid(b, a, ncol = 1, labels = "auto", align = "hv")

```

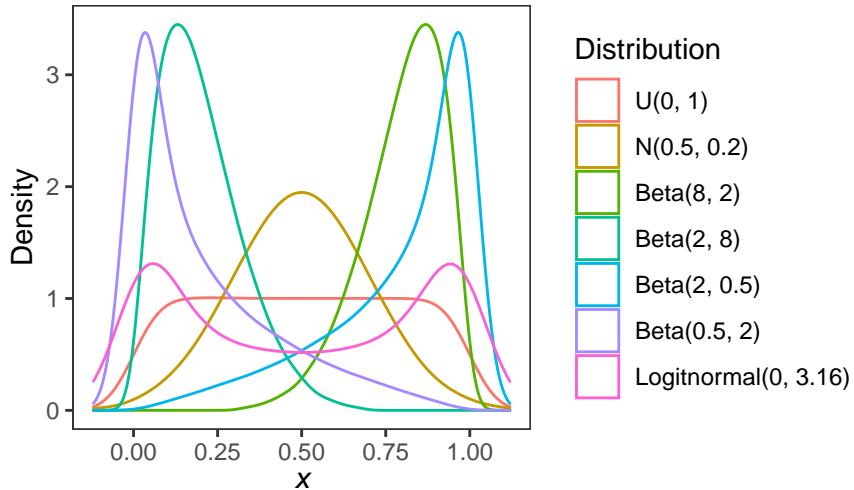
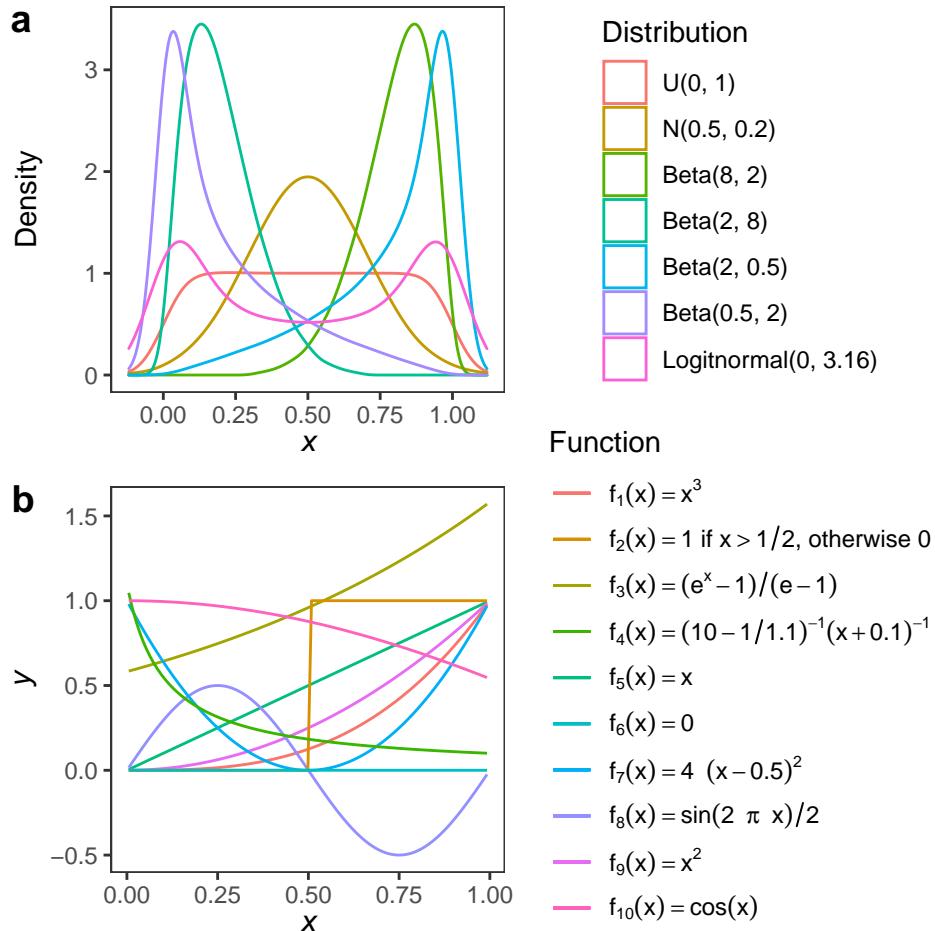


Figure 2: Distributions used in the metafunction of Becker (2020).



EXEMPLIFY THE METAFUNCTION WITH AN EXAMPLE -----

```
N <- 10000 # Sample size
R <- 100 # Number of bootstrap replications
k <- 17 # Number of model inputs
```

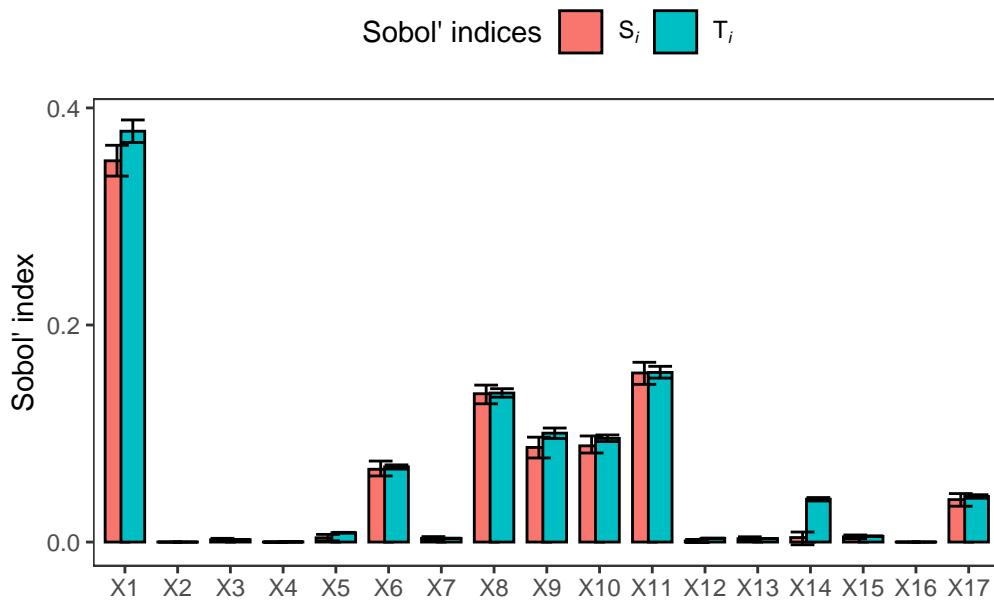
```

k_2 <- 0.5 # Fraction of active pairwise interactions
k_3 <- 0.2 # Fraction of active three-wise interactions
epsilon <- 666 # to reproduce the results
params <- paste("X", 1:k, sep = "")
A <- sobol_matrices(N = N, params = params)

# Compute
Y <- sensobol::metafunction(data = A, k_2 = k_2, k_3 = k_3, epsilon = epsilon)
indices <- data.table(sensobol::sobol_indices(Y = Y, N = N, params = params, boot = TRUE, R = 1000))
.[, parameters:= factor(parameters, levels = paste("X", 1:k, sep = ""))]

ggplot(indices, aes(parameters, original, fill = sensitivity)) +
  geom_bar(stat = "identity",
            position = position_dodge(0.6),
            color = "black") +
  geom_errorbar(aes(ymin = low.ci,
                     ymax = high.ci),
                position = position_dodge(0.6)) +
  scale_x_discrete(labels = ggplot2:::parse_safe) +
  scale_y_continuous(breaks = pretty_breaks(n = 3)) +
  labs(x = "",
       y = "Sobol' index") +
  scale_fill_discrete(name = "Sobol' indices",
                      labels = c(expression(S[italic(i)]),
                                expression(T[italic(i)]))) +
  theme_AP() +
  theme(legend.position = "top")

```



```
# FUNCTIONS TO COMPUTE SUM OF SI AND PROPORTION OF SI > 0.05 FOR METAFUNCTION -----
```

```

# Function to replicate in parallel
RepParallel <- function(n, expr, simplify = "array", ...) {
  answer <-
    mclapply(integer(n), eval.parent(substitute(function(...) expr)), ...)
  if (!identical(simplify, FALSE) && length(answer))
    return(simplify2array(answer, higher = (simplify == "array")))
  else return(answer)
}

# Function to replicate
try_metafunction <- function(x) {
  k <- sample(x)[1]
  params <- paste("X", 1:k, sep = "")
  N <- 500
  A <- sobol_matrices(N = N, params = params)
  Y <- sensobol::metafunction(A)
  ind <- sobol_indices(Y = Y, N = N, params = params)
  out1 <- ind[sensitivity == "Si", sum(original)]
  out2 <- ind[sensitivity == "Ti", sum(original > 0.05)/.N]
  output <- c(out1, out2)
  return(output)
}

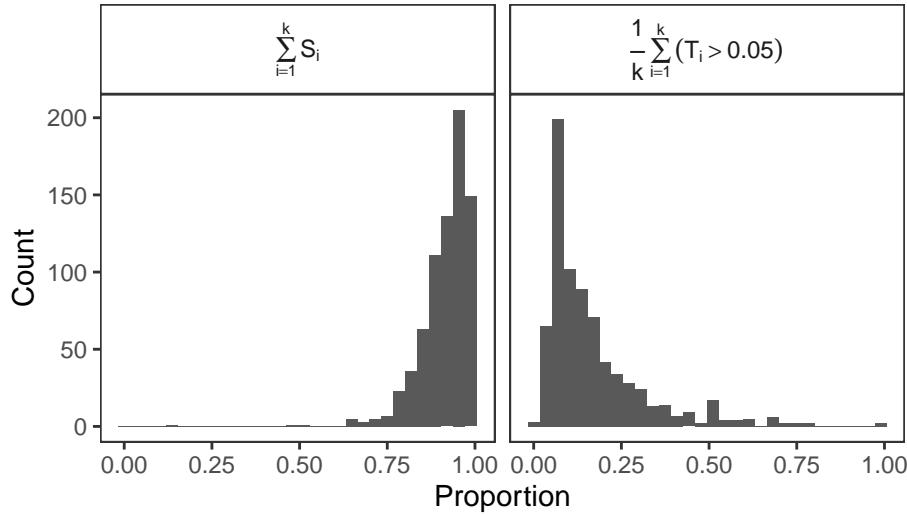
# COMPUTE SUM OF SI AND PROPORTION OF SI > 0.05 FOR METAFUNCTION -----
n <- 10^3
out <- do.call(rbind, RepParallel(n, try_metafunction(3:100), simplify = FALSE))
dt.out <- data.table(out)

# PLOT

dt.out[V1 < 1 & V1 > 0] %>%
  setnames(., c("V1", "V2"),
           c("sum(S[i], i==1, k)",
             "frac(1,k)-sum((T[i]>0.05), i == 1, k))) %>%
  melt(., measure.vars = c("sum(S[i], i==1, k)",
                           "frac(1,k)-sum((T[i]>0.05), i == 1, k))) %>%
  ggplot(., aes(value)) +
  geom_histogram() +
  facet_wrap(~variable,
             labeller = label_parsed) +
  labs(x = "Proportion", y = "Count") +
  theme_AP() +
  theme(strip.background = element_rect(fill = "white"))

## `stat_bin()` using `bins = 30` . Pick better value with `binwidth` .

```



3 The model

3.1 Settings

```
# DEFINE SETTINGS -----
N <- 2 ^ 11 # Sample size of sample matrix
h <- 0.2 # step for VARS
R <- 500 # Number of bootstrap replicas
n_cores <- ceiling(detectCores() * 0.5)
order <- "second"
params <- c("k_2", "k_3", "epsilon", "phi", "delta", "tau")
N.high <- 2 ^ 11 # Maximum sample size of the large sample matrix
```

3.2 Sample matrix

```
# CREATE SAMPLE MATRIX -----
# Main matrix without Nt and k
tmp <- sobol_matrices(N = N, params = c(params, "N_t", "k"), order = order, matrices = c("A",
A <- tmp[1:N, ]
B <- tmp[(N + 1):(2 * N), ]

first <- 1:length(params)
loop <- c(first, utils::combn(1:length(params), 2, simplify = FALSE))

X <- rbind(A, B)
for(i in loop) {
  AB <- A
  AB[, i] <- B[, i]
  X <- rbind(X, AB)
}
```

```

mat <- X

# Matrix with clusters (Nt,k)
AB <- AB2 <- AB3 <- A
AB[, c("N_t", "k")] <- B[, c("N_t", "k")] # (Nt,k)
AB2[, c("k_2", "k_3", "epsilon", "phi")] <- B[, c("k_2", "k_3", "epsilon", "phi")] # Model un
AB3[, c("delta", "tau")] <- B[, c("delta", "tau")] # Manageable uncertainties

# Final matrix
mat <- rbind(mat, AB, AB2, AB3)

# TRANSFORM MATRIX -----
mat[, 1] <- round(qunif(mat[, 1], 0.3, 0.5), 2) # k_2
mat[, 2] <- round(qunif(mat[, 2], 0.1, 0.3), 2) # k_3
mat[, 3] <- floor(qunif(mat[, 3], 1, 200)) # Epsilon
mat[, 4] <- floor(mat[, 4] * (8 - 1 + 1)) + 1 # Phi
mat[, 5] <- floor(mat[, 5] * (3 - 1 + 1)) + 1 # Delta
mat[, 6] <- floor(mat[, 6] * (2 - 1 + 1)) + 1 # Tau
mat[, 7] <- floor(qunif(mat[, 7], 10, 1000)) # Nt
mat[, 8] <- floor(qunif(mat[, 8], 3, 100)) # k

# Define the base sample matrix
N.all <- apply(mat, 1, function(x) ceiling(x["N_t"] / (x["k"] + 1)))
N.saltelli <- apply(mat, 1, function(x) ceiling(x["N_t"] / (x["k"] + 2)))
N.azzini <- apply(mat, 1, function(x) ceiling(x["N_t"] / (2 * x["k"] + 2)))
N.vars <- apply(mat, 1, function(x) floor(x["N_t"] / ((x["k"]) * ((1 / h) - 1) + 1)))

tmp <- data.table(cbind(mat, N.all, N.azzini, N.vars))

# Now constrain N as a function of vars
tmp <- tmp[, N.all:= ifelse(N.vars == 1 | N.vars == 0, ceiling(2 * (4 * k + 1) / (k + 1)), N.a
tmp <- tmp[, N.saltelli:= ifelse(N.vars == 1 | N.vars == 0, ceiling(2 * (4 * k + 1) / (k + 2))
tmp <- tmp[, N.azzini:= ifelse(N.vars == 1 | N.vars == 0, ceiling(2 * (4 * k + 1) / (2 * k + 2
tmp <- tmp[, N.vars:= ifelse(N.vars == 1 | N.vars == 0, 2, N.vars)]

C.all <- apply(tmp, 1, function(x) x["N.all"] * (x["k"] + 1))
C.saltelli <- apply(tmp, 1, function(x) x["N.saltelli"] * (x["k"] + 2))
C.azzini <- apply(tmp, 1, function(x) x["N.azzini"] * (2 * x["k"] + 2))
C.vars <- apply(tmp, 1, function(x) x["N.vars"] * (x["k"] * ((1 / h) - 1) + 1))

mat <- cbind(tmp, C.all, C.saltelli, C.azzini, C.vars)

# EXPORT SAMPLE MATRIX -----
fwrite(mat, "mat.csv")

```

3.3 Define the model

```
# DEFINE MODEL -----  
  
model_Ti <- function(k, N.all, N.azzini, N.saltelli, N.vars, h,  
                      N.high, k_2, k_3, epsilon, phi, delta, tau) {  
  ind <- list()  
  estimators <- c("jansen", "homma", "monod", "azzini", "glen", "owen", "vars", "saltelli")  
  if(tau == 1) {  
    method <- "R"  
  } else if(tau == 2) {  
    method <- "QRN"  
  }  
  set.seed(epsilon)  
  all.but.azzini <- sobol_matrices(N = N.all, params = paste("X", 1:k, sep = ""),  
                                    matrices = c("A", "AB"), method = method)  
  set.seed(epsilon)  
  saltelli <- sobol_matrices(N = N.saltelli, params = paste("X", 1:k, sep = ""),  
                            matrices = c("A", "B", "BA"), method = method)  
  set.seed(epsilon)  
  azzini <- sobol_matrices(N = N.azzini, params = paste("X", 1:k, sep = ""),  
                           matrices = c("A", "B", "AB", "BA"), method = method)  
  set.seed(epsilon)  
  owen.matrix <- sobol_matrices(N = N.azzini, params = paste("X", 1:k, sep = ""),  
                                matrices = c("A", "B", "BA", "CB"), method = method)  
  set.seed(epsilon)  
  vars.matrix <- vars_matrices(star.centers = N.vars, params = paste("X", 1:k, sep = ""),  
                               h = h, method = method)  
  set.seed(epsilon)  
  large.matrix <- sobol_matrices(N = N.high, params = paste("X", 1:k, sep = ""),  
                                 matrices = c("A", "AB"), method = method)  
  set.seed(epsilon)  
  all.matrices <- random_distributions(X = rbind(all.but.azzini,  
                                             azzini,  
                                             owen.matrix,  
                                             vars.matrix,  
                                             saltelli,  
                                             large.matrix),  
                                         phi = phi)  
  output <- sensobol::metafunction(data = all.matrices,  
                                    k_2 = k_2,  
                                    k_3 = k_3,  
                                    epsilon = epsilon)  
  full.ind <- sobol_Ti(d = tail(output, nrow(large.matrix)),  
                        N = N.high,  
                        params = paste("X", 1:k, sep = ""),  
                        total = "jansen")  
  full.ind[, sample.size:= "N"]
```

```

# Define indices of Y for estimators
Nt.all.but.azzini <- N.all * (k + 1)
Nt.azzini.owen <- N.azzini * ((2 * k) + 2)
Nt.vars <- N.vars * (k * ((1 / h) - 1) + 1)
Nt.saltelli <- N.saltelli * (k + 2)
lg.all.but.azzini <- 1:Nt.all.but.azzini
lg.azzini <- (length(lg.all.but.azzini) + 1):(length(lg.all.but.azzini) + Nt.azzini.owen)
lg.owen <- (max(lg.azzini) + 1):(max(lg.azzini) + Nt.azzini.owen)
lg.vars <- (max(lg.owen) + 1):(max(lg.owen) + Nt.vars)
lg.saltelli <- (max(lg.vars) + 1):(max(lg.vars) + Nt.saltelli)

for(i in estimators) {
  if(i == "jansen" | i == "homma" | i == "monod" | i == "glen") {
    y <- output[lg.all.but.azzini]
    n <- N.all
  } else if(i == "saltelli") {
    y <- output[lg.saltelli]
    n <- N.saltelli
  } else if(i == "azzini") {
    y <- output[lg.azzini]
    n <- N.azzini
  } else if(i == "owen") {
    y <- output[lg.owen]
    n <- N.azzini
  } else if(i == "vars") {
    y <- output[lg.vars]
    star.centers <- N.vars
  }
  if(!i == "vars") {
    ind[[i]] <- sobol_Ti(d = y, N = n, params = paste("X", 1:k, sep = ""), total = i)
  }
  if(i == "vars") {
    ind[[i]] <- vars_ti(Y = y, star.centers = star.centers,
                           params = paste("X", 1:k, sep = ""), h = h)
  }
  ind[[i]][, sample.size:= "n"]
  ind[[i]] <- rbind(ind[[i]], full.ind)
}
# Arrange data
out <- rbindlist(ind, idcol = "estimator")
out.wide <- dcast(out, estimator + parameters ~ sample.size, value.var = "Ti")
  # Replace NaN
for (i in seq_along(out.wide))
  set(out.wide, i=which(is.nan(out.wide[[i]])), j = i, value = 0)
  # Replace Inf
for (i in seq_along(out.wide))
  set(out.wide, i=which(is.infinite(out.wide[[i]])), j = i, value = 0)

```

```

# Replace Na
for (i in seq_along(out.wide))
  set(out.wide, i=which(is.na(out.wide[[i]])), j = i, value = 0)
# CHECK DELTA
if(delta == 1) { # Regular Pear
  final <- out.wide[, .(correlation = cor(N, n)), estimator]
} else if(delta == 2) { # kendall tau
  final <- out.wide[, .(correlation = pcaPP::cor.fk(N, n)), estimator]
} else { # Savage ranks
  final <- out.wide[, lapply(.SD, savage_scores), .SDcols = c("N", "n"), estimator][
    , .(correlation = cor(N, n)), estimator]
}
return(final)
}

```

3.4 Run the model

```

# RUN MODEL -----
# Define parallel computing
cl <- makeCluster(n_cores)
registerDoParallel(cl)

# Compute
Y.ti <- foreach(i=1:nrow(mat),
                 .packages = c("sensobol", "data.table", "pcaPP",
                               "logitnorm", "dplyr")) %dopar%
{
  model_Ti(k = mat[[i, "k"]],
            k_2 = mat[[i, "k_2"]],
            k_3 = mat[[i, "k_3"]],
            epsilon = mat[[i, "epsilon"]],
            phi = mat[[i, "phi"]],
            delta = mat[[i, "delta"]],
            tau = mat[[i, "tau"]],
            N.all = mat[[i, "N.all"]],
            N.azzini = mat[[i, "N.azzini"]],
            N.saltelli = mat[[i, "N.saltelli"]],
            N.vars = mat[[i, "N.vars"]],
            N.high = N.high,
            h = h)
}

# Stop parallel cluster
stopCluster(cl)

```

3.5 Arrange output

```
# ARRANGE OUTPUT ----

out_cor <- rbindlist(Y.ti, idcol = "row")

mt.dt <- data.table(mat) %>%
  .[, row:= 1:.N]

full_output <- merge(mt.dt, out_cor) %>%
  .[, Nt:= ifelse(estimator == "azzini" | estimator == "owen", N.azzini * (2 * k + 2),
                 ifelse(estimator == "vars", N.vars * (k * ((1 / h) - 1) + 1),
                       ifelse(estimator == "saltelli", N.saltelli * (k + 2), N.all * (k + 1)))
  .[, estimator:= ifelse(estimator %in% "azzini", "Azzini and Rosati",
                           ifelse(estimator %in% "homma", "Homma and Saltelli",
                                 ifelse(estimator %in% "monod", "Janon/Monod",
                                       ifelse(estimator %in% "jansen", "Jansen",
                                             ifelse(estimator %in% "glen", "Glen and Isaacs",
                                               ifelse(estimator %in% "owen", "pseudo-Owen",
                                                 ifelse(estimator %in% "saltelli", "Saltelli et al.", "Unknown"))))))))

# Define A matrix
A <- full_output[,.SD[1:N], estimator]

# Define matrix for total variance
VY.dt <- full_output[row %in% c(1:N)][
  , .(VY = 1 / N * sum((correlation - ((1 / N) * sum(correlation))) ^ 2)),
  estimator]

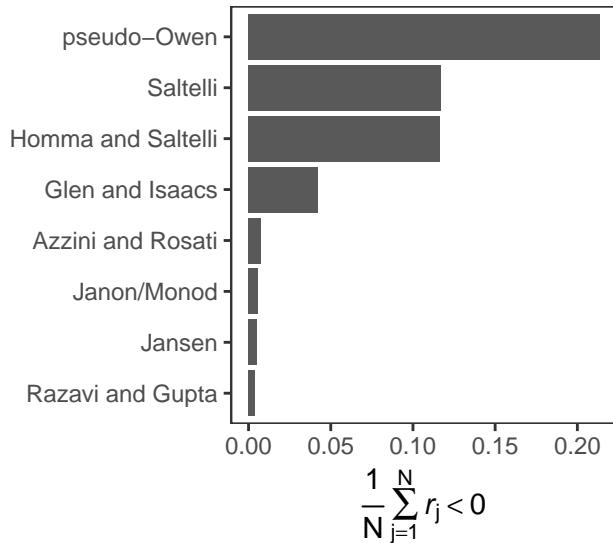
# EXPORT OUTPUT ----

fwrite(A, "A.csv")
fwrite(full_output, "full_output.csv")
```

4 Uncertainty analysis

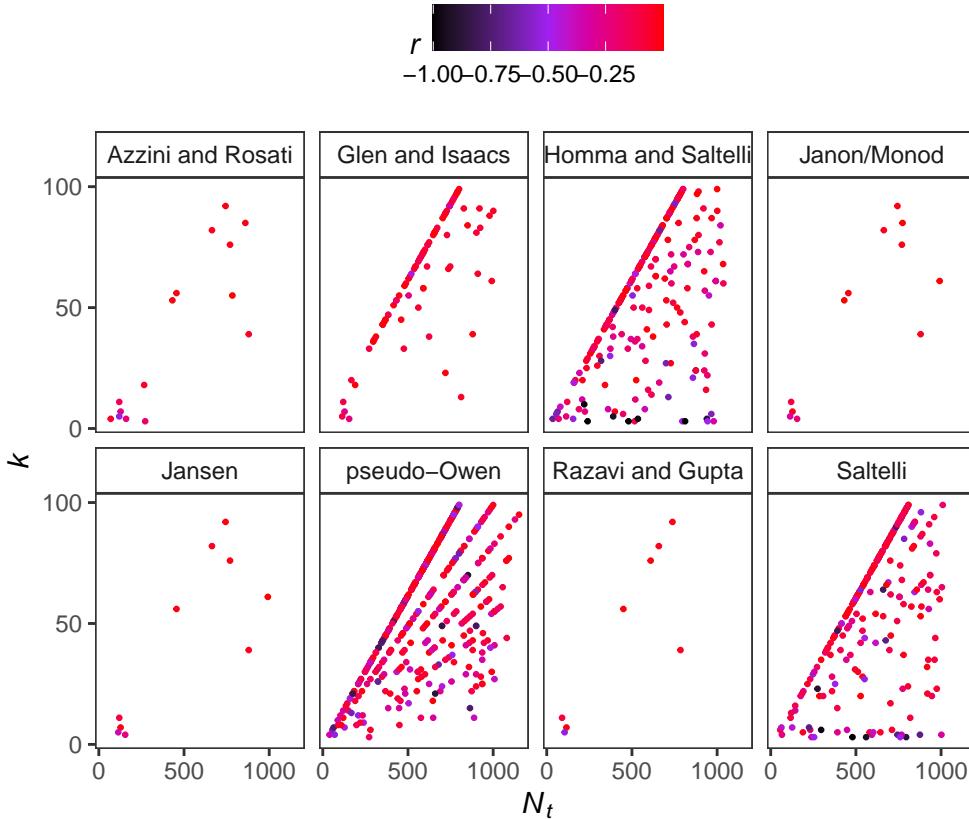
```
# PLOT PROPORTION OF NEGATIVE VALUES ----

A[, sum(correlation < 0) / .N, estimator] %>%
  ggplot(., aes(reorder(estimator, V1), V1)) +
  geom_bar(stat = "identity") +
  coord_flip() +
  labs(y = expression(frac(1, N) ~ sum(italic(r)[j] < 0, j == 1, N)),
       x = "") +
  theme_AP()
```



```
# MAP VALUES WITH NEGATIVE R -----
index.neg <- A[, .I[correlation < 0]]

A[index.neg] %>%
  ggplot(., aes(Nt, k, color = correlation)) +
  geom_point(size = 0.5) +
  scale_colour_gradient(colours = c("black", "purple", "red"),
                        name = expression(italic(r))) +
  scale_x_continuous(breaks = pretty_breaks(n = 3)) +
  scale_y_continuous(breaks = pretty_breaks(n = 3)) +
  labs(x = expression(italic(N[t])),
       y = expression(italic(k))) +
  facet_wrap(~estimator,
             ncol = 4) +
  theme_AP() +
  theme(legend.position = "top",
        strip.background = element_rect(fill = "white"))
```



```

# PLOT OUTPUT -----
# Compute median and quantiles
dt_median <- A[, .(median = median(correlation),
                  low.ci = quantile(correlation, 0.25),
                  high.ci = quantile(correlation, 0.75)), estimator]

A[, .(median = median(correlation),
      low.ci = quantile(correlation, 0.25),
      high.ci = quantile(correlation, 0.75)), estimator][order(median)]

##          estimator     median     low.ci    high.ci
## 1:   pseudo-Owen 0.2298777 0.02805562 0.4878208
## 2:       Saltelli 0.3192782 0.11163476 0.5663318
## 3: Homma and Saltelli 0.3302772 0.11494772 0.5721159
## 4:   Glen and Isaacs 0.4019751 0.18866989 0.6371165
## 5:  Azzini and Rosati 0.7778500 0.62032658 0.8996873
## 6:   Janon/Monod 0.8665914 0.75282215 0.9487651
## 7:       Jansen 0.8739175 0.75804357 0.9540298
## 8: Razavi and Gupta 0.9101595 0.81319190 0.9719266

a <- ggplot(A, aes(correlation)) +
  geom_rect(data = dt_median,
            aes(xmin = low.ci,
                xmax = high.ci,

```

```

        ymin = -Inf,
        ymax = Inf),
    fill = "blue",
    color = "white",
    alpha = 0.2,
    inherit.aes = FALSE) +
geom_histogram() +
geom_vline(data = dt_median, aes(xintercept = median),
    lty = 2,
    color = "red") +
facet_wrap(~estimator,
    ncol = 4) +
scale_x_continuous(breaks = pretty_breaks(n = 3)) +
scale_y_continuous(breaks = pretty_breaks(n = 3)) +
labs(x = expression(italic(r)),
    y = "Counts") +
theme_AP() +
theme(strip.background = element_rect(fill = "white"))

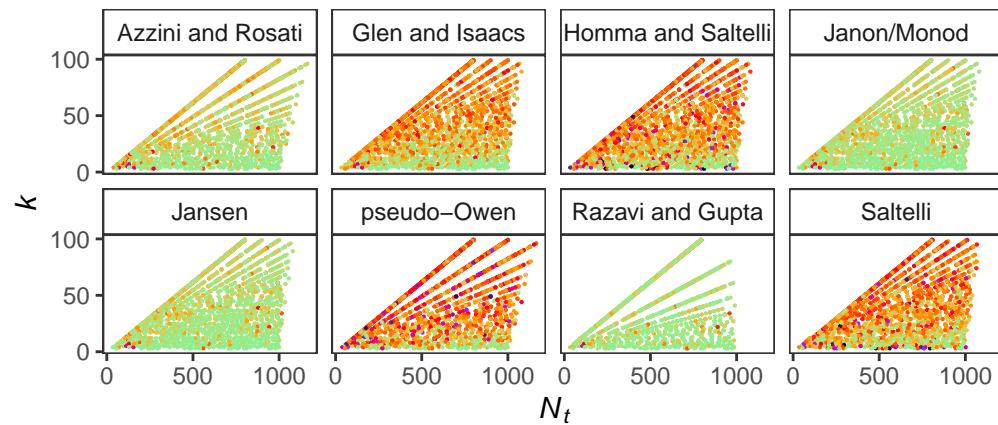
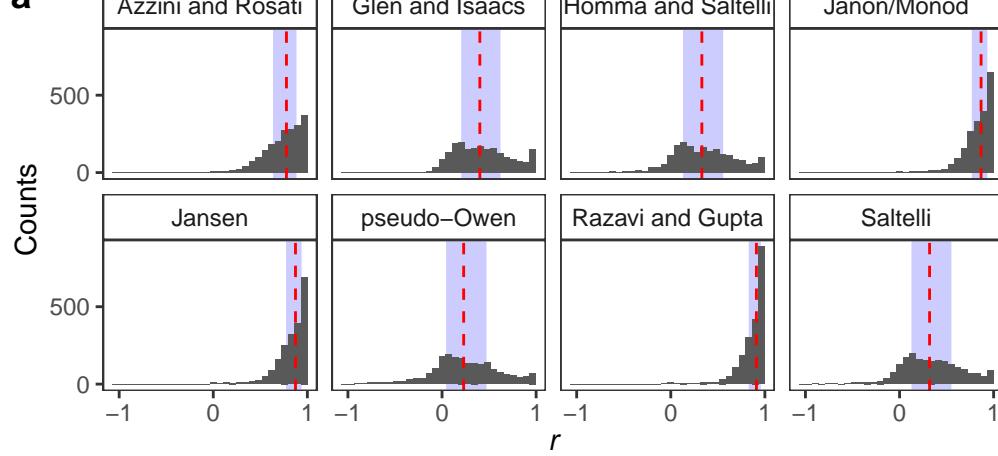
# Scatterplot
b <- ggplot(A, aes(Nt, k, color = correlation)) +
    geom_point(size = 0.1) +
    scale_colour_gradientn(colours = c("black", "purple", "red", "orange", "lightgreen"),
        name = expression(italic(r))) +
    scale_x_continuous(breaks = pretty_breaks(n = 3)) +
    scale_y_continuous(breaks = pretty_breaks(n = 3)) +
    labs(x = expression(italic(N[t])),
        y = expression(italic(k))) +
    facet_wrap(~estimator,
        ncol = 4) +
    theme_AP() +
    theme(legend.position = "top",
        strip.background = element_rect(fill = "white"))

# Ratio
c <- A[, ratio:= Nt / k] %>%
    ggplot(., aes(ratio, correlation)) +
    geom_point(alpha = 0.1, size = 0.2) +
    facet_wrap(~estimator,
        ncol = 4) +
    geom_smooth() +
    labs(x = expression(italic(N[t]/k)),
        y = expression(italic(r))) +
    scale_x_log10() +
    scale_y_continuous(breaks = pretty_breaks(n = 3)) +
    theme_AP() +
    theme(strip.background = element_rect(fill = "white"))

```

```
# Merge plot
plot_grid(a, b, ncol = 1, labels = "auto", rel_heights = c(0.85, 1))

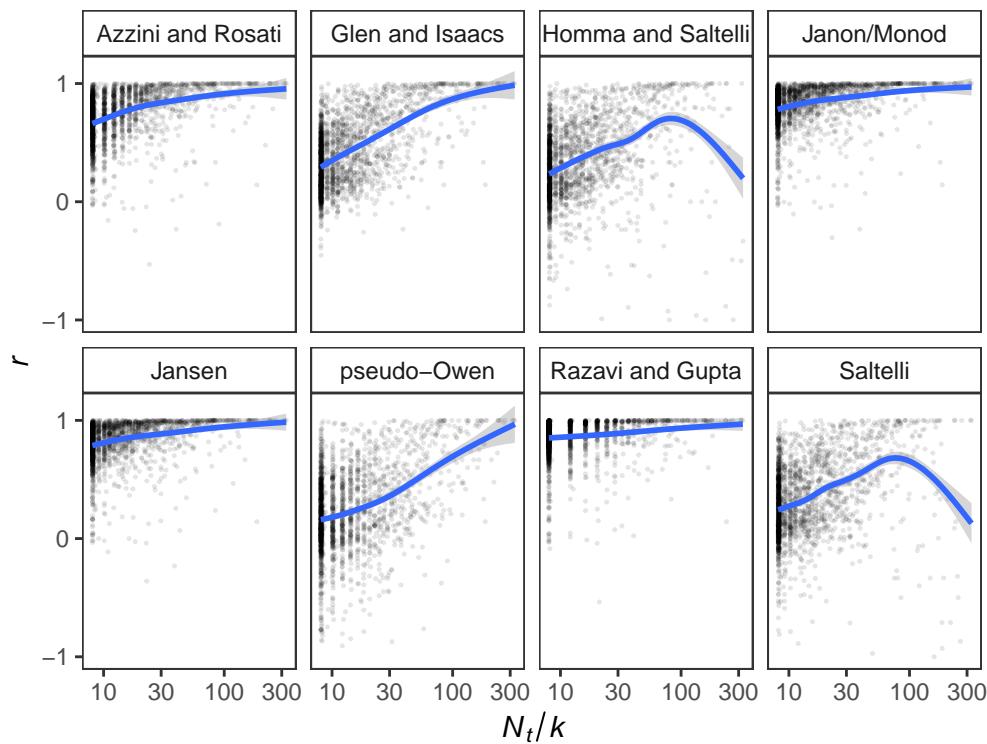
## `stat_bin()` using `bins = 30` . Pick better value with `binwidth` .
```



```
# DISPLAY THE RATIO NT/K FOR EACH SIMULATION AND ESTIMATOR -----
```

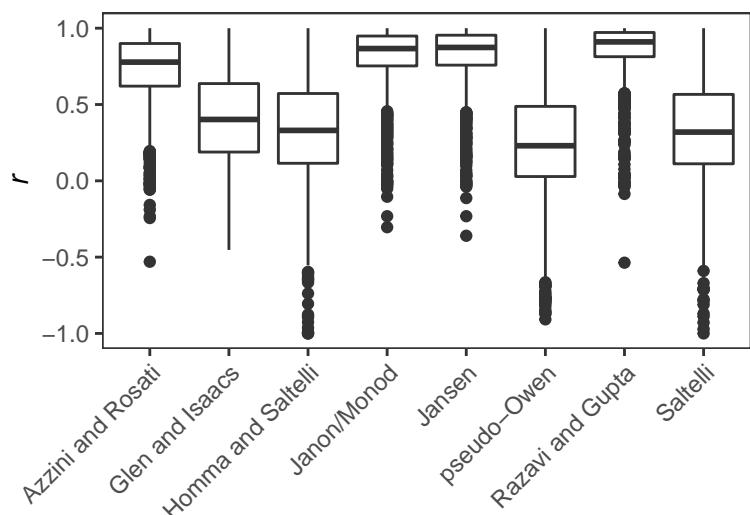
c

```
## `geom_smooth()` using method = 'gam' and formula 'y ~ s(x, bs = "cs")'
```



PLOT BOXPLOT -----

```
ggplot(A, aes(estimator, correlation)) +
  geom_boxplot() +
  labs(x = "",
       y = expression(italic(r))) +
  theme_AP() +
  theme(axis.text.x = element_text(angle = 45, hjust = 1))
```



PLOT MEDIANS -----

```
vv <- seq(5, 100, 5)
```

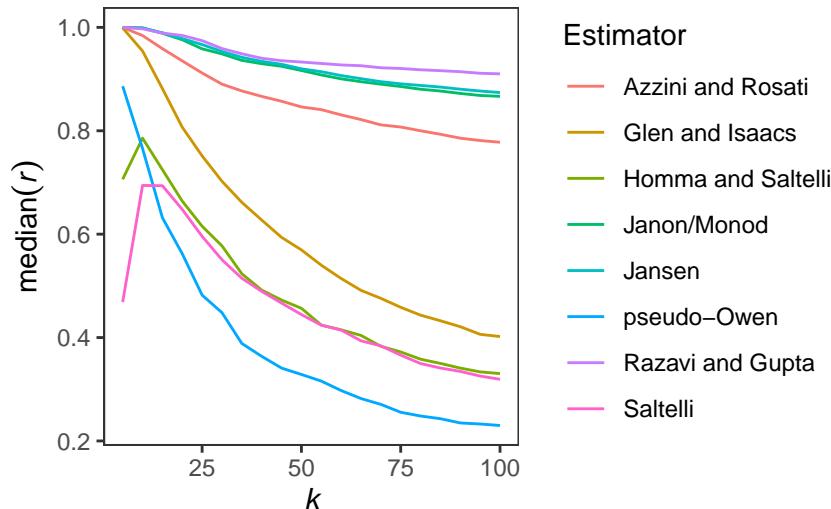
```

dt <- lapply(vv, function(x) A[k <= x, median(correlation), estimator])

names(dt) <- vv

rbindlist(dt, idcol = "k") %>%
  .[, k:= as.numeric(k)] %>%
  ggplot(., aes(k, V1, color = estimator)) +
  scale_color_discrete(name = "Estimator") +
  labs(x = expression(italic(k)),
       y = expression(median(italic(r)))) +
  geom_line() +
  theme_AP()

```



```

# Median Nt/k
dt.tmp <- A[, .(min = min(ratio), max = max(ratio))]

v <- seq(0, ceiling(dt.tmp$max), 20)
a <- c(v[1], rep(v[-c(1, length(v))], each = 2), v[length(v)])
indices <- matrix(a, ncol = 2 ,byrow = TRUE)

out <- list()
for(i in 1:nrow(indices)) {
  out[[i]] <- A[ratio > indices[i, 1] & ratio < indices[i, 2]]
}

names(out) <- rowmeans(indices)

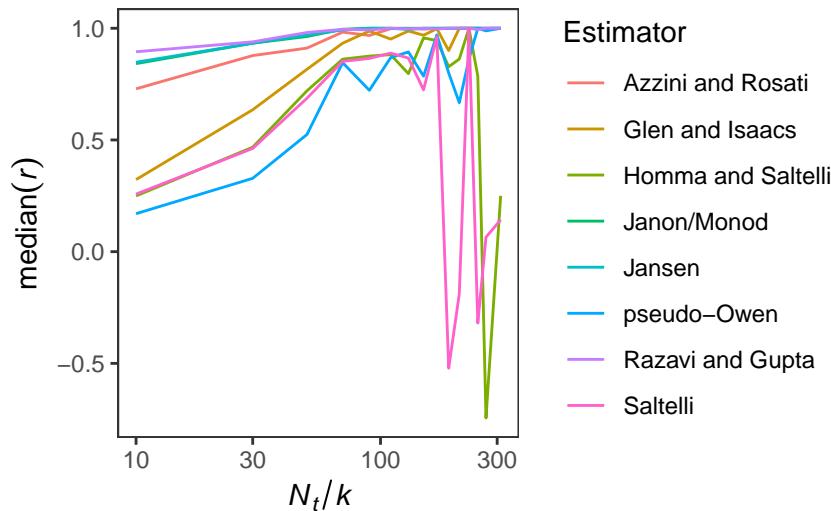
# Plot
lapply(out, function(x) x[, median(correlation, na.rm = TRUE), estimator]) %>%
  rbindlist(., idcol = "N") %>%
  .[, N:= as.numeric(N)] %>%
  ggplot(., aes(N, V1, group = estimator, color = estimator)) +

```

```

geom_line() +
  labs(x = expression(italic(N[t])/k)),
  y = expression(median(italic(r)))) +
  scale_color_discrete(name = "Estimator") +
  scale_x_log10() +
  theme_AP()

```

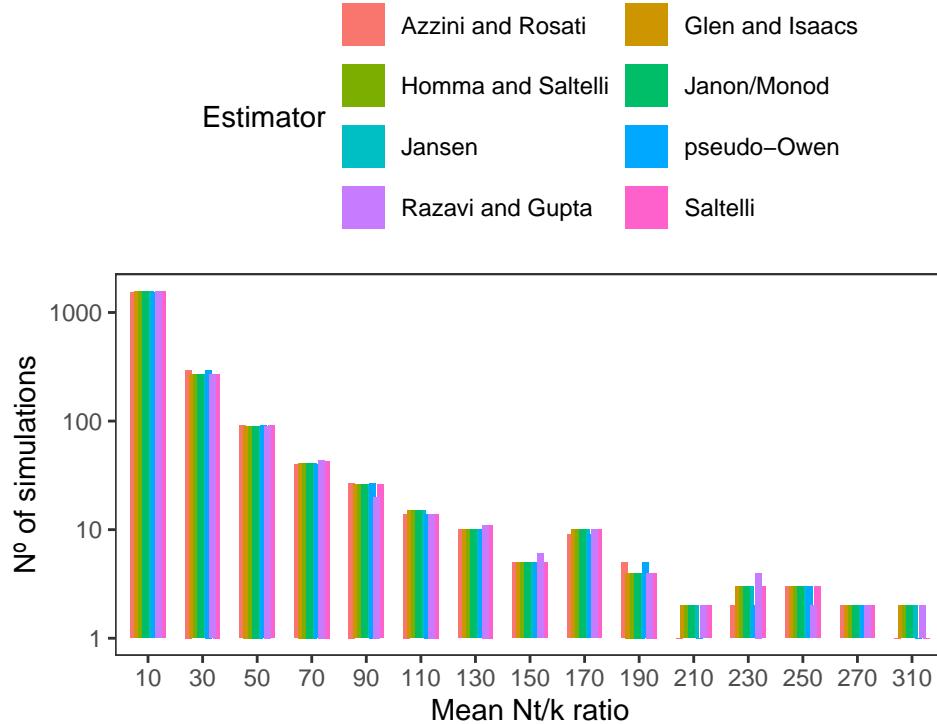


```
# PLOT NUMBER OF SIMULATIONS AGAINST NT/K RATIO -----
```

```

rbindlist(out, idcol = "samples")[, .N, .(estimator, samples)] %>%
  .[, samples := factor(samples, levels = rowmeans(indices))] %>%
  ggplot(., aes(samples, N, fill = estimator)) +
  scale_y_log10() +
  scale_fill_discrete(name = "Estimator") +
  labs(x = "Mean Nt/k ratio",
       y = "N° of simulations") +
  geom_bar(stat = "identity",
           position = position_dodge(0.6)) +
  theme_AP() +
  theme(legend.position = "top") +
  guides(fill = guide_legend(nrow = 4, byrow = TRUE))

```



5 Sensitivity analysis

5.1 Scatterplots

```
# SCATTERPLOTS OF MODEL OUTPUT AGAINST PARAMETERS ----

A <- setnames(A, c("N_t", "k_2", "k_3"), c("N[t]", "k[2]", "k[3]"))

scatter.dt <- melt(A, measure.vars = c("k[2]", "k[3]",
                                         "epsilon", "phi", "delta", "tau")) %>%
  split(., .$estimator)

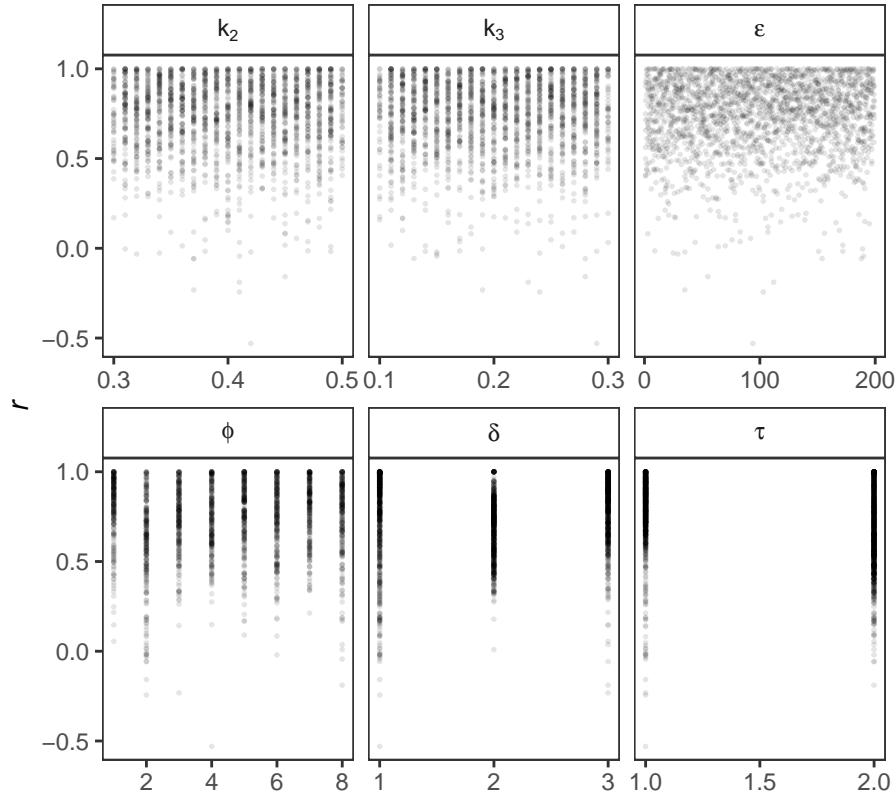
gg <- list()
for(i in names(scatter.dt)) {
  gg[[i]] <- ggplot(scatter.dt[[i]], aes(value, correlation)) +
    geom_point(alpha = 0.1, size = 0.3) +
    facet_wrap(~ variable,
               scales = "free_x",
               labeller = label_parsed,
               ncol = 3) +
    scale_color_manual(values = c("#00BFC4", "#F8766D")) +
    scale_x_continuous(breaks = pretty_breaks(n = 3)) +
    theme_AP() +
    labs(x = "", y = expression(italic(r))) +
    theme(strip.background = element_rect(fill = "white")) +
    ggtitle(names(scatter.dt[i]))}
```

```
}
```

```
gg
```

```
## $`Azzini and Rosati`
```

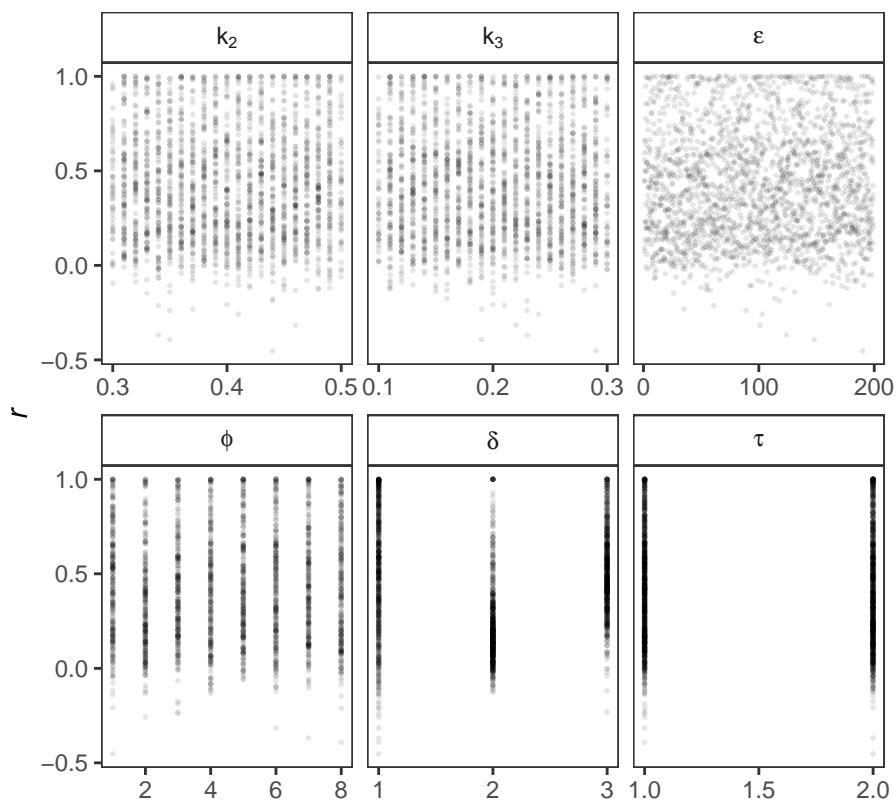
Azzini and Rosati



```
##
```

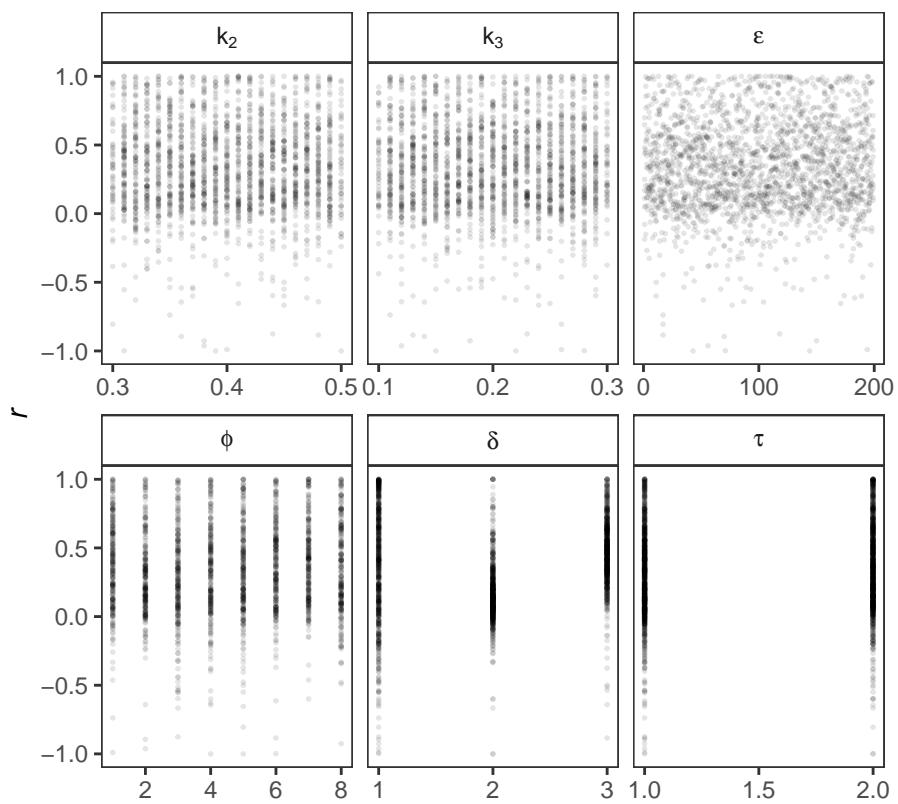
```
## $`Glen and Isaacs`
```

Glen and Isaacs



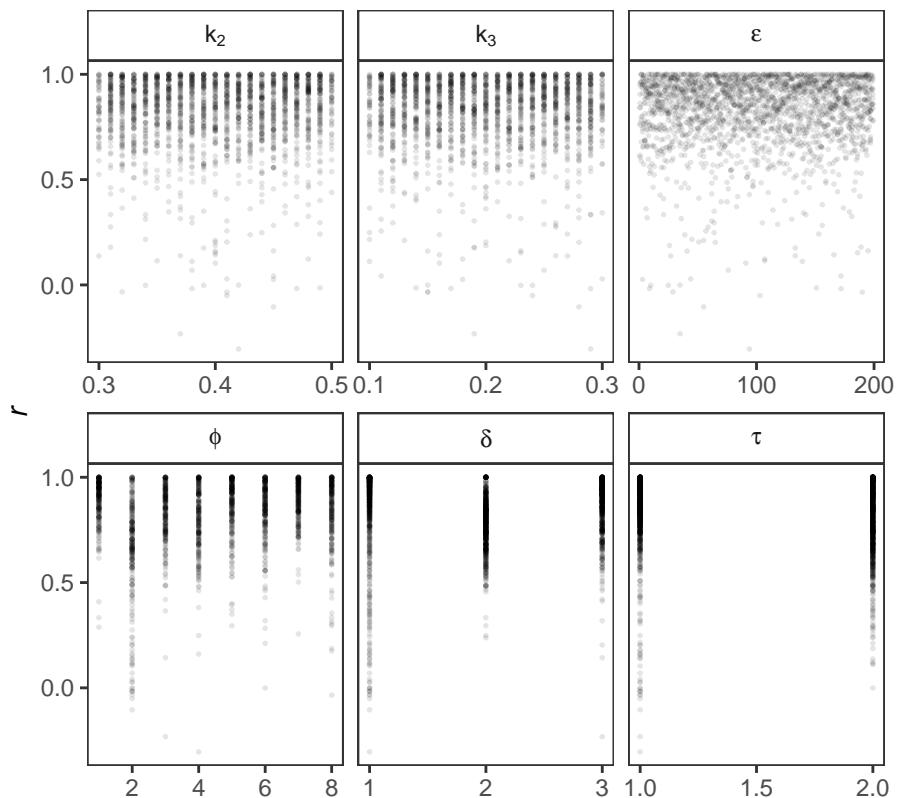
```
##  
## $`Homma and Saltelli`
```

Homma and Saltelli



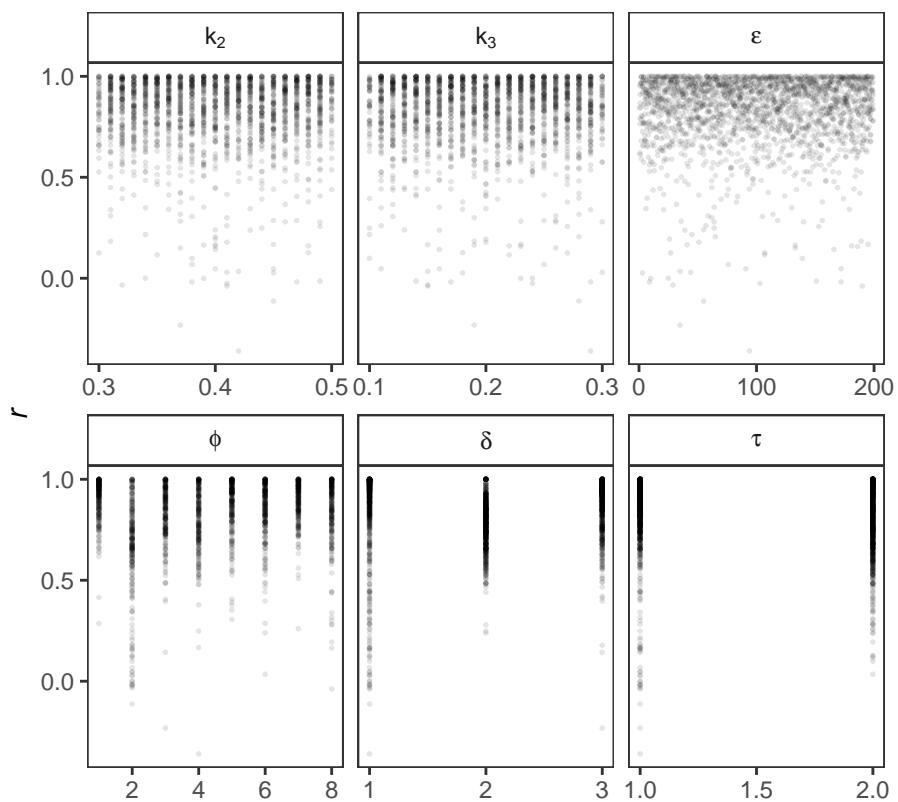
```
##  
## $`Janon/Monod`
```

Janon/Monod



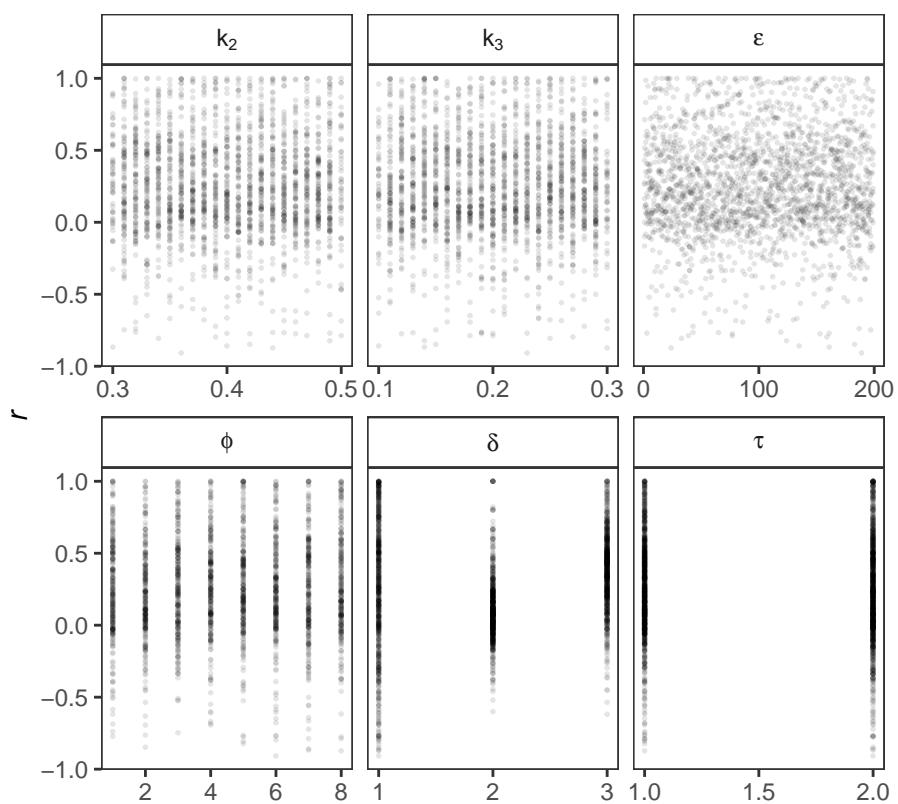
```
##  
## $Jansen
```

Jansen



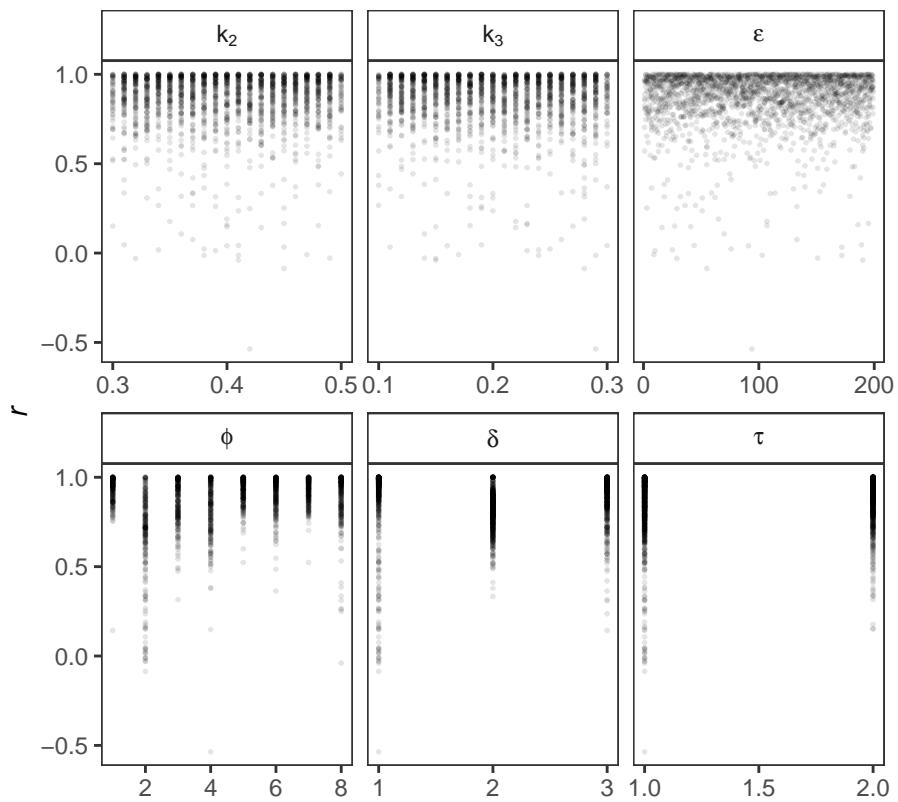
```
##  
## $`pseudo-Owen`
```

pseudo-Owen



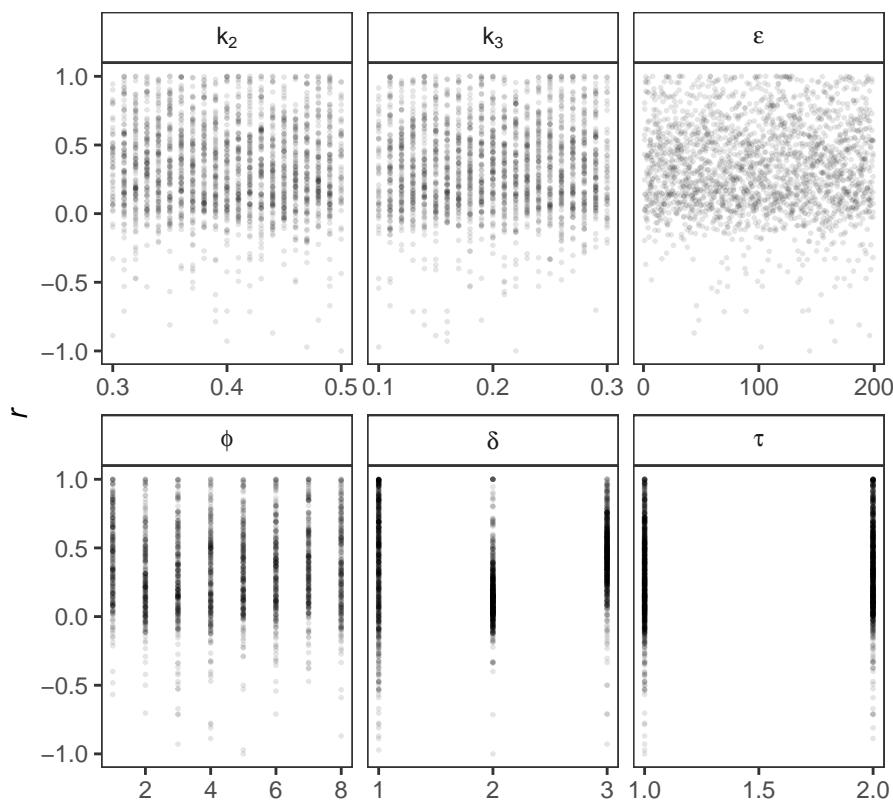
```
##  
## $`Razavi and Gupta`
```

Razavi and Gupta



```
##  
## $Saltelli
```

Saltelli



5.2 Sobol' indices

```
# SENSITIVITY ANALYSIS -----
# Show rows with NA
full_output[is.na(correlation), ]

## Empty data.table (0 rows and 20 cols): row,k_2,k_3,epsilon,phi,delta...
# Substitute NA by 0
full_output <- full_output[, correlation:= ifelse(is.na(correlation) == TRUE, 0, correlation)]

# New vector of parameters to plot better
params.new <- c("k[2]", "k[3]", "epsilon", "phi", "delta", "tau")

# Compute Sobol' indices except for the cluster Nt,k
indices <- full_output[!row %in% mat[, tail(.I, N)]][
  , sobol_indices(Y = correlation,
    N = N,
    params = params.new,
    first = "jansen",
    boot = TRUE,
    R = R,
    order = order),
```

```

estimator]

# Compute Sobol' indices for cluster
index.clusters <- mat[, tail(.I, 3 * N)]

indicesC <- rbind(full_output[row %in% 1:(2 * N)], full_output[row %in% index.clusters)][
  , sobol_indices(Y = correlation,
    N = N,
    params = c("N[t]~k", "f(x)", "delta~tau"),
    first = "jansen",
    boot = TRUE,
    R = R,
    order = "first"),
  estimator]

# Final
indices <- rbind(indices, indicesC[parameters == "N[t]~k"]) %>%
  merge(., VY.dt, by = "estimator") %>%
  .[, unnormalized:= original * VY]

# Export
fwrite(indices, "indices.csv")
fwrite(indicesC, "indicesC.csv")

# PLOT SOBOL' INDICES -----
indices.tmp <- copy(indices)
indicesC.tmp <- copy(indicesC)

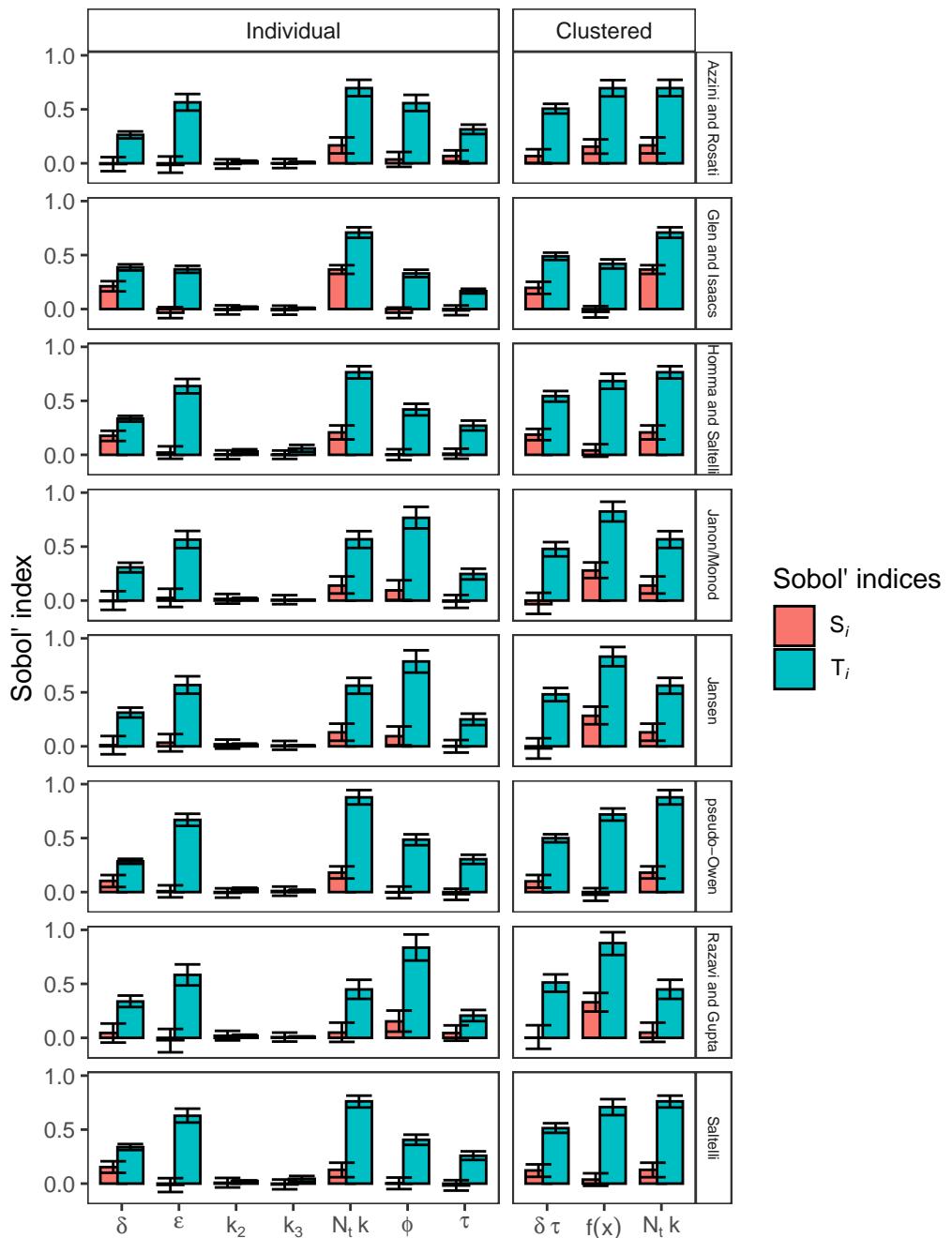
colsDelete <- c("VY", "unnormalized")
indices.tmp <- indices.tmp[, (colsDelete) := NULL] %>%
  .[, type:= "Individual"]

indicesC.tmp <- indicesC.tmp[, type:= "Clustered"] %>%
  .[, type:= factor(type, levels = c("Individual", "Clustered"))]

rbind(indices.tmp, indicesC.tmp) %>%
  .[sensitivity == "Si" | sensitivity == "Ti"] %>%
  ggplot(., aes(parameters, original, fill = sensitivity)) +
  geom_bar(stat = "identity",
    position = position_dodge(0.6),
    color = "black") +
  geom_errorbar(aes(ymin = low.ci,
    ymax = high.ci),
    position = position_dodge(0.6)) +
  scale_x_discrete(labels = ggplot2:::parse_safe) +
  scale_y_continuous(breaks = pretty_breaks(n = 3)) +

```

```
facet_grid(estimator~type,
           scales = "free_x",
           space = "free_x") +
  labs(x = "",
       y = "Sobol' index") +
  scale_fill_discrete(name = "Sobol' indices",
                      labels = c(expression(S[italic(i)]),
                                expression(T[italic(i)]))) +
  theme_AP() +
  theme(legend.position = "right",
        strip.text.y = element_text(size = 6),
        strip.background = element_rect(fill = "white"))
```



```
# PLOT UNNORMALIZED SOBOL' INDICES -----
```

```
indicesC <- indicesC %>%
  merge(., VY.dt, by = "estimator") %>%
  .[, unnormalized:= original * VY]

all.indicesC <- list(indices, indicesC)

hh <- list()
for(i in seq_along(all.indicesC)) {
```

```

hh[[i]] <- ggplot(all.indicesC[[i]][sensitivity == "Si" |
                                         sensitivity == "Ti"],
                   aes(parameters, unnormalized, fill = sensitivity)) +
  geom_bar(stat = "identity",
            position = position_dodge(0.6),
            color = "black") +
  scale_x_discrete(labels = ggplot2:::parse_safe) +
  scale_y_continuous(breaks = pretty_breaks(n = 3)) +
  facet_wrap(~estimator,
             ncol = 4) +
  labs(x = "",
       y = expression(T[italic(i)] * V(Y))) +
  scale_fill_discrete(name = "Sobol' indices",
                      labels = c(expression(S[italic(i)]),
                                 expression(T[italic(i)]))) +
  theme_AP() +
  theme(legend.position = "none",
        strip.background = element_rect(fill = "white"))
}

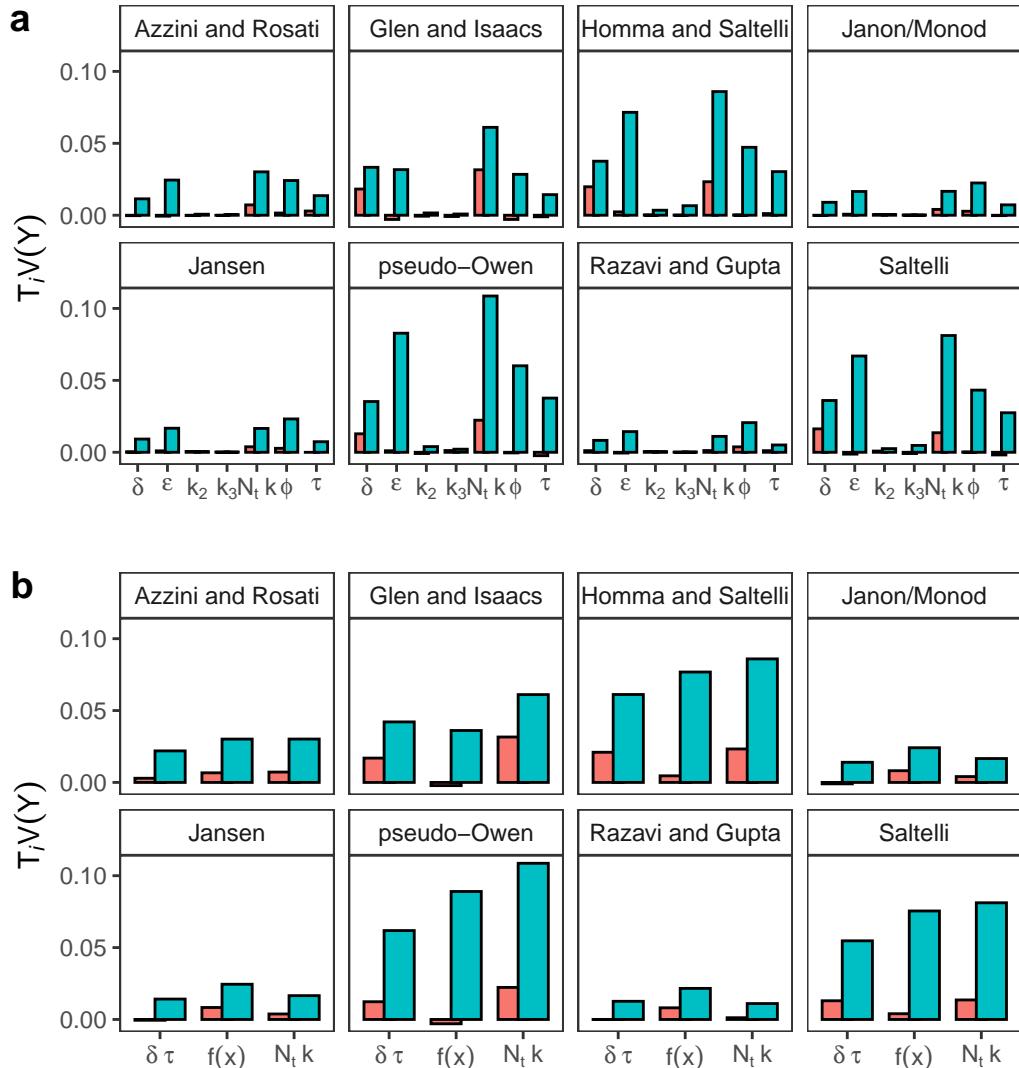
# Get legend
legend <- get_legend(hh[[1]]) + theme(legend.position = "top")

top <- plot_grid(hh[[1]], hh[[2]] + theme(legend.position = "none"),
                  align = "hv", labels = "auto", ncol = 1)

plot_grid(legend, top, ncol = 1, rel_heights = c(0.1, 1))

```

Sobol' indices S_i T_i



SUM OF FIRST-ORDER INDICES -----

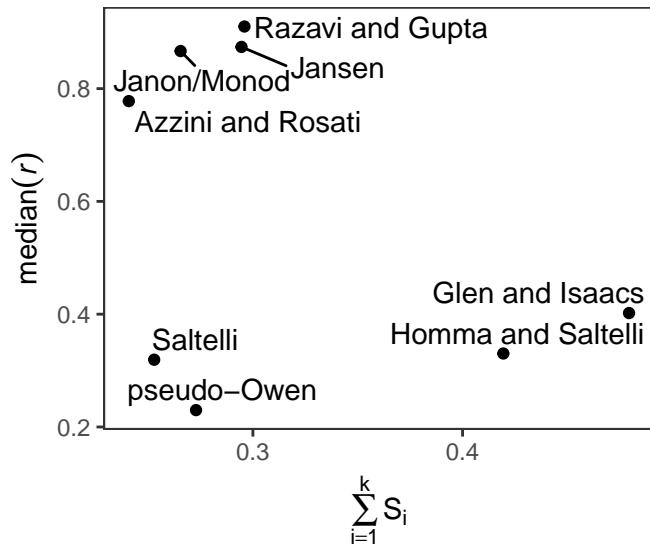
```

indices[sensitivity == "Si", sum(original), estimator]

##           estimator      V1
## 1: Azzini and Rosati 0.2408162
## 2:   Glen and Isaacs 0.4793858
## 3: Homma and Saltelli 0.4194474
## 4:     Janon/Monod 0.2655164
## 5:       Jansen 0.2945569
## 6: Razavi and Gupta 0.2959547
## 7:     Saltelli 0.2529163
## 8:   pseudo-Owen 0.2728713

```

```
# Plot
merge(indices[sensitivity == "Si", sum(original), estimator],
      dt_median, by = "estimator") %>%
  ggplot(., aes(V1, median)) +
  geom_point() +
  labs(x = expression(sum(S[i], i==1, k)),
       y = expression(median(italic(r)))) +
  geom_text_repel(aes(label = estimator)) +
  theme_AP()
```



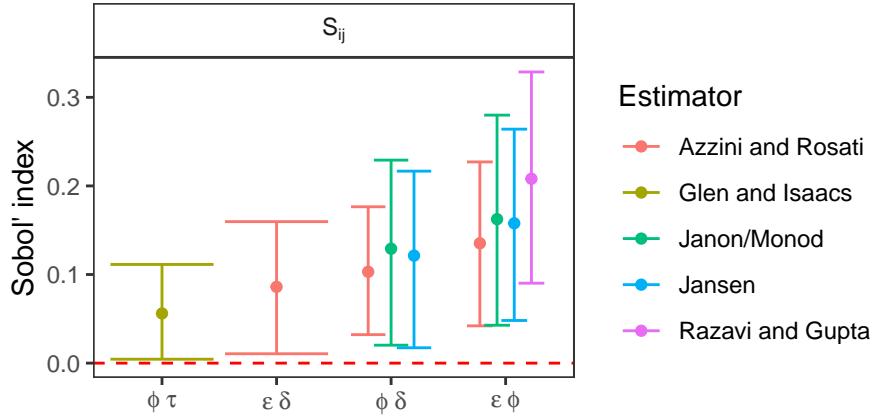
```
# PLOT SECOND-ORDER EFFECTS -----
```

```
indices[sensitivity == "Sij"] %>%
  .[low.ci > 0] %>%
  .[, sensitivity:= ifelse(sensitivity %in% "Sij", "S[ij]", sensitivity)] %>%
  .[, parameters:= gsub(parameters,
                        pattern = ".",
                        replacement = "~",
                        fixed = TRUE)] %>%
  ggplot(., aes(reorder(parameters, original), original, color = estimator)) +
  geom_point(position = position_dodge(0.6)) +
  geom_errorbar(aes(ymax = high.ci, ymin = low.ci),
                position = position_dodge(0.6)) +
  geom_hline(yintercept = 0,
             lty = 2,
             color = "red") +
  facet_grid(~sensitivity,
             scales = "free_x",
             space = "free_x",
             labeller = label_parsed) +
  scale_x_discrete(labels = ggplot2:::parse_safe) +
  scale_color_discrete(name = "Estimator") +
```

```

  labs(x = "",
    y = "Sobol' index") +
  theme_AP() +
  theme(strip.background = element_rect(fill = "white"))

```



```
# EXPORT SOBOL' INDICES -----
```

```
fwrite(dt_median, "dt_median.csv")
```

```
# CALCULATE EXPLORATIVITY AND ECONOMY -----
```

```
# Create functions
```

```
economy_fun <- function(n, k, h) {
  Owen <- k / (2 * (k + 1))
  Razavi <- ((1 - h) * 2 * k) / (h * (1 + (1 / h - 1) * k))
  Azzini <- k / (k + 1)
  Glen <- k / (k + 1)
  Saltelli <- k / (k + 2)
  Homma <- k / (k + 1)
  Jansen <- k / (k + 1)
  Janon <- k / (k + 1)
  eff <- c(Owen, Razavi, Azzini, Glen, Saltelli,
           Homma, Jansen, Janon)
  return(eff)
}
```

```
explorativity_fun <- function(n, k, h) {
```

```
  Owen <- 3 / (2 * (1 + k))
  Razavi <- 1 / (h * (1 + (1 / h - 1) * k))
  Azzini <- 1 / (k + 1)
  Glen <- 2 / (k + 1)
  Saltelli <- 2 / (k + 1)
  Homma <- 2 / (k + 1)
  Jansen <- 2 / (k + 1)
  Janon <- 2 / (k + 1)
  ec <- c(Owen, Razavi, Azzini, Glen,
```

```

        Saltelli, Homma, Jansen, Janon)
    return(ec)
}

full_fun <- function(n, k, h) {
  arrangement <- data.table(c("pseudo-Owen",
                             "Razavi and Gupta",
                             "Azzini and Rosati",
                             "Glen and Isaacs",
                             "Saltelli",
                             "Homma and Saltelli",
                             "Jansen",
                             "Janon/Monod"))

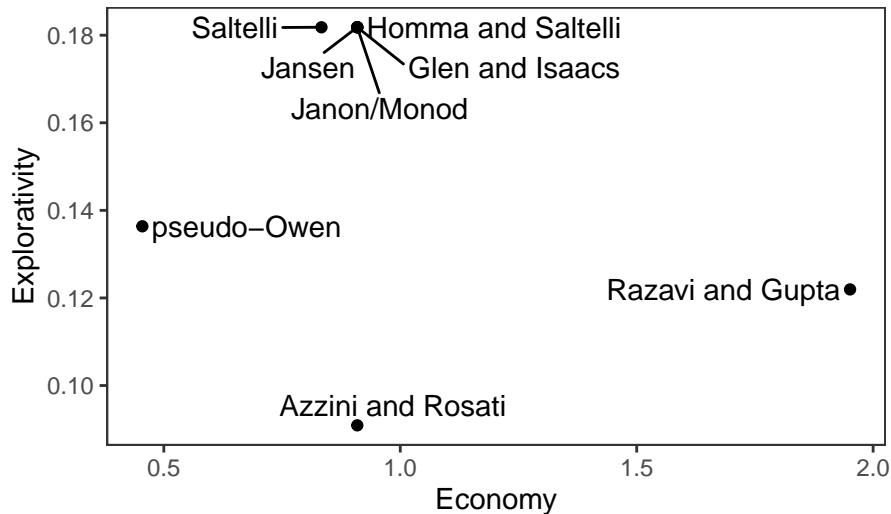
  ec <- economy_fun(n, k, h)
  ex <- explorativity_fun(n, k, h)
  out <- data.table(cbind(ec, ex))
  final <- cbind(arrangement, out)
  setnames(final, "V1", "Arrangement")
  return(final)
}

vec.k <- c(10, 100, 1000)
data.ex <- lapply(vec.k, function(k) full_fun(n = 4, k = k, h = 0.2))
names(data.ex) <- vec.k

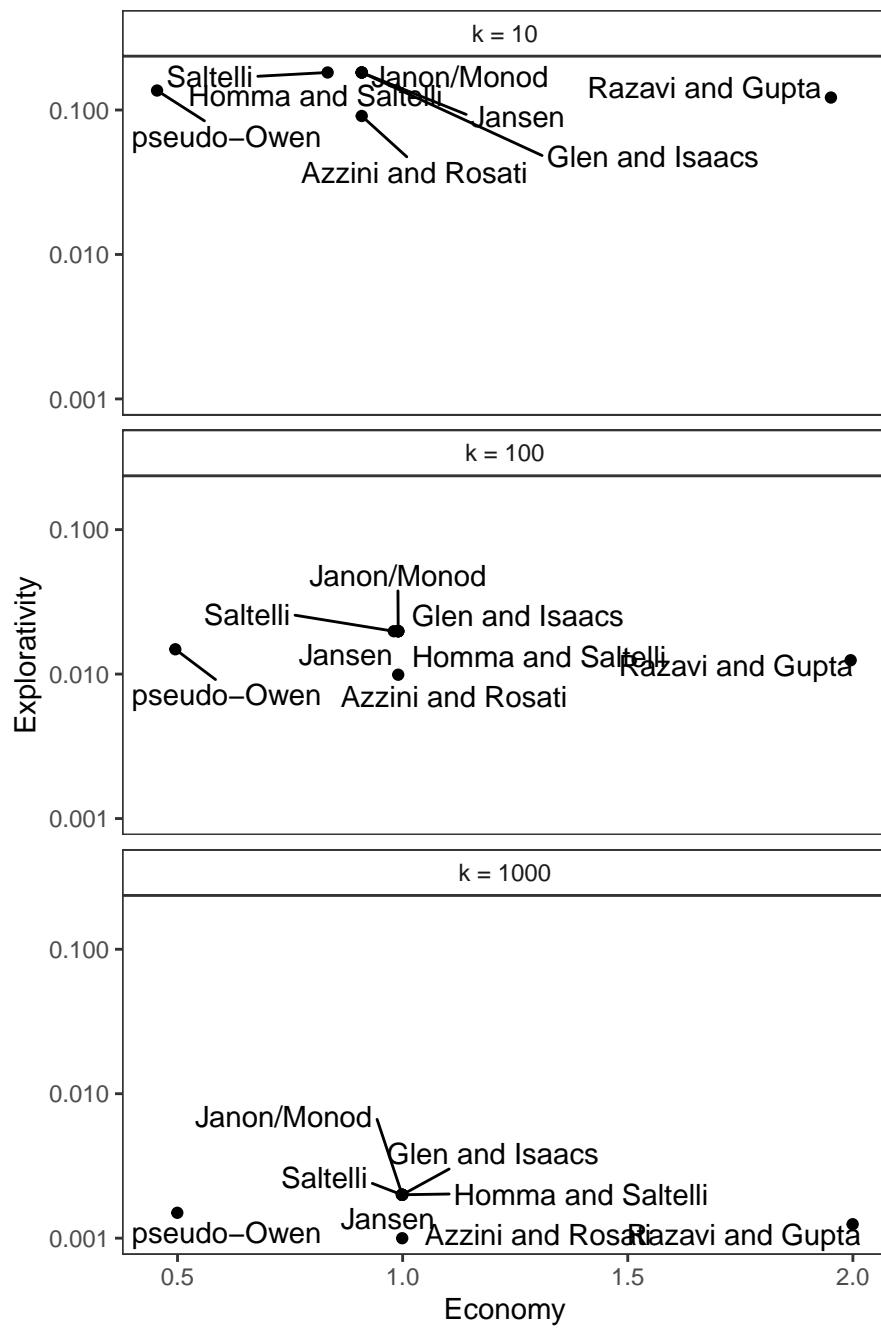
plot.data.ex <- rbindlist(data.ex, idcol = "k") %>%
  .[, k:= as.numeric(k)]

# PLOT EXPLORATIVITY -----
ggplot(plot.data.ex[k == 10], aes(ec, ex)) +
  geom_point() +
  labs(x = "Economy",
       y = "Explorativity") +
  geom_text_repel(aes(label = Arrangement)) +
  theme_AP()

```



```
# PLOT EXPLORATIVITY -----
plot.data.ex[, k:= paste("k = ", k, sep = "")] %>%
  ggplot(., aes(ec, ex)) +
  geom_point() +
  labs(x = "Economy",
       y = "Explorativity") +
  geom_text_repel(aes(label = Arrangement)) +
  facet_wrap(~k, ncol = 1) +
  scale_y_log10() +
  theme_AP() +
  theme(strip.background = element_rect(fill = "white"))
```



6 Session information

```
# SESSION INFORMATION -----  
  
sessionInfo()  
  
## R version 4.0.3 (2020-10-10)  
## Platform: x86_64-apple-darwin17.0 (64-bit)  
## Running under: macOS Catalina 10.15.6  
##  
## Matrix products: default  
## BLAS: /Library/Frameworks/R.framework/Versions/4.0/Resources/lib/libRblas.dylib  
## LAPACK: /Library/Frameworks/R.framework/Versions/4.0/Resources/lib/libRlapack.dylib  
##  
## locale:  
## [1] en_US.UTF-8/en_US.UTF-8/en_US.UTF-8/C/en_US.UTF-8/en_US.UTF-8  
##  
## attached base packages:  
## [1] parallel stats      graphics grDevices utils      datasets  methods  
## [8] base  
##  
## other attached packages:  
## [1] pcaPP_1.9-73           checkpoint_0.4.9       ggrepel_0.8.2  
## [4] sensobol_0.3            logitnorm_0.8.37      benchmarkme_1.0.4  
## [7] cowplot_1.1.0          scales_1.1.1          data.table_1.13.2  
## [10] Rfast_2.0.1            RcppZiggurat_0.1.6    doParallel_1.0.16  
## [13] iterators_1.0.13        foreach_1.5.1        forcats_0.5.0  
## [16] stringr_1.4.0           dplyr_1.0.2           purrr_0.3.4  
## [19] readr_1.3.1             tidyverse_1.3.0        tibble_3.0.4  
## [22] ggplot2_3.3.2          tidyverse_1.3.0        RcppArmadillo_0.10.1.0.0  
## [25] Rcpp_1.0.5  
##  
## loaded via a namespace (and not attached):  
## [1] httr_1.4.2              jsonlite_1.7.1        modelr_0.1.8  
## [4] Rdpack_2.0              assertthat_0.2.1      blob_1.2.1  
## [7] cellranger_1.1.0        yaml_2.2.1            remotes_2.2.0  
## [10] pillar_1.4.6            backports_1.2.0       lattice_0.20-41  
## [13] glue_1.4.2              digest_0.6.27         rbibutils_1.3  
## [16] rvest_0.3.6             colorspace_1.4-1      htmltools_0.5.0  
## [19] Matrix_1.2-18           pkgconfig_2.0.3       broom_0.7.0  
## [22] haven_2.3.1            mvtnorm_1.1-1         farver_2.0.3  
## [25] generics_0.0.2          ellipsis_0.3.1       withr_2.3.0  
## [28] cli_2.1.0              magrittr_1.5           crayon_1.3.4  
## [31] readxl_1.3.1           evaluate_0.14         fs_1.5.0  
## [34] fansi_0.4.1             xml2_1.3.2            tools_4.0.3  
## [37] hms_0.5.3              gbRd_0.4-11           lifecycle_0.2.0  
## [40] munsell_0.5.0           reprex_0.3.0          compiler_4.0.3  
## [43] rlang_0.4.8             grid_4.0.3            rstudioapi_0.11
```

```

## [46] labeling_0.4.2           rmarkdown_2.5.2        gtable_0.3.0
## [49] codetools_0.2-16          DBI_1.1.0            curl_4.3
## [52] benchmarkmeData_1.0.4    R6_2.5.0             lubridate_1.7.9
## [55] knitr_1.30                stringi_1.5.3       vctrs_0.3.4
## [58] dbplyr_1.4.4              tidyselect_1.1.0    xfun_0.18

## Return the machine CPU
cat("Machine:      "); print(get_cpu()$model_name)

## Machine:
## [1] "Intel(R) Core(TM) i9-9900K CPU @ 3.60GHz"

## Return number of true cores
cat("Num cores:   "); print(detectCores(logical = FALSE))

## Num cores:
## [1] 8

## Return number of threads
cat("Num threads: "); print(detectCores(logical = TRUE))

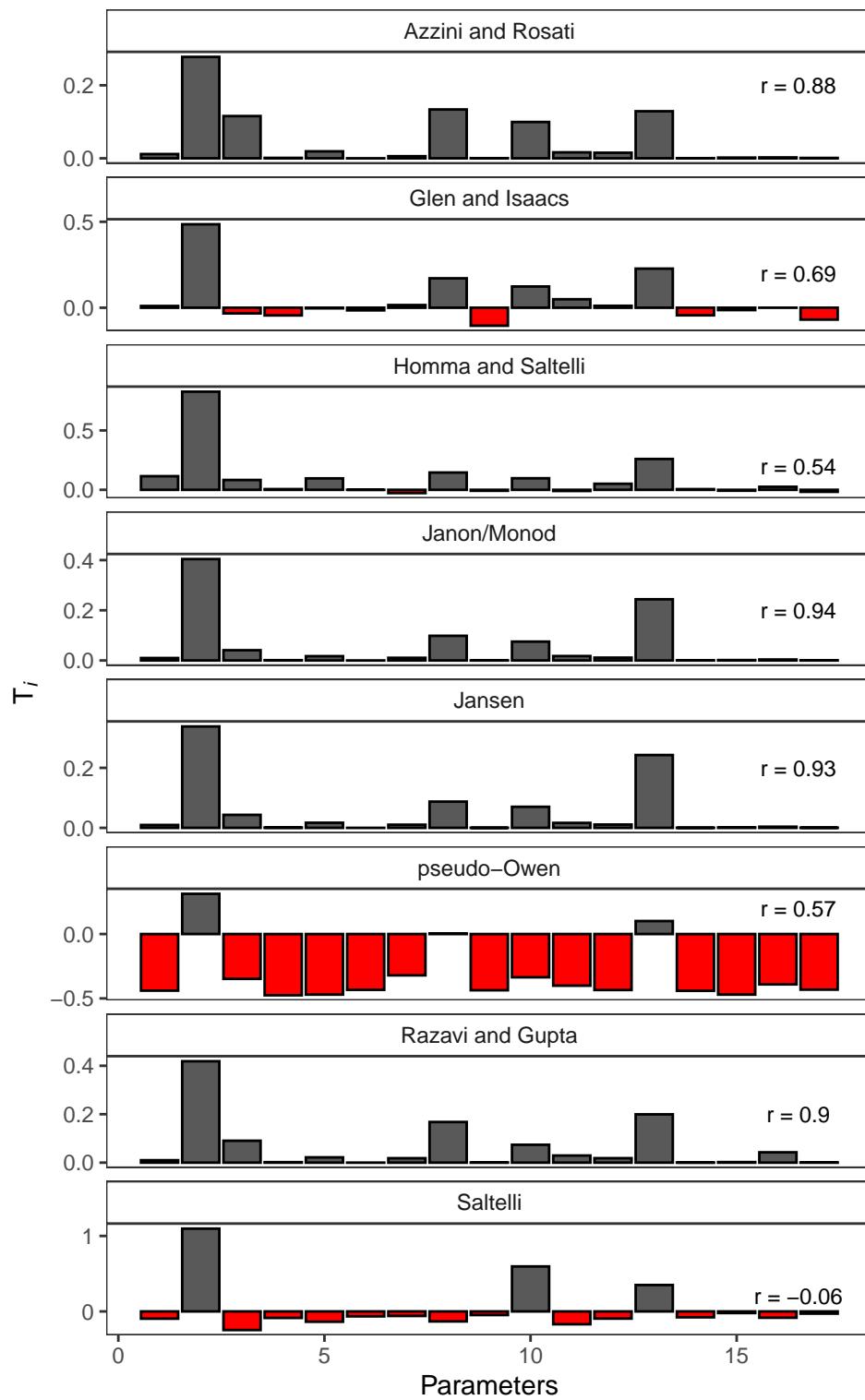
## Num threads:
## [1] 16

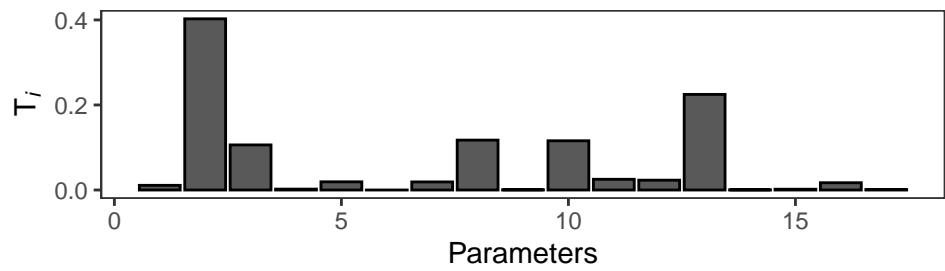
## Return the machine RAM
cat("RAM:         "); print (get_ram()); cat("\n")

## RAM:
## 34.4 GB

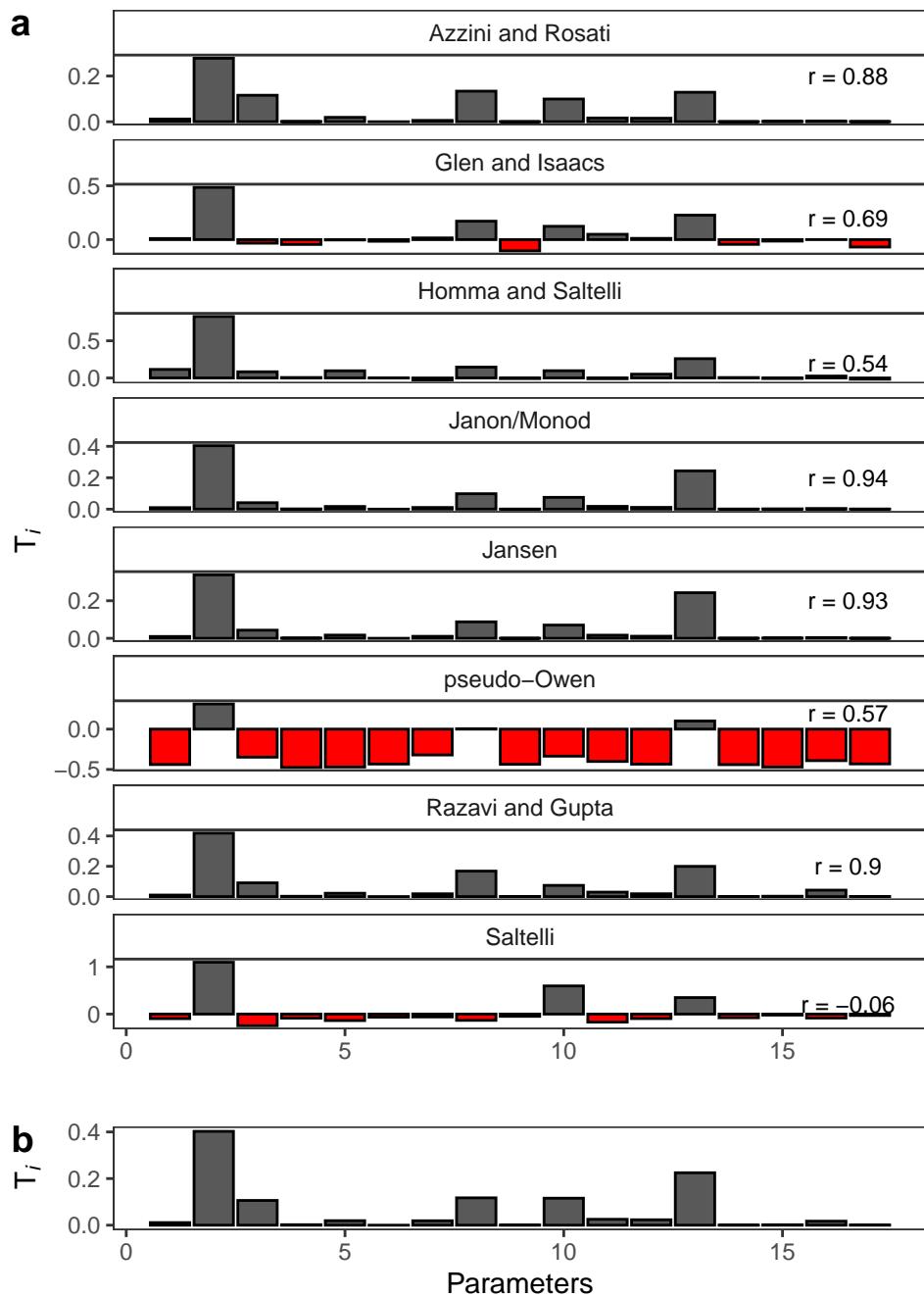
```

7 References





```
## Warning: Graphs cannot be horizontally aligned unless the axis parameter is set.  
## Placing graphs unaligned.
```



Becker, William. 2020. "Metafunctions for benchmarking in sensitivity analysis." *Reliability Engineering & System Safety*, August, 107189. <https://doi.org/10.1016/j.ress.2020.107189>.