

Compare

Alan n. Inglis

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Load libraries:

```
library(dbarts) # for model
library(bartMan) # for visualizations
library(vivid) # for agnostic visualizations
library(ggplot2) # for visualizations
```

Read in and setup data:

```
# Create some data
f <- function(x) {
  10 * sin(pi * x[, 1] * x[, 2]) + 20 * (x[, 3] - 0.5)^2 +
  10 * x[, 4] + 5 * x[, 5]
}

set.seed(1701)
sigma <- 1.0
n <- 250
x <- matrix(runif(n * 10), n, 10)
colnames(x) <- paste0("x", 1:10)
Ey <- f(x)
y <- rnorm(n, Ey, sigma)
fData <- as.data.frame(cbind(x, y))

x <- fData[, 1:10]
y <- fData$y
```

Build models

```
set.seed(1701)
dB20 <- bart(x.train = x,
             y.train = y,
             ntree = 20,
             keeptrees = TRUE,
             nskip = 100,
             ndpost = 1000
)
```

```

##
## Running BART with numeric y
##
## number of trees: 20
## number of chains: 1, number of threads 1
## tree thinning rate: 1
## Prior:
## k prior fixed to 2.000000
## degrees of freedom in sigma prior: 3.000000
## quantile in sigma prior: 0.900000
## scale in sigma prior: 0.002377
## power and base for tree prior: 2.000000 0.950000
## use quantiles for rule cut points: false
## proposal probabilities: birth/death 0.50, swap 0.10, change 0.40; birth 0.50
## data:
## number of training observations: 250
## number of test observations: 0
## number of explanatory variables: 10
## init sigma: 2.766176, curr sigma: 2.766176
##
## Cutoff rules c in  $x \leq c$  vs  $x > c$ 
## Number of cutoffs: (var: number of possible c):
## (1: 100) (2: 100) (3: 100) (4: 100) (5: 100)
## (6: 100) (7: 100) (8: 100) (9: 100) (10: 100)
##
## Running mcmc loop:
## iteration: 100 (of 1000)
## iteration: 200 (of 1000)
## iteration: 300 (of 1000)
## iteration: 400 (of 1000)
## iteration: 500 (of 1000)
## iteration: 600 (of 1000)
## iteration: 700 (of 1000)
## iteration: 800 (of 1000)
## iteration: 900 (of 1000)
## iteration: 1000 (of 1000)
## total seconds in loop: 0.123965
##
## Tree sizes, last iteration:
## [1] 3 2 3 2 3 4 6 2 4 3 2 2 2 3 2 4 2 3
## 4 4
##
## Variable Usage, last iteration (var:count):
## (1: 7) (2: 8) (3: 8) (4: 8) (5: 5)
## (6: 1) (7: 0) (8: 2) (9: 0) (10: 1)
##
## DONE BART

set.seed(1701)
dB100 <- bart(x.train = x,
              y.train = y,
              ntree = 100,
              kepttrees = TRUE,
              nskip = 100,

```

```

)                                ndpost = 1000

##
## Running BART with numeric y
##
## number of trees: 100
## number of chains: 1, number of threads 1
## tree thinning rate: 1
## Prior:
## k prior fixed to 2.000000
## degrees of freedom in sigma prior: 3.000000
## quantile in sigma prior: 0.900000
## scale in sigma prior: 0.002377
## power and base for tree prior: 2.000000 0.950000
## use quantiles for rule cut points: false
## proposal probabilities: birth/death 0.50, swap 0.10, change 0.40; birth 0.50
## data:
## number of training observations: 250
## number of test observations: 0
## number of explanatory variables: 10
## init sigma: 2.766176, curr sigma: 2.766176
##
## Cutoff rules c in x<=c vs x>c
## Number of cutoffs: (var: number of possible c):
## (1: 100) (2: 100) (3: 100) (4: 100) (5: 100)
## (6: 100) (7: 100) (8: 100) (9: 100) (10: 100)
##
## Running mcmc loop:
## iteration: 100 (of 1000)
## iteration: 200 (of 1000)
## iteration: 300 (of 1000)
## iteration: 400 (of 1000)
## iteration: 500 (of 1000)
## iteration: 600 (of 1000)
## iteration: 700 (of 1000)
## iteration: 800 (of 1000)
## iteration: 900 (of 1000)
## iteration: 1000 (of 1000)
## total seconds in loop: 0.603402
##
## Tree sizes, last iteration:
## [1] 2 2 2 3 2 3 4 2 2 2 2 3 3 2 2 2 2
## 7 2 3 2 2 2 2 2 3 2 2 4 3 2 1 2 2 4 3
## 3 2 2 3 3 1 3 2 2 2 4 2 6 2 2 3 4 2 4 3
## 3 3 2 1 3 2 2 4 3 2 2 2 1 2 2 3 2 2 3 2
## 2 3 2 2 4 2 3 4 3 2 5 3 2 2 2 3 2 3 2 4
## 3 2
##
## Variable Usage, last iteration (var:count):
## (1: 26) (2: 20) (3: 22) (4: 18) (5: 11)
## (6: 12) (7: 12) (8: 10) (9: 14) (10: 10)
##
## DONE BART

```

```

set.seed(1701)
dB200 <- bart(x.train = x,
              y.train = y,
              ntree = 200,
              keeptrees = TRUE,
              nskip = 100,
              ndpost = 1000
            )

##
## Running BART with numeric y
##
## number of trees: 200
## number of chains: 1, number of threads 1
## tree thinning rate: 1
## Prior:
## k prior fixed to 2.000000
## degrees of freedom in sigma prior: 3.000000
## quantile in sigma prior: 0.900000
## scale in sigma prior: 0.002377
## power and base for tree prior: 2.000000 0.950000
## use quantiles for rule cut points: false
## proposal probabilities: birth/death 0.50, swap 0.10, change 0.40; birth 0.50
## data:
## number of training observations: 250
## number of test observations: 0
## number of explanatory variables: 10
## init sigma: 2.766176, curr sigma: 2.766176
##
## Cutoff rules c in x<=c vs x>c
## Number of cutoffs: (var: number of possible c):
## (1: 100) (2: 100) (3: 100) (4: 100) (5: 100)
## (6: 100) (7: 100) (8: 100) (9: 100) (10: 100)
##
## Running mcmc loop:
## iteration: 100 (of 1000)
## iteration: 200 (of 1000)
## iteration: 300 (of 1000)
## iteration: 400 (of 1000)
## iteration: 500 (of 1000)
## iteration: 600 (of 1000)
## iteration: 700 (of 1000)
## iteration: 800 (of 1000)
## iteration: 900 (of 1000)
## iteration: 1000 (of 1000)
## total seconds in loop: 1.142502
##
## Tree sizes, last iteration:
## [1] 2 3 2 2 3 2 2 2 2 2 4 2 3 1 2 2 2 3
## 4 3 4 4 2 2 1 2 4 2 2 2 3 2 2 2 3 2 2 2
## 2 2 3 3 4 6 3 2 2 1 3 2 2 3 2 2 2 2 2 1
## 2 1 2 3 2 2 2 3 2 2 2 1 2 3 2 3 2 3 1 3
## 2 2 2 2 3 2 2 3 2 3 2 3 2 2 2 1 2 1 2 2
## 3 2 2 1 2 3 4 4 2 2 3 2 2 5 2 2 2 2 2 2

```

```
## 3 2 2 3 2 2 3 4 2 2 1 4 3 2 2 2 2 3 4 3
## 2 2 2 2 3 4 2 3 2 2 2 1 2 2 5 2 4 2 2 3
## 3 2 4 2 2 2 2 3 2 2 3 2 3 2 3 2 3 2 3 2
## 2 2 3 3 2 2 3 2 2 3 2 3 1 3 3 2 2 2 3 2
## 2 2
##
## Variable Usage, last iteration (var:count):
## (1: 38) (2: 35) (3: 27) (4: 25) (5: 22)
## (6: 28) (7: 22) (8: 21) (9: 22) (10: 34)
##
## DONE BART
```

Create dataframe of trees

```
dbT20 <- extractTreeData(model = dB20, data = fData)
dbT100 <- extractTreeData(model = dB100, data = fData)
dbT200 <- extractTreeData(model = dB200, data = fData)
```

Figure 11:

20 trees

```
# Set var order
newOrder <- c("x1","x2","x3","x4","x5", 'x6', 'x7', 'x8', 'x9', 'x10')

# vsup matrix
vsupMat <- viviBartMatrix(dbT20, type = 'vsup', metric = 'propMean',
                          metricError = "CV", reorder = F)

# reorder
vsupMat$actualMatrix <- vsupMat$actualMatrix[newOrder,newOrder]
vsupMat$uncertaintyMatrix <- vsupMat$uncertaintyMatrix[newOrder,newOrder]

colors <- scales::colour_ramp(
  colors = c(blue = '#FFFFCC', red = '#800026')
)((0:7)/7)

newCols <- RColorBrewer::brewer.pal(9, 'GnBu')
colors2 <- newCols[-1]

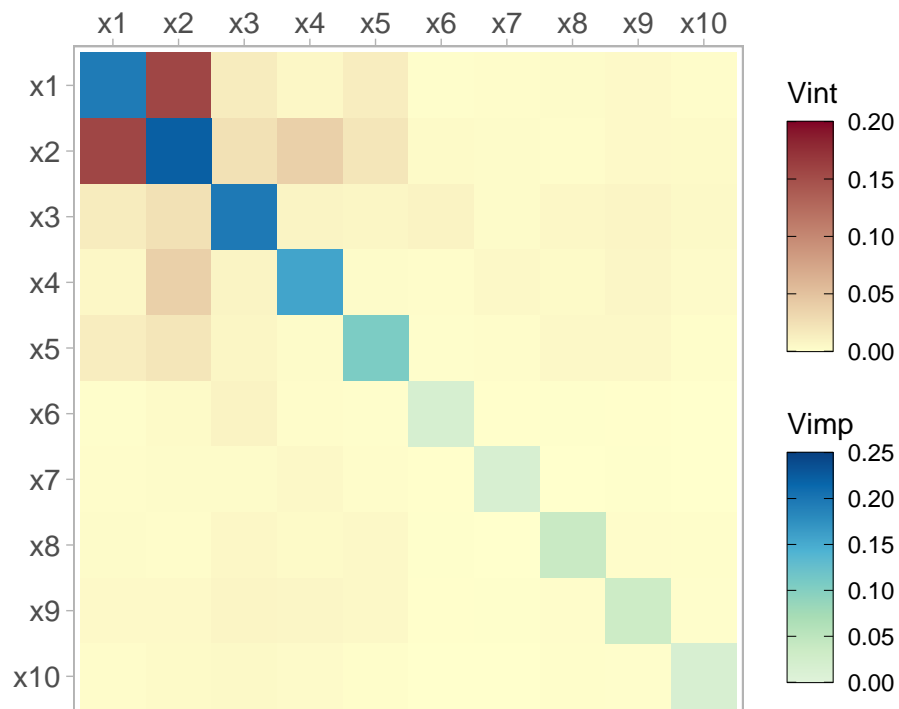
# -----

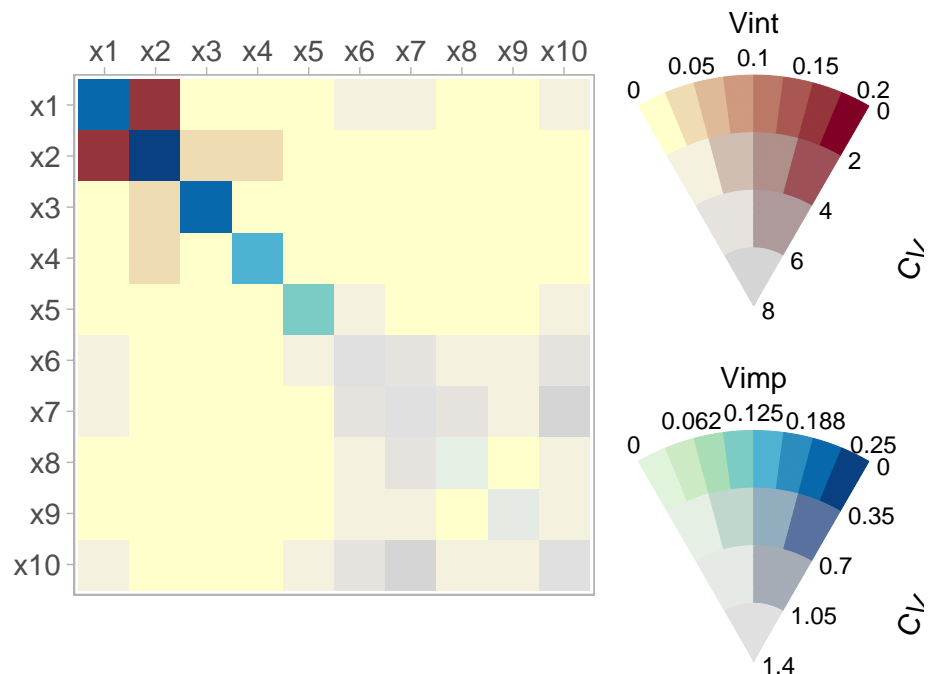
vivid::viviHeatmap(vsupMat$actualMatrix,
                   intPal = colors,
                   impPal = colors2)
```

```

# plot vsup
viviBartPlot(vsupMat,
             intPal = colors,
             impPal = colors2,
             max_desat = 1,
             pow_desat = 0.6,
             max_light = 0.6,
             pow_light = 1,
             label = "CV"
             ) +
  theme(axis.text.x = element_text(hjust = 0.5))

```



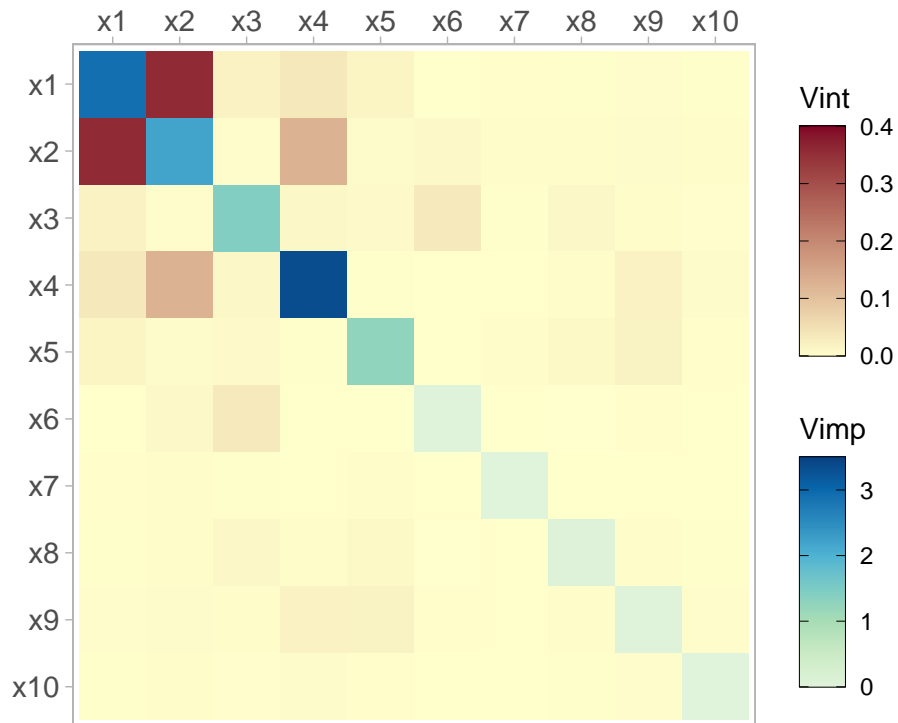


vivid for 20 trees

```
response = 'y'
data = fData
# create predict function
responseIdx <- which(colnames(data) == response)
pFun <- function(fit, data, prob=TRUE) apply(predict(fit, data[, -responseIdx]), 2, mean)

# run vivid
set.seed(1701)
mat <- vivid::vivi(fit = dB20,
                  data = fData,
                  response = 'y',
                  reorder = F,
                  gridSize = 10,
                  nmax = 500,
                  normalized = FALSE,
                  class = 1,
                  predictFun = pFun)

mat <- mat[newOrder, newOrder]
viviHeatmap(mat,
            intPal = colors,
            impPal = colors2)
```



100 trees

```
# Set var order
newOrder <- c("x1","x2","x3","x4","x5", 'x6', 'x7', 'x8', 'x9', 'x10')

# vsup matrix
vsupMat <- viviBartMatrix(dbT100, type = 'vsup',
                          metric = 'propMean',
                          metricError = "CV", reorder = F)

# reorder
vsupMat$actualMatrix <- vsupMat$actualMatrix[newOrder,newOrder]
vsupMat$uncertaintyMatrix <- vsupMat$uncertaintyMatrix[newOrder,newOrder]

colors <- scales::colour_ramp(
  colors = c(blue = '#FFFFCC', red = '#800026')
)((0:7)/7)

newCols <- RColorBrewer::brewer.pal(9, 'GnBu')
colors2 <- newCols[-1]

# -----

vivid::viviHeatmap(vsupMat$actualMatrix,
                   impLims = c(0, 0.16),
                   intPal = colors,
```

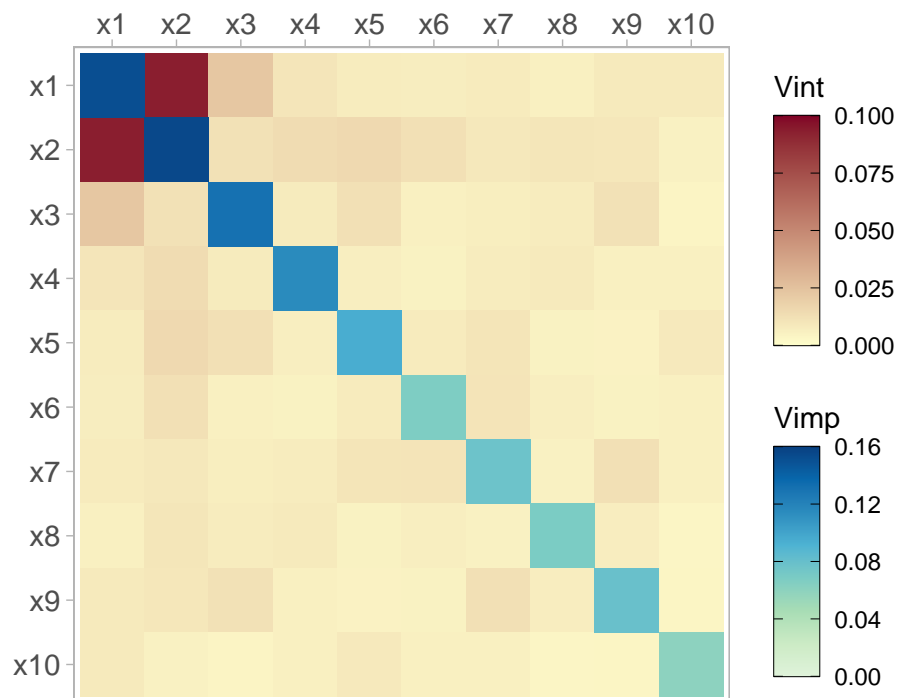


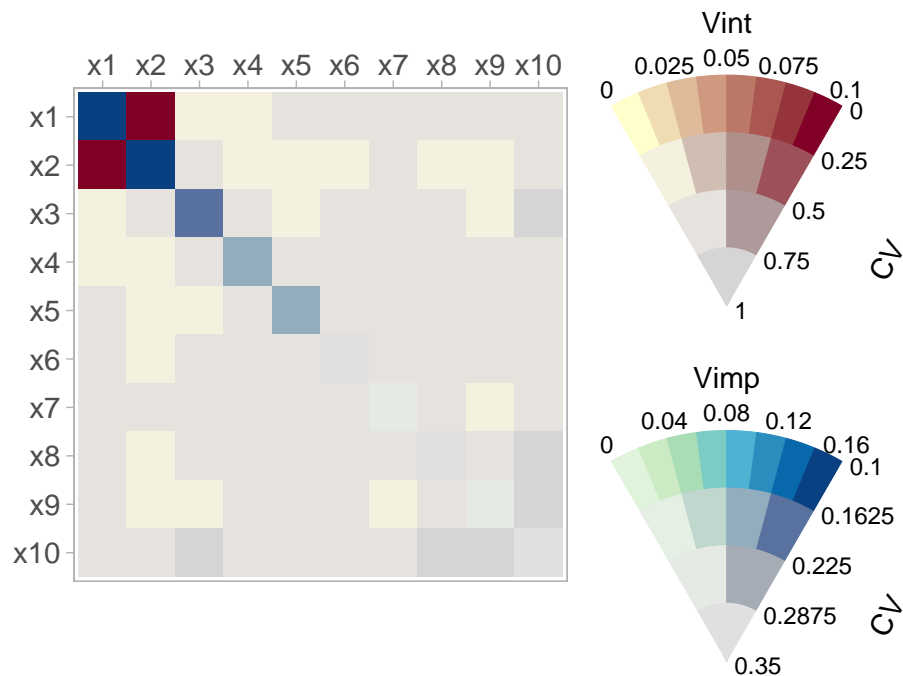
```

    impPal = colors2)

# plot vsup
viviBartPlot(vsupMat,
  intPal = colors,
  impPal = colors2,
  impLims = c(0, 0.16),
  max_desat = 1,
  pow_desat = 0.6,
  max_light = 0.6,
  pow_light = 1,
  label = "CV"
) +
  theme(axis.text.x = element_text(hjust = 0.5))

```



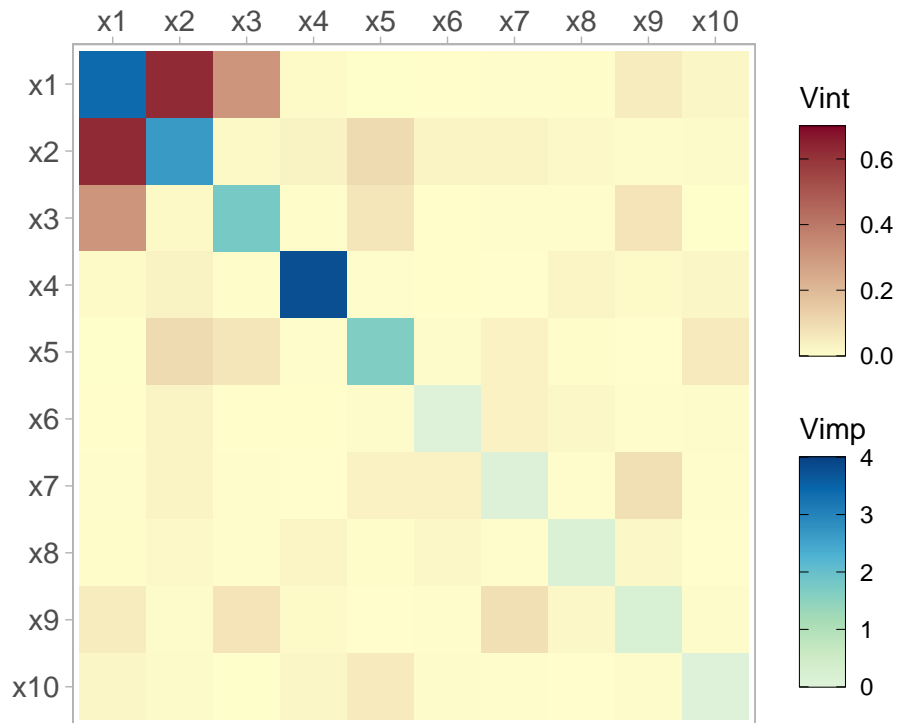


vivid for 100 trees

```
response = 'y'
data = fData
# create predict function
responseIdx <- which(colnames(data) == response)
pFun <- function(fit, data, prob=TRUE) apply(predict(fit, data[, -responseIdx]), 2, mean)

# run vivid
set.seed(1701)
mat <- vivid::vivi(fit = dB100,
                  data = fData,
                  response = 'y',
                  reorder = F,
                  gridSize = 10,
                  nmax = 500,
                  normalized = FALSE,
                  class = 1,
                  predictFun = pFun)

mat <- mat[newOrder, newOrder]
viviHeatmap(mat,
            intPal = colors,
            impPal = colors2)
```



200 trees

```
# Set var order
newOrder <- c("x1","x2","x3","x4","x5", 'x6', 'x7', 'x8', 'x9', 'x10')

# vsup matrix
vsupMat <- viviBartMatrix(dbT200, type = 'vsup',
                          metric = 'propMean',
                          metricError = "CV", reorder = F)

# reorder
vsupMat$actualMatrix <- vsupMat$actualMatrix[newOrder,newOrder]
vsupMat$uncertaintyMatrix <- vsupMat$uncertaintyMatrix[newOrder,newOrder]

colors <- scales::colour_ramp(
  colors = c(blue = '#FFFFCC', red = '#800026')
)((0:7)/7)

newCols <- RColorBrewer::brewer.pal(9, 'GnBu')
colors2 <- newCols[-1]

# -----

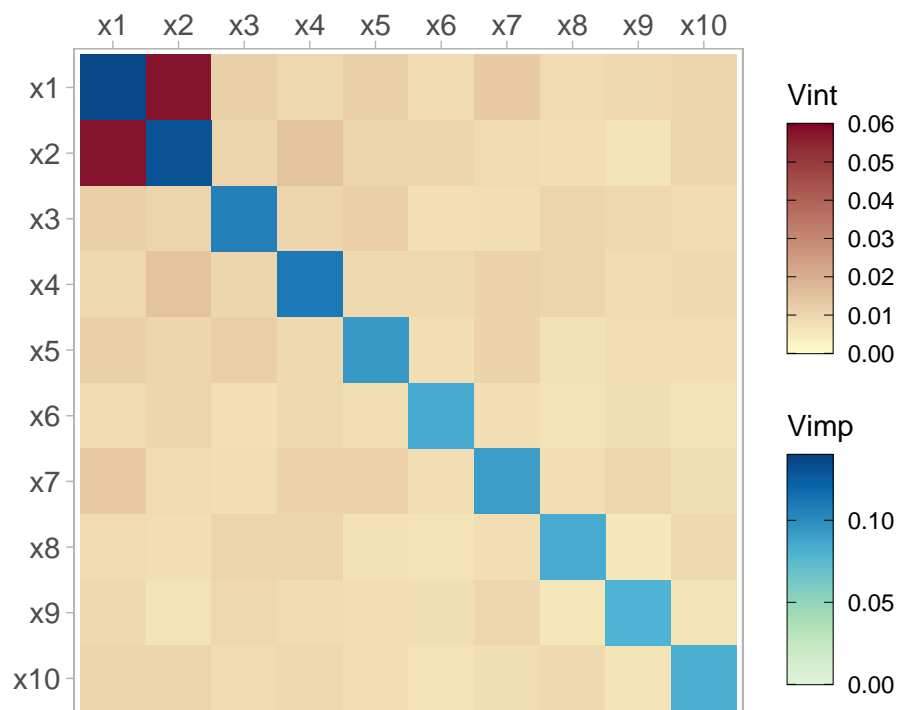
vivid::viviHeatmap(vsupMat$actualMatrix,
                   impLims = c(0, 0.14),
                   intPal = colors,
```

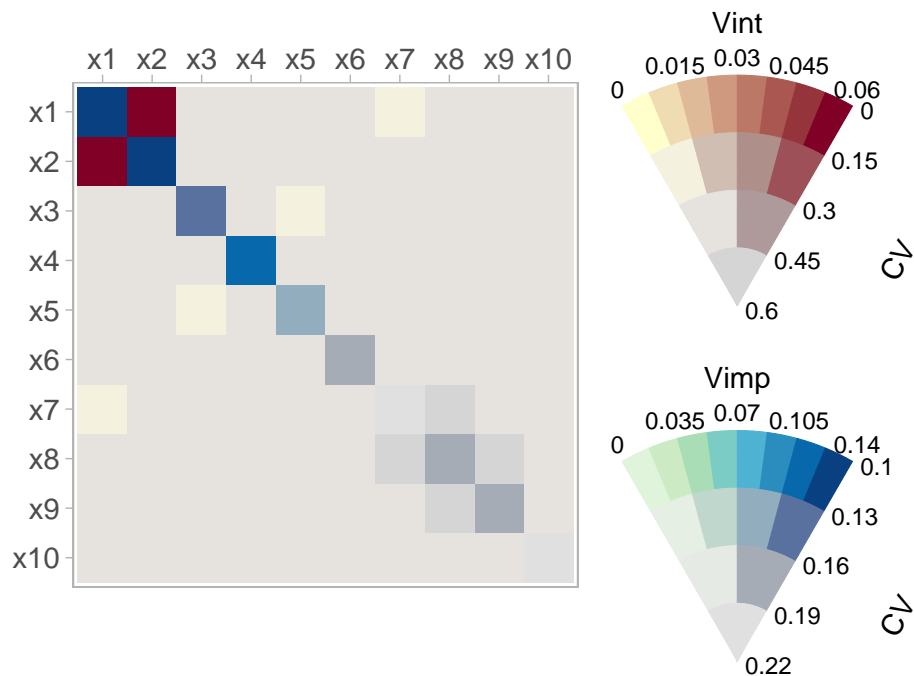
```

    impPal = colors2)

# plot vsup
viviBartPlot(vsupMat,
  intPal = colors,
  impPal = colors2,
  impLims = c(0, 0.14),
  max_desat = 1,
  pow_desat = 0.6,
  max_light = 0.6,
  pow_light = 1,
  label = "CV"
) +
  theme(axis.text.x = element_text(hjust = 0.5))

```





vivid for 200 trees

```
response = 'y'
data = fData
# create predict function
responseIdx <- which(colnames(data) == response)
pFun <- function(fit, data, prob=TRUE) apply(predict(fit, data[, -responseIdx]), 2, mean)

# run vivid
set.seed(1701)
mat <- vivid::vivi(fit = dB200,
                  data = fData,
                  response = 'y',
                  reorder = F,
                  gridSize = 10,
                  nmax = 500,
                  normalized = FALSE,
                  class = 1,
                  predictFun = pFun)

mat <- mat[newOrder, newOrder]
viviHeatmap(mat,
            intPal = colors,
            impPal = colors2)
```

