

# Probabilistic Nested Designs with Sensitivity Indices, version 7

## Load packages.

```
In [1]: require(abind)
require(data.table)
require(deSolve)
require(magrittr)
require(np)
require(rTensor)

require(ggplot2)
require(GGally)
```

Loading required package: abind  
Loading required package: data.table  
Loading required package: deSolve  
Loading required package: magrittr  
Loading required package: np  
Nonparametric Kernel Methods for Mixed Datatypes (version 0.60-9)  
[vignette("np\_faq",package="np") provides answers to frequently asked questions]  
[vignette("np",package="np") an overview]  
[vignette("entropy\_np",package="np") an overview of entropy-based methods]  
Loading required package: rTensor  
Loading required package: ggplot2  
Loading required package: GGally  
Registered S3 method overwritten by 'GGally':  
  method from  
  +.gg      ggplot2

## Function to generate simulations.

```

In [51]: # Create a multivariate function with specified properties:
#         tmax: maximum time
#         multiplicities: number of correlations each parameter has
#         degrees: polynomial degree of each parameter
#         dimension: the dimension of the output
#         returns a multivariate function of the vector of parameters and t
ime
makeGenerator <- function(multiplicities, degrees, dimension) {

  single <- function(degree) {
    x0 <- runif(1)
    z0 <- runif(1)
    function(x) {
      if (x < x0)
        0
      else
        z0 * (x - x0)^degree
    }
  }

  locations <- lapply(multiplicities, function(m) sample(1:dimension, m))
  functions <- lapply(degrees, single)

  start <- runif(dimension, -0.25, 0.75)
  coefs <- matrix(runif(dimension^2, -0.25, 0.75), dimension, dimension)
  shift <- matrix(runif(dimension^2, -0.25, 0.75), dimension, dimension)

  function(x, ts) {
    z <- rep(0, dimension)
    for (i in 1:length(locations))
      for (j in locations[[i]])
        z[j] <- z[j] + functions[[i]](x[i])
    ode(start, ts, function(t, y, params) {list((coefs %*% y) * z
* (1 - ((shift %*% y) * z)))})
  }
}

```

## Functions for nesting designs

```
In [52]: ssa.digits <- 30
```

```
In [53]: ssa.scale <- 2^ssa.digits
```

```
In [54]: ssa.level <- function(i) {  
  if (i == 0)  
    return(ssa.digits)  
  n <- 0  
  while (bitwAnd(i, 1) == 0) {  
    n <- n + 1  
    i <- bitwShiftR(i, 1)  
  }  
  n  
}
```

```
In [55]: ssa.depth <- function(i) {  
  ssa.digits - ssa.level(i)  
}
```

```
In [56]: ssa.corner <- function(i, offset=0){  
  d <- ssa.level(i) - offset  
  c(i - 2^d, i + 2^d)  
}
```

```
In [57]: ssa.corners <- function(i1, i2, i3, offset=0) {  
  axis <- function(s1, s2, s3) {  
    if (s1 == 0)  
      4 * s1 + 2 * s2 + s3  
    else  
      4 * (1 - s1) + 2 * (1 - s2) + (1 - s3)  
  }  
  merge(merge(merge(  
    data.table(  
      d1 = 1,  
      d2 = 2,  
      d3 = 4  
    ),  
    data.table(d1=1, i1=ssa.corner(i1, offset=offset), s1=0:1), b  
y="d1", allow.cartesian=TRUE),  
    data.table(d2=2, i2=ssa.corner(i2, offset=offset), s2=0:1), b  
y="d2", allow.cartesian=TRUE),  
    data.table(d3=4, i3=ssa.corner(i3, offset=offset), s3=0:1), b  
y="d3", allow.cartesian=TRUE  
  )[,  
    .(  
      i1,  
      i2,  
      i3,  
      axis=mapapply(axis, s1, s2, s3)  
    )]  
}
```

```
In [58]: ssa.candidates <- function(i1, i2, i3) {  
  CJ(  
    i1=c(i1, ssa.corner(i1)),  
    i2=c(i2, ssa.corner(i2)),  
    i3=c(i3, ssa.corner(i3))  
  ),  
  .(  
    i1,  
    i2,  
    i3,  
    x1 = i1 / ssa.scale,  
    x2 = i2 / ssa.scale,  
    x3 = i3 / ssa.scale  
  )  
}
```

```

In [59]: ssa.start <- function() {
  result <- CJ(
    i1=c(0, ssa.scale),
    i2=c(0, ssa.scale),
    i3=c(0, ssa.scale)
  )[,
    .(
      sequence = 1:8      ,
      generation = 0      ,
      i1          ,
      i2          ,
      i3          ,
      x1 = i1 / ssa.scale,
      x2 = i2 / ssa.scale,
      x3 = i3 / ssa.scale,
      compute = TRUE      ,
      measure = FALSE     ,
      probed  = TRUE      ,
      s1 = 0              ,
      s2 = 0              ,
      s3 = 0              ,
      s  = 0              ,
    )
  ]
  result <- rbind(
    result,
    data.table(
      sequence = 9      ,
      generation = NA    ,
      i1 = ssa.scale / 2,
      i2 = ssa.scale / 2,
      i3 = ssa.scale / 2,
      x1 = 0.5          ,
      x2 = 0.5          ,
      x3 = 0.5          ,
      compute = FALSE   ,
      measure = TRUE     ,
      probed  = FALSE   ,
      s1 = 0. / 0.      ,
      s2 = 0. / 0.      ,
      s3 = 0. / 0.      ,
      s  = 0. / 0.      ,
    )
  )
  setkeyv(result, c("i1", "i2", "i3"))
}

```

```

In [60]: ssa.compute <- function(f, ts, xs, ys) {
  needed <- xs[compute == TRUE]
  for (row in 1:nrow(needed)) {
    ys <- rbind(
      ys,
      data.table(f(as.numeric(needed[row, .(x1, x2, x3)]), ts))
    [,
      .(
        sequence = needed[row, sequence],
        t = time,
        y1 = `1`,
        y2 = `2`,
        y3 = `3`
      )
    ]
  )
  }
  xs[compute == TRUE, compute := FALSE]
  ys
}

```

```

In [61]: ssa.measure <- function(xs, ys, focus=2) {

  result <- merge(
    merge(
      xs[measure == TRUE][, ssa.corners(i1, i2, i3), by=.(center=sequence)],
      xs,
      by=c("i1", "i2", "i3"),
      allow.cartesian=TRUE
    )[, .(center, axis, sequence)],
    ys,
    by="sequence",
    allow.cartesian=TRUE
  )[,
    .(y1 = mean(y1), y2 = mean(y2), y3 = mean(y3)), by=.(sequence
=center, axis, t)
  ][,
    .(s1 = sd(y1), s2 = sd(y2), s3 = sd(y3)), by=.(sequence, t)
  ][,
    .(s1 = max(s1), s2 = max(s2), s3 = max(s3)), by=sequence
  ]

  xs[
    sequence %in% result$sequence,
    `:=`(
      s1=result$s1,
      s2=result$s2,
      s3=result$s3,
      generation=xs[, 1 + max(generation, na.rm=TRUE)]
    )
  ]

  y1min = min(ys$y1)
  y1max = max(ys$y1)
  y2min = min(ys$y2)
  y2max = max(ys$y2)
  y3min = min(ys$y3)
  y3max = max(ys$y3)

  normalize <- function (i, s1, s2, s3)
    max(s1 / (y1max - y1min), s2 / (y2max - y2min), s3 / (y3max -
y3min)) * focus^ssa.depth(i)
  xs[, `:=`(measure = FALSE, s = mapply(normalize, i1, s1, s2, s3))
  ]

  result
}

```

```

In [62]: ssa.probe <- function(xs, alpha=1, maxdepth=-log(sqrt(.Machine$doubl
e.eps),2)) {
  choices <- xs[probed == FALSE & !is.na(s), .(sequence, s)]
  if (nrow(choices) == 1)
    choice <- choices$sequence
  else
    choice <- sample(choices$sequence, 1, prob = choices$s^alpha)
  probe <- xs[sequence == choice, .(sequence, i1, i2, i3)]
  xs[
    probed == FALSE & mapply(ssa.depth, i1) < maxdepth,
    .(sequence, i1, i2, i3, rank=frank(-s, ties.method="random"))
  ][
    rank == 1,
    .(sequence, i1, i2, i3)
  ]
  candidates <- ssa.candidates(probe$i1, probe$i2, probe$i3)
  setkeyv(candidates, c("i1", "i2", "i3"))
  candidates <- candidates[!xs]
  n <- xs[, max(sequence)]
  if (nrow(candidates) > 0) {
    candidates <- candidates[, .(
      sequence = n + (1:nrow(candidates)),
      generation = NA,
      i1, i2, i3,
      x1, x2, x3,
      compute = TRUE,
      measure = FALSE,
      probed = FALSE,
      s1 = 0. / 0.,
      s2 = 0. / 0.,
      s3 = 0. / 0.,
      s = 0. / 0.
    )]
    n <- candidates[, max(sequence)]
  }
  probes <- ssa.corners(probe$i1, probe$i2, probe$i3, offset=1)
  probes <- probes[, .(
    sequence = n + (1:nrow(probes)),
    generation = NA,
    i1, i2, i3,
    x1 = i1 / ssa.scale,
    x2 = i2 / ssa.scale,
    x3 = i3 / ssa.scale,
    compute = FALSE,
    measure = TRUE,
    probed = FALSE,
    s1 = 0. / 0.,
    s2 = 0. / 0.,
    s3 = 0. / 0.,
    s = 0. / 0.
  )]
  if (nrow(candidates) > 0)
    result <- rbind(
      xs,
      candidates,
      probes
    )
}

```



```

    )
  else
    result <- rbind(
      xs,
      probes
    )
  result[sequence == probe$sequence, `:=`(compute=TRUE, measure=FALSE, probed=TRUE)]
  setkeyv(result, c("i1", "i2", "i3"))
  list(
    sequence=probe$sequence,
    xs=result
  )
}

```

## Reproducible random numbers.

```
In [63]: RNGkind("Mersenne-Twister", "Inversion", "Rejection")
```

```
In [64]: set.seed(46)
```

## Create a simulation function.

```
In [65]: f <- makeGenerator(c(2, 2, 3), c(0, 1, 2), 3)
```

## Example application

### Set time resolution.

```
In [17]: ts <- (0:20) / 2
```

### Initialize experiment at bounds of domain.

```
In [18]: xs <- ssa.start()
```

In [19]: xs

A data.table: 9 x 15

sequence	generation	i1	i2	i3	x1	x2	x3	compute	m
<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<lgl>	
1	0	0	0	0	0.0	0.0	0.0	TRUE	
2	0	0	0	1073741824	0.0	0.0	1.0	TRUE	
3	0	0	1073741824	0	0.0	1.0	0.0	TRUE	
4	0	0	1073741824	1073741824	0.0	1.0	1.0	TRUE	
9	NA	536870912	536870912	536870912	0.5	0.5	0.5	FALSE	
5	0	1073741824	0	0	1.0	0.0	0.0	TRUE	
6	0	1073741824	0	1073741824	1.0	0.0	1.0	TRUE	
7	0	1073741824	1073741824	0	1.0	1.0	0.0	TRUE	
8	0	1073741824	1073741824	1073741824	1.0	1.0	1.0	TRUE	

## Iterate several times.

There are two adjustable parameters: `focus` and `alpha`. The `focus` parameter controls the emphasis on depth, with `focus = 0` ignoring depth and `focus = 2` being neutral with respect to depth. The `alpha` parameter controls how peaked the probability of selecting an area to probe is on the nonlinearity there, with `alpha = 0` for no sensitivity to nonlinearity and `alpha = 1` for probability of selection proportional to the amount of nonlinearity.

```
In [21]: ys <- NULL
for (k in 1:2500) {
  # Perform pending computations.
  ys <- rbind(ys, ssa.compute(f, ts, xs, NULL))
  # Measure divergences.
  ssa.measure(xs, ys, focus = 0.25)
  # Select where to probe further.
  result <- ssa.probe(xs, alpha = 1)
  # Update grid.
  xs <- result$xs
}
ys <- rbind(ys, ssa.compute(f, ts, xs, NULL))
```

## Save results.

```
In [22]: write.table(xs, file="xs-v7.csv", row.names=FALSE, sep=",")
xs %>% dim
```

```
54744 15
```

```
In [23]: write.table(ys, file="ys-v7.csv", row.names=FALSE, sep=",")
ys %>% dim
```

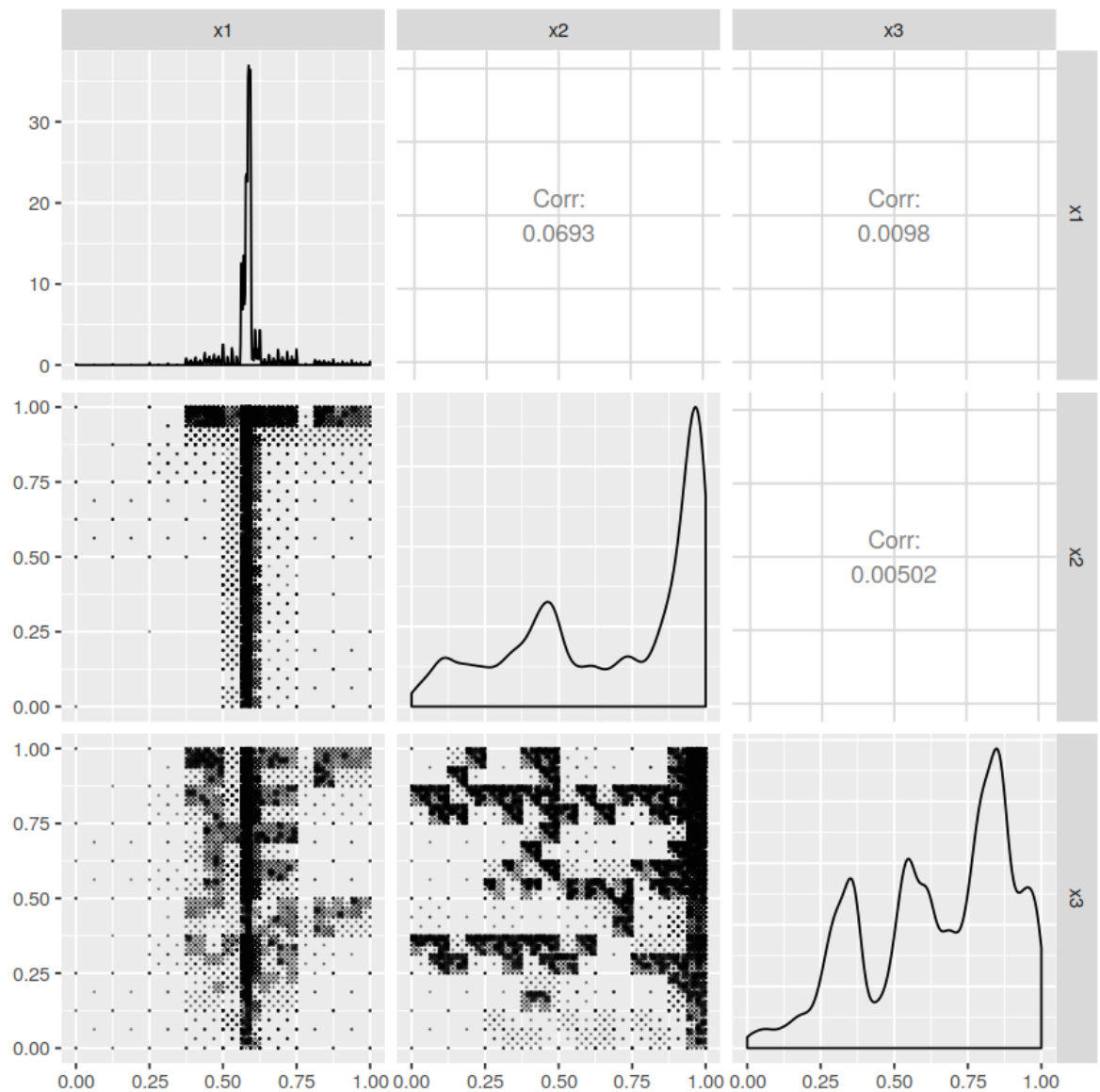
```
782103 5
```

```
In [2]: xs <- fread("xs-v7.csv")
setkeyv(xs, c("i1", "i2", "i3"))

ys <- fread("ys-v7.csv")
setkeyv(ys, c("sequence", "t"))
```

**Plot sampling pattern.**

```
In [24]: ggpairs(
  xS,
  6:8,
  # mapping = aes(color=factor(mapply(ssa.depth, i1))),
  lower = list(continuous = wrap("points", alpha = 0.2, size=0.05))
)
```



**Compute sensitivity indices.**

**Organize the data for sensitivity analysis, just using the final timestep.**

```
In [3]: xys.wide <- merge(
  xs[probed == TRUE, .(sequence, x1, x2, x3)],
  ys[t == 10, .(sequence, y1, y2, y3)]
)
xys.wide %>% head
```

A data.table: 6 x 7

sequence	x1	x2	x3	y1	y2	y3
<int>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>
1	0	0	0	-0.1900320	0.5144967	0.4093612
2	0	0	1	5.0782072	1.9488670	2.2320137
3	0	1	0	-0.1437119	0.5511418	0.4093612
4	0	1	1	4.9929679	1.9513684	2.2540234
5	1	0	0	-0.1900320	8.2070979	-1.8065843
6	1	0	1	5.2555080	1.3837551	1.9296318

**Function for computing first-order sensitivity index.**

***Saltelli's method.***

This sometimes yields values above one or below zero, but the bootstrapping seems to work better.

```

In [70]: sensitivity_indices <- function(xy) {
  # Convert to long format.
  xy.tall <- melt(
    melt(
      xy,
      id.vars=c("sequence", "y1", "y2", "y3"),
      measure.vars=c("x1", "x2", "x3"),
      variable.name="i",
      value.name="x"
    ),
    id.vars=c("sequence", "i", "x"),
    measure.vars=c("y1", "y2", "y3"),
    variable.name="j",
    value.name="y"
  )
  # Compute `E[y_j]`.
  ybar.all <- xy.tall[, .(ybar.all=mean(y)), by=.(i, j)]
  # Compute `Var[y_j]`.
  var.all <- merge(ybar.all, xy.tall)[, .(var.all=mean((y - ybar.all)^2)), by=.(i, j)]
  # Compute `E[y_j | x_i = x]`.
  ybar.x <- xy.tall[, .(ybar.x=mean(y)), by=.(i, j, x)]
  # Compute `Var[E[y_j | x_i = x]]`.
  var.x <- merge(ybar.all, ybar.x)[, .(var.x=mean((ybar.x - ybar.all)^2)), by=.(i, j)]
  # Compute the first-order sensitivity `Var[E[y_j|x_i=x]] / Var[y]`
  `.`
  s.x <- merge(var.all, var.x)[, .(i, j, s.x=var.x/var.all)]
  # Return the result.
  s.x
}

```

### ***Literal method.***

This yields values between zero and one, but the bootstrapping doesn't seem to work well.

```

In [71]: sensitivity_indices <- function(xy) {
  # Convert to long format.
  xy.tall <- melt(
    melt(
      xy,
      id.vars=c("sequence", "y1", "y2", "y3"),
      measure.vars=c("x1", "x2", "x3"),
      variable.name="i",
      value.name="x"
    ),
    id.vars=c("sequence", "i", "x"),
    measure.vars=c("y1", "y2", "y3"),
    variable.name="j",
    value.name="y"
  )
  # Compute `Var[y_j]`.
  var.all <- xy.tall[, .(var.all=var(y)), by=.(i, j)]
  # Compute `E[y_j | x_i = x]`.
  ybar.x <- xy.tall[, .(ybar.x=mean(y)), by=.(i, j, x)]
  # Compute `Var[ E[y_j | x_i = x] ]`.
  var.x <- ybar.x[, .(var.x=var(ybar.x)), by=.(i, j)]
  # Compute the first-order sensitivity `Var[E[y_j|x_i=x]] / Var[y]`
  .
  s.x <- merge(var.all, var.x)[, .(i, j, s.x=var.x/var.all)]
  # Return the result.
  s.x
}

```

### Function to resample sensitivity indices.

This is a very simple bootstrap, just resampling with replacement. It looks like it is biased.

```

In [72]: sensitivity_indices_resample <- function(xy, n) {
  rows <- nrow(xy)
  result <- cbind(n=0, sensitivity_indices(xy))
  for (i in 1:n) {
    xy.sample <- xy[sample(1:rows, rows, replace=TRUE)]
    xy.sample[, sequence:=1:rows]
    result <- rbind(
      result,
      cbind(n=i, sensitivity_indices(xy.sample))
    )
  }
  result
}

```

### Compute the indices.

```
In [73]: s.x <- sensitivity_indices_resample(xys.wide, 1000)
s.x[n == 0]
```

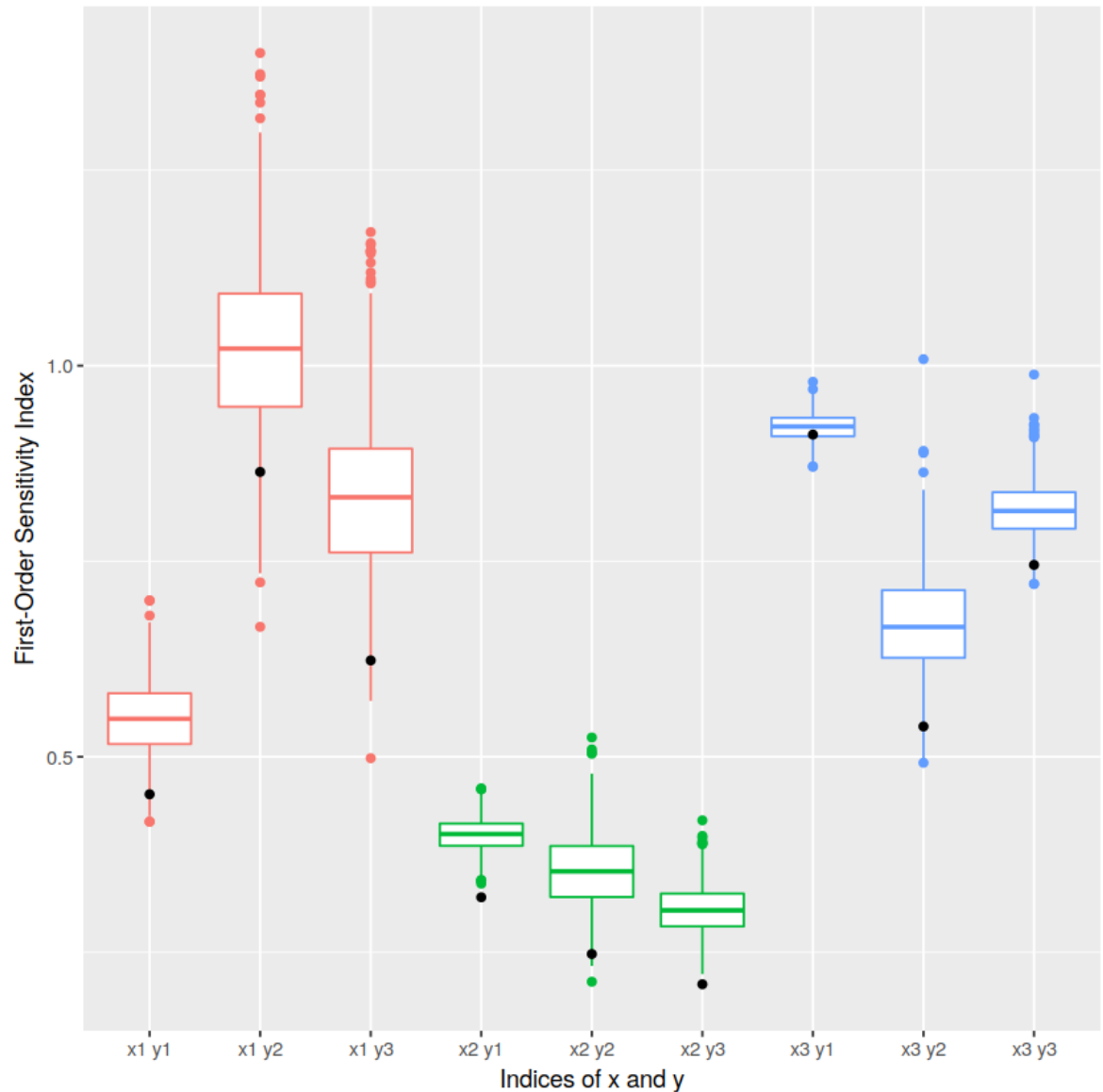
A data.table: 9 x 4

n	i	j	s.x
<dbl>	<fct>	<fct>	<dbl>
0	x1	y1	0.4521335
0	x1	y2	0.8641704
0	x1	y3	0.6232122
0	x2	y1	0.3204271
0	x2	y2	0.2479287
0	x2	y3	0.2092908
0	x3	y1	0.9118776
0	x3	y2	0.5387907
0	x3	y3	0.7453109

**Plot the indices.**



```
In [74]: ggplot(s.x[n != 0], aes(x=paste(i, j), y=s.x, color=i)) +
  geom_boxplot() +
  xlab("Indices of x and y") +
  ylab("First-Order Sensitivity Index") +
  guides(color=FALSE) +
  geom_point(data=s.x[n == 0], aes(x=paste(i, j), y=s.x), color="black")
```



The first-order sensitivity index is  $s_{ij} = \text{Var}[E[y_j | x_i]] / \text{Var}[y_j]$ .

The solid black dots show the unbootstrapped estimates. **It is suspicious that the bootstrapped estimates are almost always higher by more than one quartile.**

It looks like  $y_2$  and  $y_3$  have strong dependence on  $x_1$ , and  $y_1$  and  $y_3$  have strong dependence on  $x_3$ . The results do not seem dependence of  $x_2$ . The dependence of  $y_1$  on  $x_1$  and of  $y_2$  on  $x_3$  is moderate.

## Experiment with partitioning data into two contiguous sets and then computing sensitivity indices.

Even though there are the global sensitivities listed above, there might be regions of  $(x_1, x_2, x_3)$  where some of those sensitivities disappear or others appear. To investigate this, let's try partitioning  $(x_1, x_2, x_3)$  by hyperplanes of constant  $x_1$ ,  $x_2$ , or  $x_3$  and then computing the sensitivity indices on either side of partition. (This is somewhat like Monte-Carlo filtering.)

This is an attempt to find boundaries in the phase diagram of  $(x_1, x_2, x_3)$ .

**Separate by x values into a left and right partition.**

```
In [75]: us <- xs[,
          .(u=round((1:6)/7, 3)),
          by=. (sequence)
        ],
          .(k=c("split on x1", "split on x2", "split on x3")),
          by=. (sequence, u)
        ]
```

```
In [76]: uxys.wide <- merge(us, xys.wide, by="sequence", allow.cartesian=TRUE)
uxys.wide <- rbind(
  uxys.wide[k == "split on x1" & x1 < u, .(k, u, s='below split',
sequence, x1, x2, x3, y1, y2, y3)],
  uxys.wide[k == "split on x1" & x1 >= u, .(k, u, s='above split',
sequence, x1, x2, x3, y1, y2, y3)],
  uxys.wide[k == "split on x2" & x2 < u, .(k, u, s='below split',
sequence, x1, x2, x3, y1, y2, y3)],
  uxys.wide[k == "split on x2" & x2 >= u, .(k, u, s='above split',
sequence, x1, x2, x3, y1, y2, y3)],
  uxys.wide[k == "split on x3" & x3 < u, .(k, u, s='below split',
sequence, x1, x2, x3, y1, y2, y3)],
  uxys.wide[k == "split on x3" & x3 >= u, .(k, u, s='above split',
sequence, x1, x2, x3, y1, y2, y3)]
)
uxys.wide %>% dim
```

45144 10

```

In [77]: us <- us[, .(s=c("below split", "above split")), by=.(k, u)]
us.x <- NULL
for (row in 1:nrow(us))
  us.x <- rbind(
    us.x,
    cbind(
      us[row, .(k, u, s)],
      sensitivity_indices_resample(merge(us[row], uxys.wide, by
=c("k", "u", "s")), 1000)
    )
  )
us.x %>% dim

```

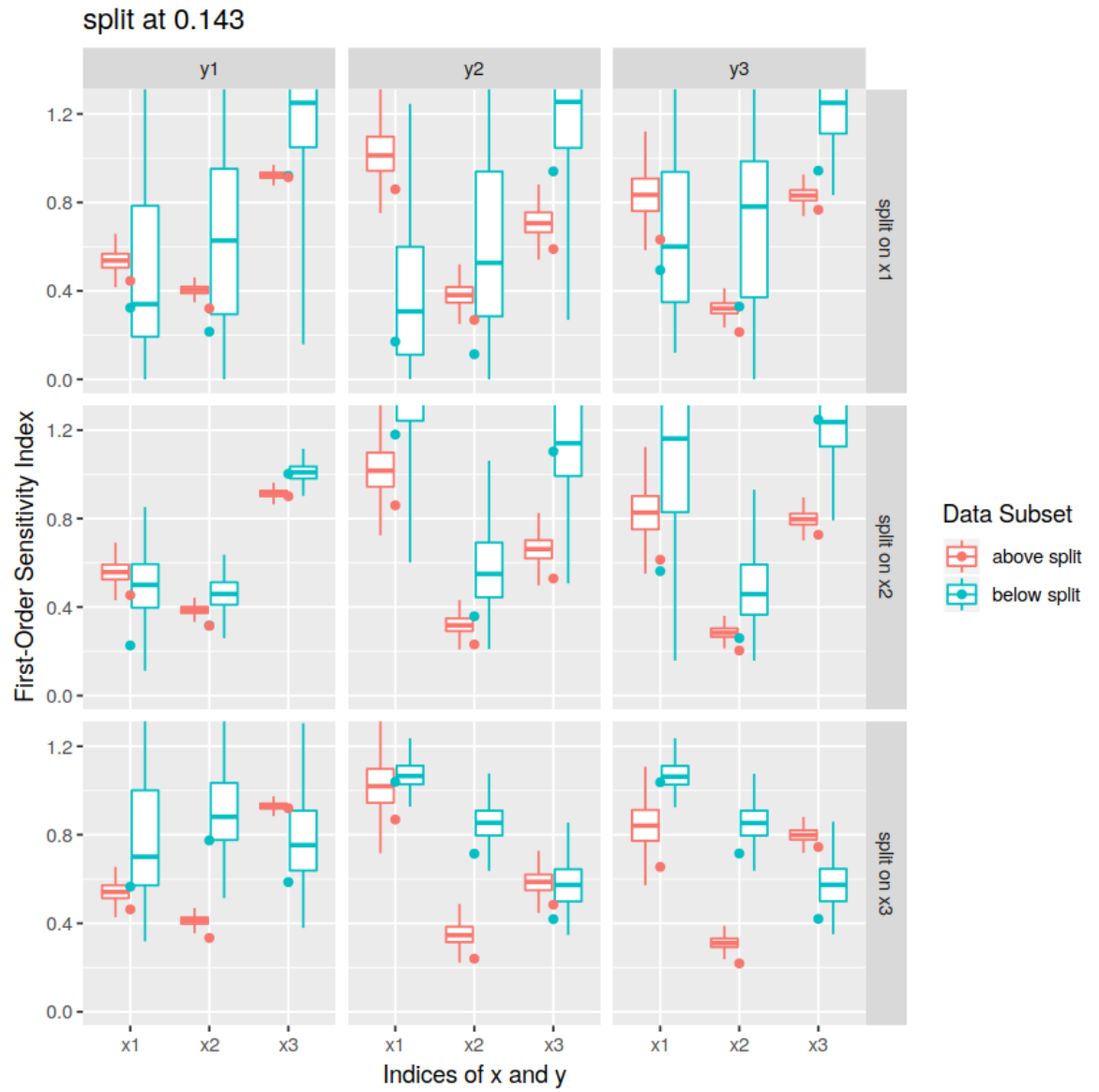
324324 7

```

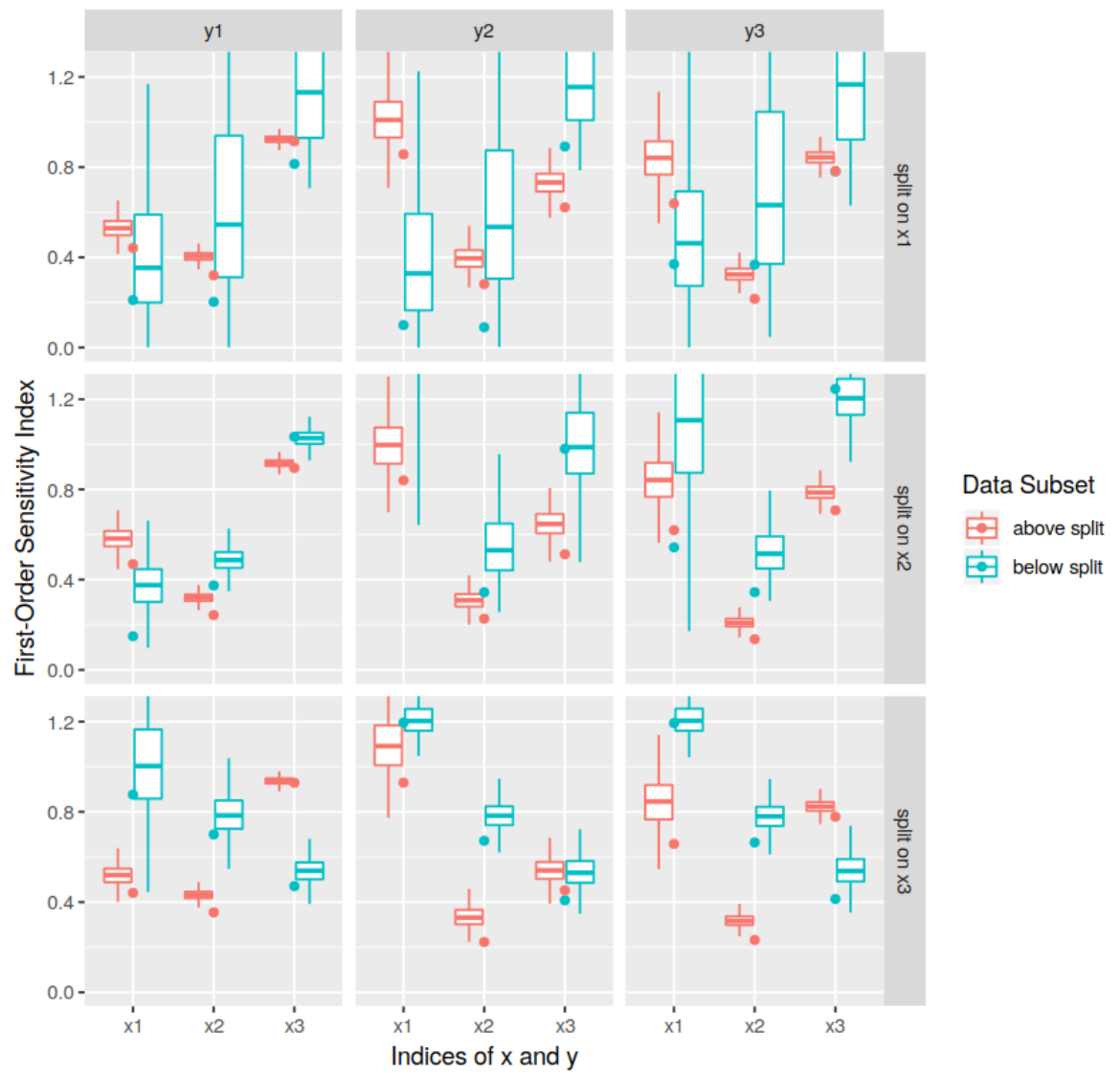
In [78]: for (u1 in unique(us$u)) {
  g <- ggplot(us.x[u1 == u & n != 0], aes(x=i, y=s.x, color=s)) +
    geom_boxplot(outlier.shape=NA) +
    coord_cartesian(ylim = c(0, 1.25)) +
    xlab("Indices of x and y") +
    ylab("First-Order Sensitivity Index") +
    guides(color=guide_legend(title="Data Subset")) +
    geom_point(data=us.x[u1 == u & n == 0], aes(x=i, y=s.x, color
=s)) +
    facet_grid(k ~ j) +
    ggtitle(paste("split at", u1))
  print(g)
}

```

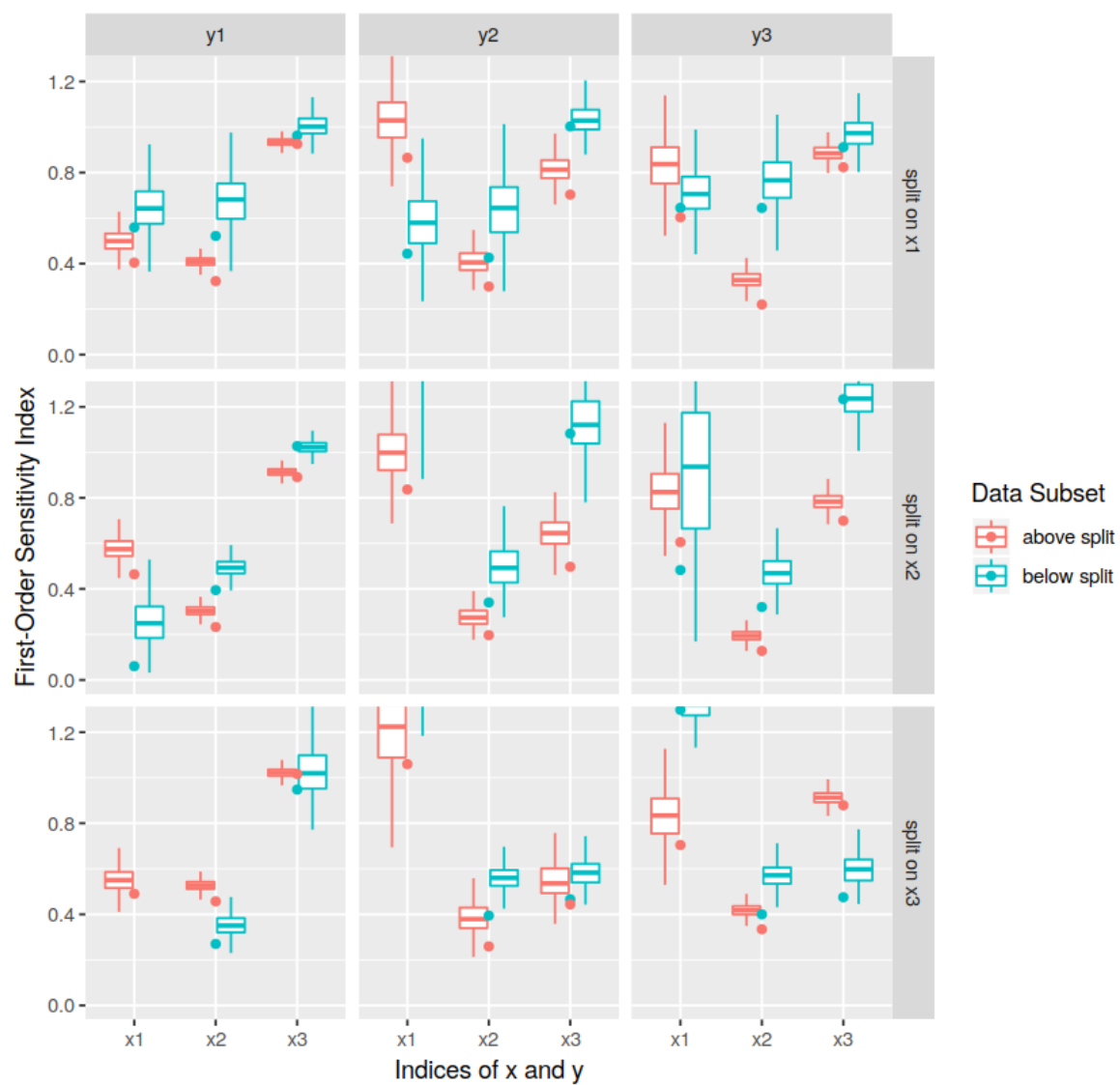
Warning message:  
"Removed 268 rows containing non-finite values (stat\_boxplot)."  
Warning message:  
"Removed 15 rows containing non-finite values (stat\_boxplot)."



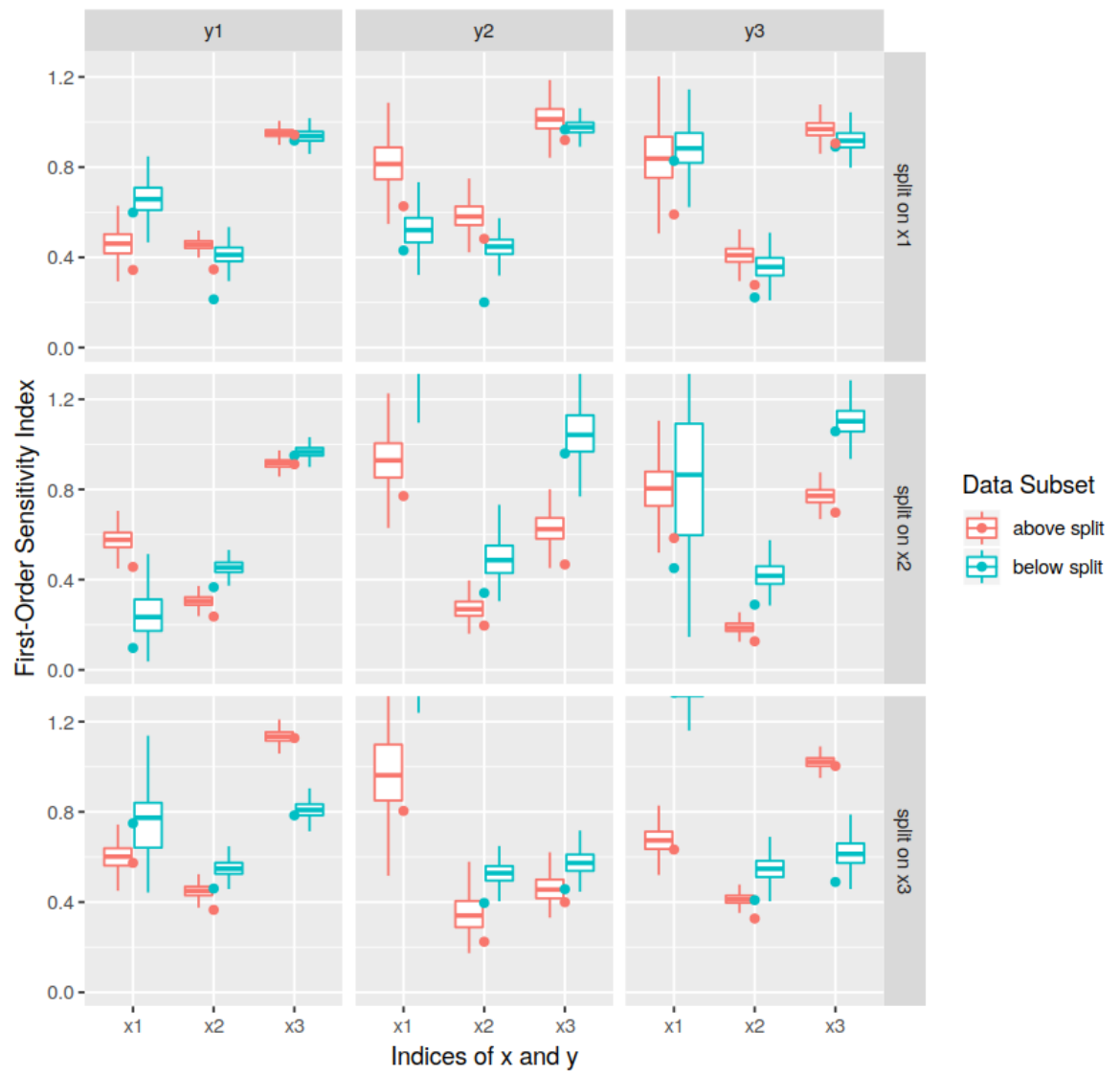
split at 0.286



split at 0.429

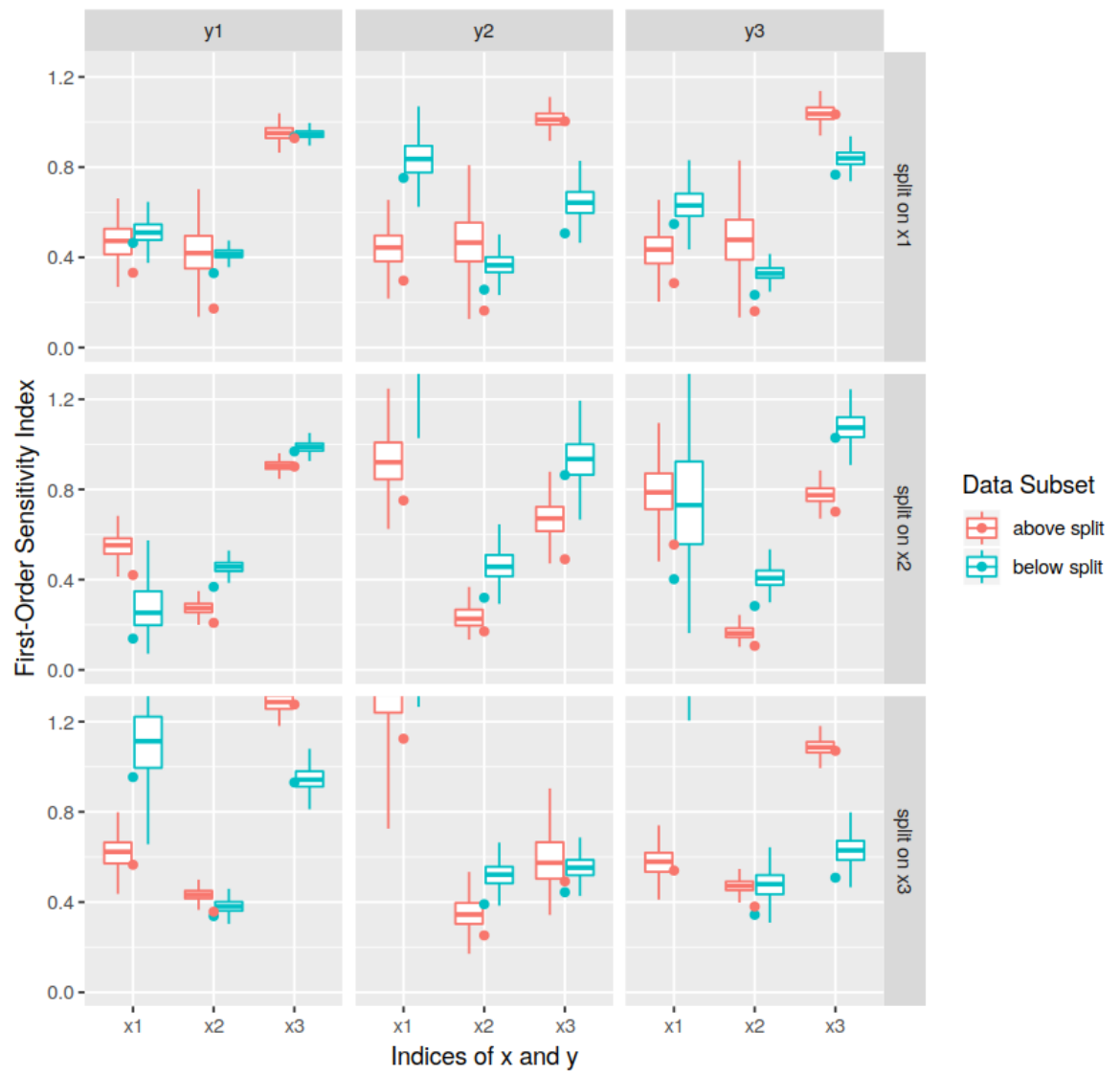


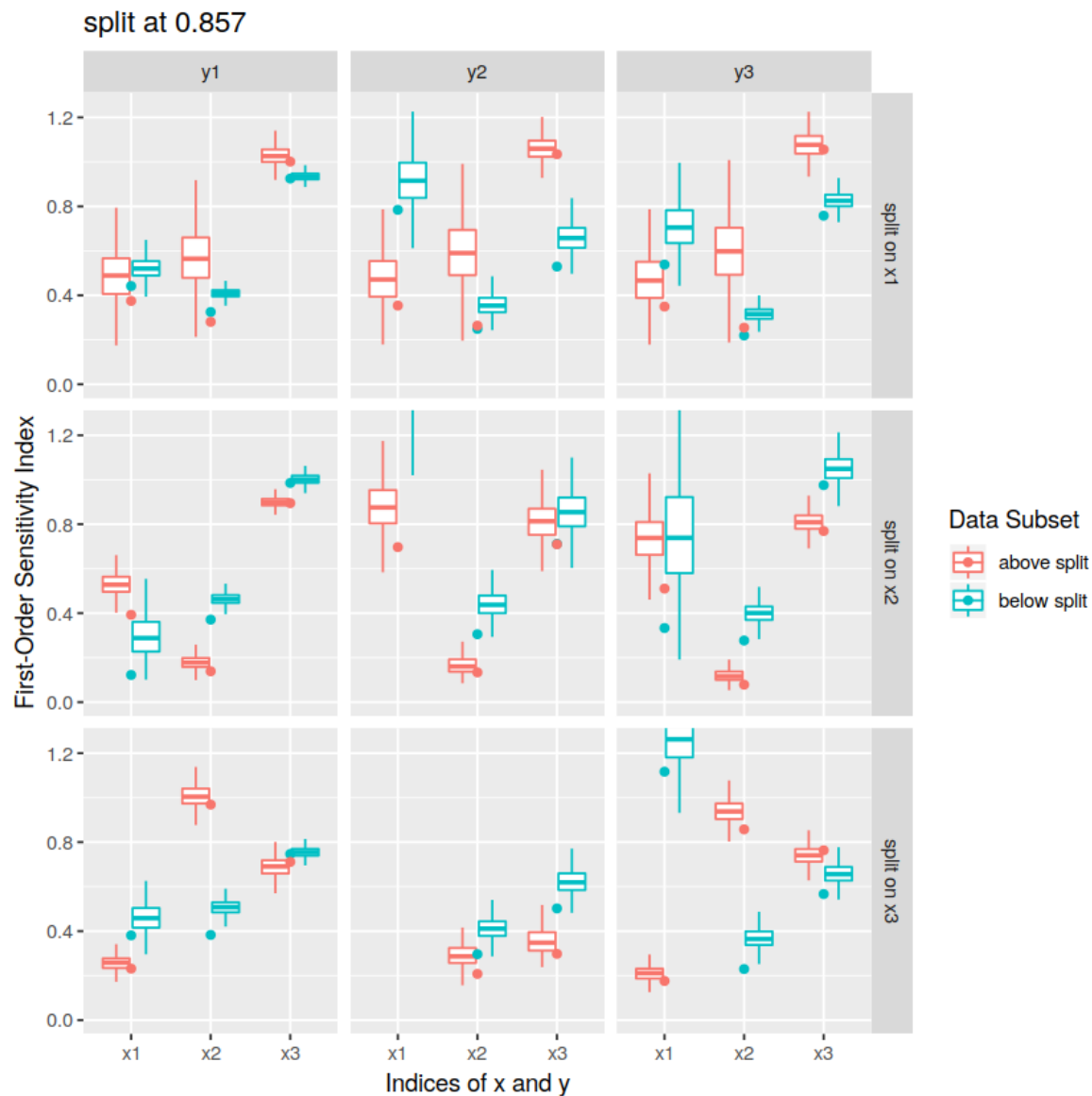
split at 0.571





split at 0.714





The solid dots show the unbootstrapped estimates.

Splitting  $x_1$ ,  $x_2$ , or  $x_3$  at  $1/7$  doesn't yield significant differences in the sensitivity indices on either side of the split. By the time the split reaches beyond  $2/7$ , significant differences appear. For example, that of  $y_3$  on  $x_1$  when split into the insignificant  $x_3 > 6/7$  vs the significant  $x_3 < 6/7$ .

**Instead of computing the sensitivity indices for each split, just test whether the probability densities on either side of the split are different.**

There doesn't seem to be a multivariate KS test available in R, so instead use `npdeneqtest` from the `np` package.

```

In [41]: tus <- us[, .(1), by=.(k, u)]
         une <- NULL
         for (row in 1:nrow(tus)) {
           z <- merge(tus[row], uxys.wide, by=c("k", "u"))
           une <- rbind(
             une,
             cbind(
               tus[row, .(k, u)],
               Tn=npdeneqtest(
                 z[s == "below split", .(y1, y2, y3)],
                 z[s == "above split", .(y1, y2, y3)],
                 boot.num=250
               )$Tn
             )
           )
         }
         une %>% dim

```

Bootstrap replication 1/250...	Bootstrap
replication 2/250...	Bootstrap replicati
on 3/250...	Bootstrap replication 4/25
0...	Bootstrap replication 5/250...
Bootstrap replication 6/250...	Bootstrap
replication 7/250...	Bootstrap replicati
on 8/250...	Bootstrap replication 9/25
0...	Bootstrap replication 10/250...
Bootstrap replication 11/250...	Bootstr
ap replication 12/250...	Bootstrap repl
ication 13/250...	Bootstrap replication
14/250...	Bootstrap replication 15/25
0...	Bootstrap replication 16/250...
Bootstrap replication 17/250...	Bootstr
ap replication 18/250...	Bootstrap repl
ication 19/250...	Bootstrap replication
20/250...	Bootstrap replication 21/25
0...	Bootstrap replication 22/250...
Bootstrap replication 23/250...	Bootstr
ap replication 24/250...	Bootstrap repl
ication 25/250...	Bootstrap replication
26/250...	Bootstrap replication 27/25
0...	Bootstrap replication 28/250...
Bootstrap replication 29/250...	Bootstr
ap replication 30/250...	Bootstrap repl
ication 31/250...	Bootstrap replication
32/250...	Bootstrap replication 33/25
0...	Bootstrap replication 34/250...
Bootstrap replication 35/250...	Bootstr
ap replication 36/250...	Bootstrap repl
ication 37/250...	Bootstrap replication
38/250...	Bootstrap replication 39/25
0...	Bootstrap replication 40/250...
Bootstrap replication 41/250...	Bootstr
ap replication 42/250...	Bootstrap repl
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44/250...	Bootstrap replication 45/25
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Bootstrap replication 53/250...	Bootstr
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Bootstrap replication 65/250...	Bootstr
ap replication 66/250...	Bootstrap repl
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Bootstrap replication 117/250...  
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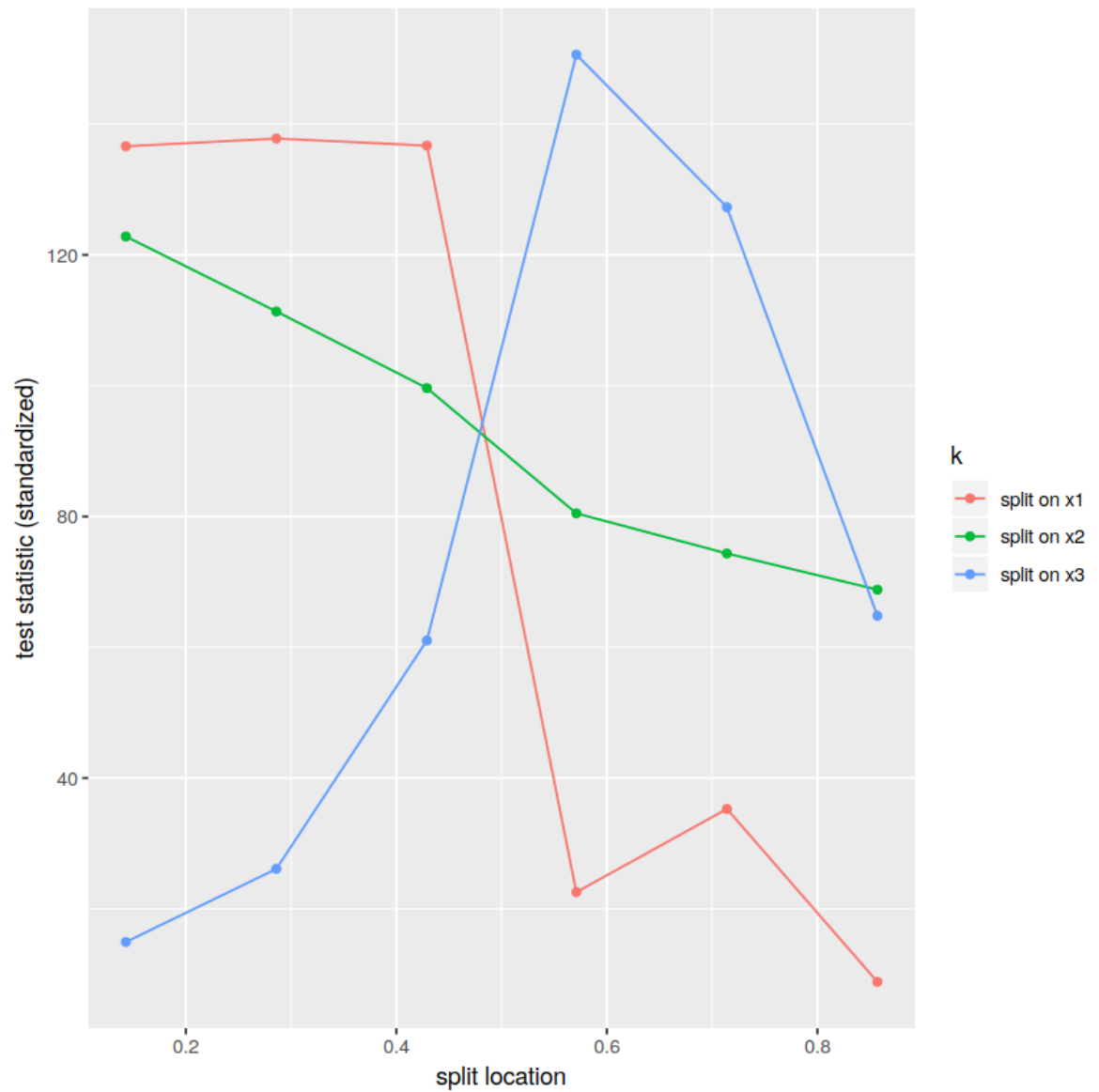
cation 136/250...  
n 137/250...  
8/250...  
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Bootstrap replication 141/250...  
trap replication 142/250...  
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Bootstrap replication 173/250...  
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 Bootstrap replication 221/250...  
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 Bootstrap replication 229/250...  
 trap replication 230/250...  
 replication 231/250...  
 cation 232/250...  
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 Bootstrap replication 237/250...  
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 Bootstrap replication 243/25  
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```
In [42]: ggplot(une, aes(x=u, y=Tn, color=k)) +
  geom_point() +
  geom_line() +
  xlab("split location") +
  ylab("test statistic (standardized)")
```



This clearly identifies the phase boundary near  $x_2 = 1/2$ .

**Repeat with bootstrap confidence intervals.**



```

In [46]: twosample_test_resample <- function(xy, n) {
  rows <- nrow(xy)
  result <- cbind(
    n=0,
    Tn=npdeneqtest(
      xy[s == "below split", .(y1, y2, y3)],
      xy[s == "above split", .(y1, y2, y3)],
      boot.num=250
    )$Tn
  )
  for (i in 1:n) {
    xy.sample <- xy[sample(1:rows, rows, replace=TRUE)]
    xy.sample[, sequence:=1:rows]
    xy.sample <- merge(tus[row], xy.sample, by=c("k", "u"))
    result <- rbind(
      result,
      cbind(
        n=i,
        Tn=npdeneqtest(
          xy.sample[s == "below split", .(y1, y2, y3)],
          xy.sample[s == "above split", .(y1, y2, y3)],
          boot.num=250
        )$Tn
      )
    )
  }
  result
}

```

```
In [47]: tus <- us[, .(1), by=.(k, u)]
         une <- NULL
         for (row in 1:nrow(tus))
           une <- rbind(
             une,
             cbind(
               tus[row, .(k, u)],
               twosample_test_resample(merge(tus[row], uxys.wide, by=c(
                 "k", "u")), 15)
             )
           )
         une %>% dim
```

Bootstrap replication 1/250...	Bootstrap
replication 2/250...	Bootstrap replicati
on 3/250...	Bootstrap replication 4/25
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Bootstrap replication 6/250...	Bootstrap
replication 7/250...	Bootstrap replicati
on 8/250...	Bootstrap replication 9/25
0...	Bootstrap replication 10/250...
Bootstrap replication 11/250...	Bootstr
ap replication 12/250...	Bootstrap repl
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14/250...	Bootstrap replication 15/25
0...	Bootstrap replication 16/250...
Bootstrap replication 17/250...	Bootstr
ap replication 18/250...	Bootstrap repl
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Bootstrap replication 157/250...  
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Bootstrap replication 165/250...  
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Warning message in EssDee(dcon):

"variable 1 appears to be constantvariable 2 appears to be constantvariable 3 appears to be constant"

Warning message in min(x[x > 0]):  
 "no non-missing arguments to min; returning Inf"

Warning message in min(x[x > 0]):  
 "no non-missing arguments to min; returning Inf"

Warning message in min(x[x > 0]):  
 "no non-missing arguments to min; returning Inf"

```
Error in system.time(myout <- .C("np_density_bw", as.double(duno), a
s.double(dord), : NA/NaN/Inf in foreign function call (arg 4)
```

Traceback:

```
1. rbind(une, cbind(tus[row, .(k, u)], twosample_test_resample(merge
(tus[row],
.      uxys.wide, by = c("k", "u")), 15)))
2. cbind(tus[row, .(k, u)], twosample_test_resample(merge(tus[row],
.      uxys.wide, by = c("k", "u")), 15))
3. twosample_test_resample(merge(tus[row], uxys.wide, by = c("k",
.      "u")), 15)
4. rbind(result, cbind(n = i, Tn = npdeneqtest(xy.sample[s == "below
split",
.      .(y1, y2, y3)], xy.sample[s == "above split", .(y1, y2, y3)],
.      boot.num = 250)$Tn)) # at line 15-25 of file <text>
5. cbind(n = i, Tn = npdeneqtest(xy.sample[s == "below split", .(y1,
.      y2, y3)], xy.sample[s == "above split", .(y1, y2, y3)], boot.n
um = 250)$Tn) # at line 15-25 of file <text>
6. npdeneqtest(xy.sample[s == "below split", .(y1, y2, y3)], xy.sampl
e[s ==
.      "above split", .(y1, y2, y3)], boot.num = 250) # at line 15-
25 of file <text>
7. npudensbw(dat = x, ...)
8. npudensbw.NULL(dat = x, ...)
9. npudensbw.default(dat = dat, bws = bws, ...)
10. eval(parse(text = paste("npudensbw.bandwidth(dat=dat, bws=tbw",
.      ifelse(any.m, ",", ""), paste(mc.names[m], ifelse(any.m,
.      "=", ""), mc.names[m], collapse = ", ", "))))))
11. eval(parse(text = paste("npudensbw.bandwidth(dat=dat, bws=tbw",
.      ifelse(any.m, ",", ""), paste(mc.names[m], ifelse(any.m,
.      "=", ""), mc.names[m], collapse = ", ", "))))))
12. npudensbw.bandwidth(dat = dat, bws = tbw)
13. system.time(myout <- .C("np_density_bw", as.double(duno), as.doub
le(dord),
.      as.double(dcon), as.double(mysd), as.integer(myopti), as.doub
le(myoptd),
.      bw = c(bws$bw[bws$icon], bws$bw[bws$iuno], bws$bw[bws$iord]),
.      fval = double(2), fval.history = double(max(1, nmulti)),
.      timing = double(1), PACKAGE = "np")[c("bw", "fval", "fval.his
tory",
.      "timing")]))
```

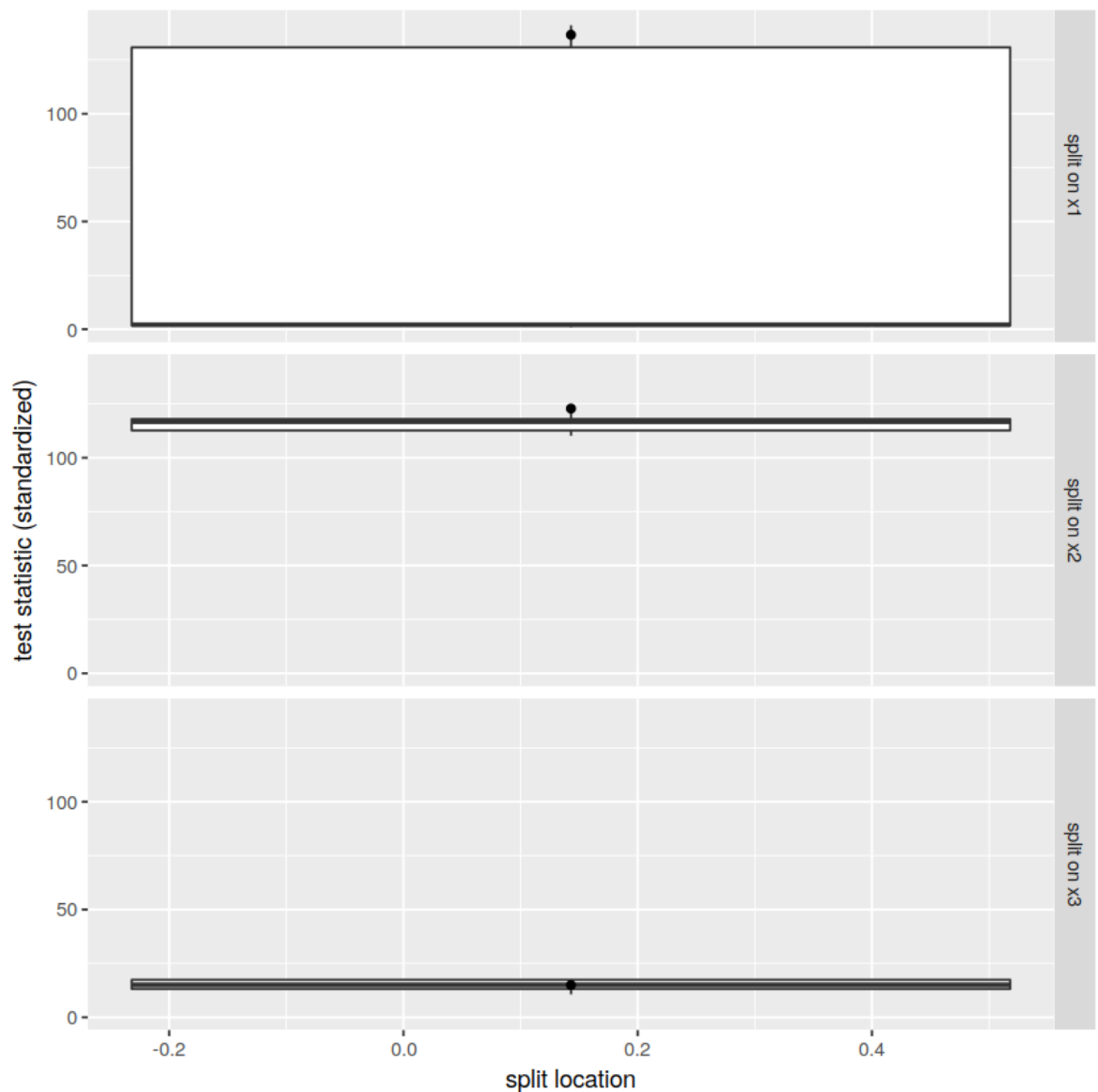
Timing stopped at: 0.005 0 0.005

**FIXME:** Look into why this failed.

```
In [80]: une %>% summary
```

	k	u	n	Tn
Length:	48	Min. :0.143	Min. : 0.00	Min. : 1.034
Class :	character	1st Qu.:0.143	1st Qu.: 3.75	1st Qu.: 11.409
Mode :	character	Median :0.143	Median : 7.50	Median : 30.714
		Mean :0.143	Mean : 7.50	Mean : 61.100
		3rd Qu.:0.143	3rd Qu.:11.25	3rd Qu.:117.182
		Max. :0.143	Max. :15.00	Max. :141.062

```
In [81]: ggplot(une[n != 0], aes(x=u, y=Tn)) +  
  geom_boxplot(outlier.shape=NA) +  
  geom_point(data=une[n == 0], aes(x=u, y=Tn)) +  
  facet_grid(k ~ .) +  
  guides(color=guide_legend(title="Data Subset")) +  
  xlab("split location") +  
  ylab("test statistic (standardized)")
```



```
In [ ]:
```