# Exploratory Plot

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# Getting data

```
allData <- NULL
for(n in c(100,500,1000,5000)){
   allData <- rbind(allData, readRDS(paste0("goodHeteroxyBig_n",n,"_2022.RDS"))$results)
}
library(ggplot2)

## Warning: package 'ggplot2' was built under R version 4.1.1
library(reshape)

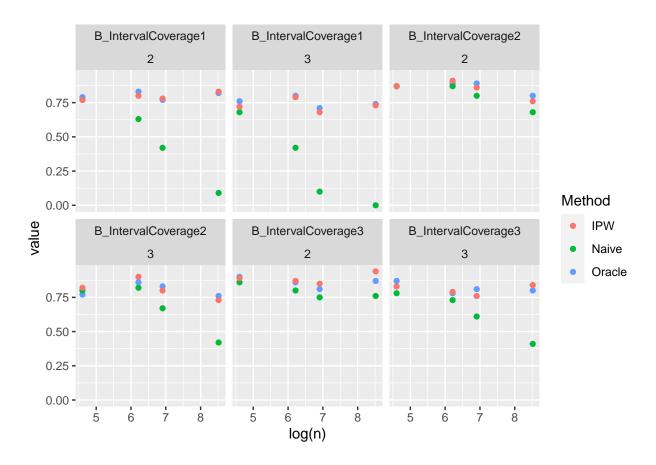
## Warning: package 'reshape' was built under R version 4.1.3
allData$Method <- paste(allData$Method,allData$x1Type)
allData <- allData[,-ncol(allData)]</pre>
```

# $X_1 \sim U[0, 1]$

### Coverage

```
diffMissInter <- allData
missPlotData <- melt(diffMissInter, id.vars=c("Method","tau","m","rho","n","ycoef"))

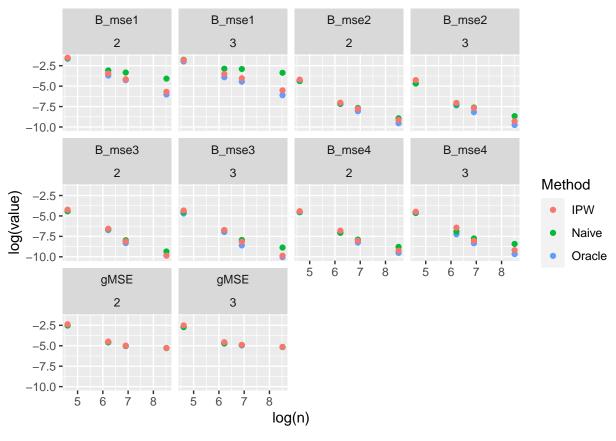
cData <- subset(missPlotData, variable %in% c("B_IntervalCoverage1","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2","B_IntervalCoverage2</pre>
```



## MSE Results

```
mseData <- subset(missPlotData, variable %in% c("B_mse1","B_mse2","B_mse3","B_mse4","gMSE"))
mseData$variable <- paste(mseData$variable)

qplot(x=log(n),y=log(value), data=mseData, color=Method) + facet_wrap(~variable*m)</pre>
```



```
mseTable <- allData[,c(1:3,5,11:15)]
for(i in 5:9){
  mseTable[,i] <- round(mseTable[,i],5)
}
mseTable</pre>
```

```
##
       Method tau m
                       n B_mse1 B_mse2 B_mse3 B_mse4
                     100 0.19241 0.01246 0.01194 0.01051 0.07970
## 1
      Oracle
              0.7 2
## 2
       Naive
              0.7 2
                     100 0.20522 0.01265 0.01207 0.01086 0.08224
## 3
         IPW
              0.7 2
                     100 0.21915 0.01496 0.01448 0.01215 0.09336
      Oracle
              0.7 3
                     100 0.13753 0.00937 0.00908 0.00954 0.07079
## 5
       Naive
              0.7 3
                     100 0.16852 0.00926 0.01076 0.00978 0.06562
## 6
         IPW
              0.7
                  3
                     100 0.16143 0.01405 0.01313 0.01130 0.08144
              0.7 2
                     500 0.02462 0.00075 0.00119 0.00086 0.01020
## 7
      Oracle
## 8
       Naive
              0.7 2
                     500 0.04594 0.00078 0.00136 0.00086 0.01019
                     500 0.03177 0.00089 0.00141 0.00111 0.01126
## 9
         IPW
              0.7 2
## 10 Oracle
                     500 0.02000 0.00064 0.00095 0.00071 0.00885
              0.7 3
       Naive
## 11
              0.7 3
                     500 0.05694 0.00071 0.00117 0.00100 0.00896
                     500 0.02984 0.00086 0.00121 0.00161 0.01040
## 12
         IPW
              0.7 3
## 13 Oracle
              0.7 2 1000 0.01401 0.00031 0.00024 0.00026 0.00659
## 14
       Naive
              0.7 2 1000 0.03555 0.00046 0.00035 0.00037 0.00654
              0.7 2 1000 0.01567 0.00042 0.00030 0.00033 0.00687
## 15
              0.7 3 1000 0.01164 0.00028 0.00018 0.00024 0.00712
## 16 Oracle
## 17
       Naive
              0.7 3 1000 0.05492 0.00049 0.00036 0.00043 0.00724
              0.7 3 1000 0.01758 0.00046 0.00029 0.00032 0.00760
## 18
         IPW
              0.7 2 5000 0.00242 0.00007 0.00005 0.00007 0.00509
       Naive 0.7 2 5000 0.01692 0.00013 0.00009 0.00015 0.00507
## 20
```

```
IPW 0.7 2 5000 0.00336 0.00011 0.00005 0.00010 0.00513
## 22 Oracle 0.7 3 5000 0.00222 0.00006 0.00004 0.00006 0.00576
## 23 Naive 0.7 3 5000 0.03442 0.00017 0.00014 0.00022 0.00577
         IPW 0.7 3 5000 0.00401 0.00009 0.00005 0.00010 0.00585
## 24
library(xtable)
## Warning: package 'xtable' was built under R version 4.1.1
xmseHetero \leftarrow xtable(mseTable, digits=c(0,0,1,0,0,5,5,5,5,5))
print(xmseHetero, include.rownames=FALSE)
## % latex table generated in R 4.1.0 by xtable 1.8-4 package
## % Mon Dec 19 16:50:39 2022
## \begin{table}[ht]
## \centering
## \begin{tabular}{lrrrrrrr}
##
     \hline
## Method & tau & m & n & B\_mse1 & B\_mse2 & B\_mse3 & B\_mse4 & gMSE \\
##
## Oracle & 0.7 & 2 & 100 & 0.19241 & 0.01246 & 0.01194 & 0.01051 & 0.07970 \\
##
     Naive & 0.7 & 2 & 100 & 0.20522 & 0.01265 & 0.01207 & 0.01086 & 0.08224 \
##
     IPW & 0.7 & 2 & 100 & 0.21915 & 0.01496 & 0.01448 & 0.01215 & 0.09336 \\
     Oracle & 0.7 & 3 & 100 & 0.13753 & 0.00937 & 0.00908 & 0.00954 & 0.07079 \\
##
     Naive & 0.7 & 3 & 100 & 0.16852 & 0.00926 & 0.01076 & 0.00978 & 0.06562 \\
##
     IPW & 0.7 & 3 & 100 & 0.16143 & 0.01405 & 0.01313 & 0.01130 & 0.08144 \\
##
##
     Oracle & 0.7 & 2 & 500 & 0.02462 & 0.00075 & 0.00119 & 0.00086 & 0.01020 \\
     Naive & 0.7 & 2 & 500 & 0.04594 & 0.00078 & 0.00136 & 0.00086 & 0.01019 \\
##
     IPW & 0.7 & 2 & 500 & 0.03177 & 0.00089 & 0.00141 & 0.00111 & 0.01126 \\
##
     Oracle & 0.7 & 3 & 500 & 0.02000 & 0.00064 & 0.00095 & 0.00071 & 0.00885 \\
##
    Naive & 0.7 & 3 & 500 & 0.05694 & 0.00071 & 0.00117 & 0.00100 & 0.00896 \\
     IPW & 0.7 & 3 & 500 & 0.02984 & 0.00086 & 0.00121 & 0.00161 & 0.01040 \\
##
##
     Oracle & 0.7 & 2 & 1000 & 0.01401 & 0.00031 & 0.00024 & 0.00026 & 0.00659 \\
    Naive & 0.7 & 2 & 1000 & 0.03555 & 0.00046 & 0.00035 & 0.00037 & 0.00654 \\
##
##
     IPW & 0.7 & 2 & 1000 & 0.01567 & 0.00042 & 0.00030 & 0.00033 & 0.00687 \\
     Oracle & 0.7 & 3 & 1000 & 0.01164 & 0.00028 & 0.00018 & 0.00024 & 0.00712 \\
##
##
     Naive & 0.7 & 3 & 1000 & 0.05492 & 0.00049 & 0.00036 & 0.00043 & 0.00724 \\
     IPW & 0.7 & 3 & 1000 & 0.01758 & 0.00046 & 0.00029 & 0.00032 & 0.00760 \\
##
##
     Oracle & 0.7 & 2 & 5000 & 0.00242 & 0.00007 & 0.00005 & 0.00007 & 0.00509 \\
##
     Naive & 0.7 & 2 & 5000 & 0.01692 & 0.00013 & 0.00009 & 0.00015 & 0.00507 \\
##
     IPW & 0.7 & 2 & 5000 & 0.00336 & 0.00011 & 0.00005 & 0.00010 & 0.00513 \\
     Oracle & 0.7 & 3 & 5000 & 0.00222 & 0.00006 & 0.00004 & 0.00006 & 0.00576 \\
##
     Naive & 0.7 & 3 & 5000 & 0.03442 & 0.00017 & 0.00014 & 0.00022 & 0.00577 \
##
##
     IPW & 0.7 & 3 & 5000 & 0.00401 & 0.00009 & 0.00005 & 0.00010 & 0.00585 \\
##
      \hline
## \end{tabular}
## \end{table}
```

### Bias

```
biasData <- subset(missPlotData, variable %in% c("B1","B2","B3","B4"))
biasData$variable <- paste(biasData$variable)

qplot(x=n,y=value, data=biasData, color=Method) + facet_wrap(~variable*m)</pre>
```

