```
## Load libraries
library(splines)
library(MASS)

## Define the number of tests
ntest <- 1000

## Set nuber of simulations
nSims <- 100##10000

##second shape parameter for beta distribution
shape2 <- 2

source("https://raw.githubusercontent.com/leekgroup/fdrreg/master/R/estPiOReg.R")</pre>
```

Function to generate p-values:

```
genPvals <- function(pi0, shape2)
{
  ntest <- length(pi0)

  nullI <- rbinom(ntest,prob=pi0,size=1)> 0

  pValues <- rep(NA,ntest)
   pValues[nullI] <- runif(sum(nullI))
  pValues[!nullI] <- rbeta(sum(!nullI),1,shape2)

  pValues
}</pre>
```

1 Probability of being a false positive as a linear function of time

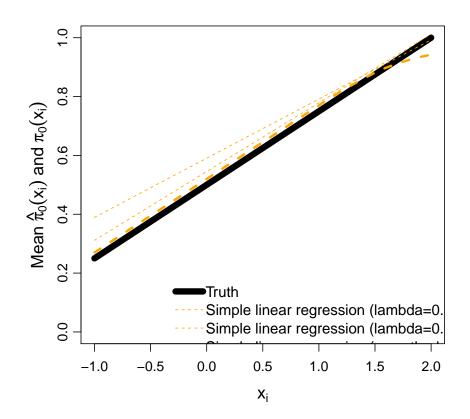
```
set.seed(1345)

## Set up the time vector and the probability of being null
tme <- seq(-1,2,length=ntest)
pi0 <- 1/4*tme+1/2

##save the value of piOhat for each simulation for lambda=0.8, 0.9, final smoothed value
piOhatMatO.8 <- piOhatMatO.9 <- piOhatMatFinal <-
matrix(NA, nrow=nSims, ncol=ntest)</pre>
```

```
for(sim in 1:nSims)
  if(sim\%100 == 0)
    print(sim)
  ##generate p-values
  pValues <- genPvals(pi0, shape2)
  ## Get the estimate
  pi0Est <- estPi0Reg(pValues, X=tme, smooth.df=3)</pre>
  pi0hatMat0.8[sim, ] <- pi0Est$pi0.lambda[,round(pi0Est$lambda,2)==0.8]
  pi0hatMat0.9[sim, ] <- pi0Est$pi0.lambda[,round(pi0Est$lambda,2)==0.9]</pre>
  piOhatMatFinal[sim, ] <- piOEst$piO</pre>
## [1] 100
## Get the mean values:
pi0hatMean0.8 <- colMeans(pi0hatMat0.8)</pre>
pi0hatMean0.9 <- colMeans(pi0hatMat0.9)</pre>
piOhatMeanFinal <- colMeans(piOhatMatFinal)</pre>
##Get the variances:
pi0hatVar0.8 <- apply(pi0hatMat0.8, 2, var)</pre>
pi0hatVar0.9 <- apply(pi0hatMat0.9, 2, var)</pre>
piOhatVarFinal <- apply(piOhatMatFinal, 2, var)</pre>
##Get the variance bounds:
# zMat <- tmeInt
# S <- zMat%*%solve(t(zMat)%*%zMat)%*%t(zMat)
# piOhatVarBound <-</pre>
# diag(S)/(4*(1-0.8)^2)
```

1.1 Plot for means



1.2 Plot for variances

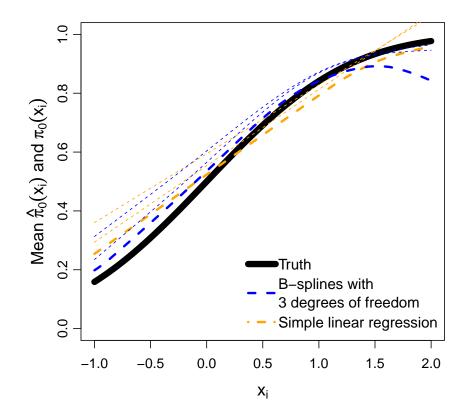
```
# par(cex.axis = 1.1, cex.main=1.3)
#
# plot(pi0hatVarBound ~ tme, col="red", ylim=c(0, max(pi0hatVarBound)),
# lwd=3, lty=3,
# type="l",
# xlab="", ylab="")
# mtext(expression(x[i]), 1, line=3, cex=1.3)
# mtext(expression(paste("Variance and upper bound of variance for ", " ", hat(pi)[0](x[i])
# points(pi0hatVar ~ tme, col="black", type="l", lwd=3, lty=1)
# legend("top", ##x=-0.4, y=0.2,
# legend=c("Empirical variance", "Upper bound"),
# col=c("black", "red"), bty="n",
# lwd=c(3,3), lty=c(1,3),
# cex=1.2, x.intersp=0.2, y.intersp=1.0)
```

2 Probability of being a false positive as a smooth function of time

```
set.seed(1345)
## Set up the time vector and the probability of being null
tme <- seq(-1,2,length=ntest)
pi0 <- pnorm(tme)</pre>
splineMat <- ns(tme,df=3)</pre>
##save the value of piOhat for each simulation for lambda=0.8, 0.9, final smoothed value for
piOhatMatLinO.8 <- piOhatMatLinFinal <-</pre>
 piOhatMatSpl0.8 <- piOhatMatSpl0.9 <- piOhatMatSplFinal <-</pre>
 matrix(NA, nrow=nSims, ncol=ntest)
for(sim in 1:nSims)
  ##generate p-values
 pValues <- genPvals(pi0, shape2)
  ## Get the estimates
 pi0Est.1 <- estPi0Reg(pValues, X=tme, smooth.df=3)</pre>
 pi0Est.2 <- estPi0Reg(pValues, X=splineMat, smooth.df=3)
 pi0hatMatLin0.8[sim, ] <- pi0Est.1$pi0.lambda[,round(pi0Est.1$lambda,2)==0.8]
 pi0hatMatLin0.9[sim, ] <- pi0Est.1$pi0.lambda[,round(pi0Est.1$lambda,2)==0.9]
```

```
piOhatMatLinFinal[sim, ] <- piOEst.1$piO</pre>
  pi0hatMatSpl0.8[sim, ] <- pi0Est.2$pi0.lambda[,round(pi0Est.2$lambda,2)==0.8]</pre>
  pi0hatMatSpl0.9[sim, ] <- pi0Est.2$pi0.lambda[,round(pi0Est.2$lambda,2)==0.9]
  piOhatMatSplFinal[sim, ] <- piOEst.2$piO</pre>
## Get the mean values:
pi0hatMeanLin0.8 <- colMeans(pi0hatMatLin0.8)</pre>
pi0hatMeanLin0.9 <- colMeans(pi0hatMatLin0.9)</pre>
piOhatMeanLinFinal <- colMeans(piOhatMatLinFinal)</pre>
pi0hatMeanSpl0.8 <- colMeans(pi0hatMatSpl0.8)</pre>
piOhatMeanSpl0.9 <- colMeans(piOhatMatSpl0.9)</pre>
piOhatMeanSplFinal <- colMeans(piOhatMatSplFinal)</pre>
##Get the variances:
pi0hatVarLin0.8 <- apply(pi0hatMatLin0.8, 2, var)</pre>
piOhatVarLin0.9 <- apply(piOhatMatLin0.9, 2, var)</pre>
piOhatVarLinFinal <- apply(piOhatMatLinFinal, 2, var)</pre>
pi0hatVarSpl0.8 <- apply(pi0hatMatSpl0.8, 2, var)</pre>
pi0hatVarSpl0.9 <- apply(pi0hatMatSpl0.9, 2, var)</pre>
piOhatVarSplFinal <- apply(piOhatMatSplFinal, 2, var)</pre>
##Get the variance bounds:
# zMat <- splineMatInt</pre>
# S <- zMat%*%solve(t(zMat)%*%zMat)%*%t(zMat)
# piOhatVarBoundFitSpl <-</pre>
  diag(S)/(4*(1-lambda)^2)
# zMat <- cbind(1, tme)
# S <- zMat%*%solve(t(zMat)%*%zMat)%*%t(zMat)
# piOhatVarBoundFitLin <-</pre>
# diag(S)/(4*(1-lambda)^2)
```

2.1 Plot for means



2.2 Plots for variances

par(cex.axis = 1.1, cex.main=1.3)

```
# plot(pi0hatVarBoundFitLin ~ tme, col="red", ylim=c(0, max(pi0hatVarBoundFitLin)),
     lwd=3, lty=3,
      type="l",
      xlab="", ylab="")
# mtext(expression(x[i]), 1, line=3, cex=1.3)
# mtext(expression(paste("Variance and upper bound of variance for ", " ", hat(pi)[0](x[i])
# points(piOhatVarFitLin ~ tme, col="black", type="l", lwd=3, lty=1)
# legend("top", ##x=-0.4, y=0.2,
         legend=c("Empirical variance", "Upper bound"),
        col=c("black", "red"), bty="n",
        lwd=c(3,3), lty=c(1,3),
         cex=1.2, x.intersp=0.2, y.intersp=1.0)
\# par(cex.axis = 1.1, cex.main=1.3)
# plot(pi0hatVarBoundFitSpl ~ tme, col="red", ylim=c(0, max(pi0hatVarBoundFitSpl)),
      lwd=3, lty=3,
      type="l",
      xlab="", ylab="")
#
# mtext(expression(x[i]), 1, line=3, cex=1.3)
# mtext(expression(paste("Variance and upper bound of variance for ", " ", hat(pi)[0](x[i])
# points(pi0hatVarFitSpl ~ tme, col="black", type="l", lwd=3, lty=1)
# legend("top", ##x=-0.4, y=0.2,
```

3 Probability of being a false positive as a sine + step function

legend=c("Empirical variance", "Upper bound"),

col=c("black", "red"), bty="n",

cex=1.2, x.intersp=0.2, y.intersp=1.0)

lwd=c(3,3), lty=c(1,3),

```
set.seed(1345)

## Set up the time vector and the probability of being null
tme1 <- seq(-1*pi,2*pi,length=ntest)
tme2 <- rep(1:0, each=ntest/2)
pi0 <- 1/4*sin(tme1) + tme2/4 + 1/2</pre>
```

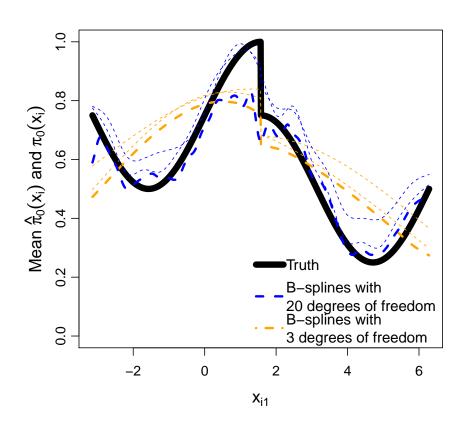
```
##pi0 <- 1/4*sin(tme1) + 0.5
range(pi0)
## [1] 0.2500028 0.9999972
splineMat3 <- cbind(ns(tme1,df=3), tme2)</pre>
splineMat20 <- cbind(ns(tme1,df=20), tme2)</pre>
##save the value of pi0hat for each simulation for lambda=0.8, 0.9, final smoothed value fo
piOhatMat3.0.8 <- piOhatMat3.0.9 <- piOhatMat3.Final <-</pre>
  piOhatMat20.0.8 <- piOhatMat20.0.9 <- piOhatMat20.Final <-</pre>
  matrix(NA, nrow=nSims, ncol=ntest)
for(sim in 1:nSims)
  ##generate p-values
  pValues <- genPvals(pi0, shape2)
  ## Get the estimates
  pi0Est.1 <- estPi0Reg(pValues, X=splineMat3, smooth.df=3)</pre>
  pi0Est.2 <- estPi0Reg(pValues, X=splineMat20, smooth.df=3)</pre>
  pi0hatMat3.0.8[sim, ] <- pi0Est.1$pi0.lambda[,round(pi0Est.1$lambda,2)==0.8]
  piOhatMat3.0.9[sim, ] <- piOEst.1$piO.lambda[,round(piOEst.1$lambda,2)==0.9]</pre>
  piOhatMat3.Final[sim, ] <- piOEst.1$piO</pre>
  pi0hatMat20.0.8[sim, ] <- pi0Est.2$pi0.lambda[,round(pi0Est.2$lambda,2)==0.8]
  pi0hatMat20.0.9[sim, ] <- pi0Est.2$pi0.lambda[,round(pi0Est.2$lambda,2)==0.9]</pre>
  piOhatMat20.Final[sim, ] <- piOEst.2$piO</pre>
## Get the mean values:
pi0hatMean3.0.8 <- colMeans(pi0hatMat3.0.8)</pre>
pi0hatMean3.0.9 <- colMeans(pi0hatMat3.0.9)</pre>
piOhatMean3.Final <- colMeans(piOhatMat3.Final)</pre>
pi0hatMean20.0.8 <- colMeans(pi0hatMat20.0.8)</pre>
pi0hatMean20.0.9 <- colMeans(pi0hatMat20.0.9)</pre>
pi0hatMean20.Final <- colMeans(pi0hatMat20.Final)</pre>
##Get the variances:
pi0hatVar3.0.8 <- apply(pi0hatMat3.0.8, 2, var)</pre>
piOhatVar3.0.9 <- apply(piOhatMat3.0.9, 2, var)</pre>
piOhatVar3.Final <- apply(piOhatMat3.Final, 2, var)</pre>
pi0hatVar20.0.8 <- apply(pi0hatMat20.0.8, 2, var)</pre>
```

```
piOhatVar20.0.9 <- apply(piOhatMat20.0.9, 2, var)
piOhatVar20.Final <- apply(piOhatMat20.Final, 2, var)

# ##Get the variance bounds:
# zMat <- splineMatInt3
# S <- zMat%*%ginv(t(zMat)%*%zMat)%*%t(zMat)
# piOhatVarBound3 <-
# diag(S)/(4*(1-lambda)^2)
#
# zMat <- splineMatInt20
# S <- zMat%*%ginv(t(zMat)%*%zMat)%*%t(zMat)
# piOhatVarBound20 <-
# diag(S)/(4*(1-lambda)^2)</pre>
```

3.1 Plot for means

```
par(cex.axis = 1.1, cex.main=1.3)
plot(pi0 ~ tme1,col="black",type="1",lwd=8, lty=1,
    xlab="", yaxt = "n",
    ylim=c(0,1), ylab="")
mtext(expression(x[i1]), 1, line=3, cex=1.3)
mtext(expression(paste("Mean ", hat(pi)[0](x[i])," and ",
                      pi[0](x[i])), 2, line=2, cex=1.3)
points(pi0hatMean3.0.8 ~ tme1,col="orange",type="l",lwd=1, lty=2)
points(pi0hatMean3.0.9 ~ tme1,col="orange",type="l",lwd=1, lty=2)
points(pi0hatMean3.Final ~ tme1,col="orange",type="1",lwd=3, lty=2)
points(pi0hatMean20.0.8 ~ tme1,col="blue",type="1",lwd=1, lty=2)
points(pi0hatMean20.0.9 ~ tme1,col="blue",type="l",lwd=1, lty=2)
points(pi0hatMean20.Final ~ tme1,col="blue",type="1",lwd=3, lty=2)
legend("bottomright", ##x=-100, y=0.3,
       legend=c("Truth", "B-splines with\n20 degrees of freedom",
                "B-splines with\n3 degrees of freedom"),
       col=c("black", "blue", "orange"), bty="n",
       lwd=c(8,3,3), lty=c(1,2,4),
       cex=1.2, x.intersp=0.2, y.intersp=1.15)
axis(side=2, at=(0:5)/5, mgp=c(3, 0.7, 0))
```



3.2 Plots for variances

```
# par(cex.axis = 1.1, cex.main=1.3)
#
# plot(piOhatVarBound3 ~ tme1, col="red", ylim=c(0, max(piOhatVarBound3)),
# lwd=3, lty=3,
# type="l",
# xlab="", ylab="")
# mtext(expression(x[i1]), 1, line=3, cex=1.3)
# mtext(expression(paste("Variance and upper bound of variance for ", " ", hat(pi)[0](x[i])
# points(piOhatVar3 ~ tme1, col="black", type="l", lwd=3, lty=1)
# legend("top", ##x=-0.4, y=0.2,
# legend=c("Empirical variance", "Upper bound"),
# col=c("black", "red"), bty="n",
# lwd=c(3,3), lty=c(1,3),
```

```
# cex=1.2, x.intersp=0.2, y.intersp=1.0)
```

```
# par(cex.axis = 1.1, cex.main=1.3)
#
# plot(pi0hatVarBound20 ~ tme1, col="red", ylim=c(0, max(pi0hatVarBound20)),
# lwd=3, lty=3,
# type="l",
# xlab="", ylab="")
# mtext(expression(x[i1]), 1, line=3, cex=1.3)
# mtext(expression(paste("Variance and upper bound of variance for ", " ", hat(pi)[0](x[i])
# points(pi0hatVar20 ~ tme1, col="black", type="l", lwd=3, lty=1)
# legend("top", ##x=-0.4, y=0.2,
# legend=c("Empirical variance", "Upper bound"),
# col=c("black", "red"), bty="n",
# lwd=c(3,3), lty=c(1,3),
# cex=1.2, x.intersp=0.2, y.intersp=1.0)
```

Session info:

```
devtools::session_info()
## Session info -----
  setting value
##
  version R version 3.3.1 (2016-06-21)
##
  system x86_64, mingw32
##
   ui
           RTerm
## language (EN)
## collate English_United States.1252
           America/New_York
## tz
## date
           2016-07-28
## Packages -----
   package * version date
                              source
            1.12.0 2016-06-24 CRAN (R 3.3.1)
## devtools
## digest
             0.6.9 2016-01-08 CRAN (R 3.3.1)
## evaluate 0.9
                    2016-04-29 CRAN (R 3.3.1)
##
  formatR
             1.4
                    2016-05-09 CRAN (R 3.3.1)
## highr
           0.6
                    2016-05-09 CRAN (R 3.3.1)
## knitr * 1.13 2016-05-09 CRAN (R 3.3.1)
## magrittr 1.5 2014-11-22 CRAN (R 3.3.1)
## MASS
        * 7.3-45 2016-04-21 CRAN (R 3.3.1)
## memoise 1.0.0 2016-01-29 CRAN (R 3.3.1)
## stringi 1.1.1 2016-05-27 CRAN (R 3.3.0)
## stringr 1.0.0 2015-04-30 CRAN (R 3.3.1)
## withr 1.0.2 2016-06-20 CRAN (R 3.3.1)
```