Algorithmic Analysis for LOTF

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1 Load required functions and library

```
library(ggplot2)
library("gridExtra")

## Warning: package 'gridExtra' was built under R version 3.6.3

source("utils.R")

## Warning: package 'ggpubr' was built under R version 3.6.3
```

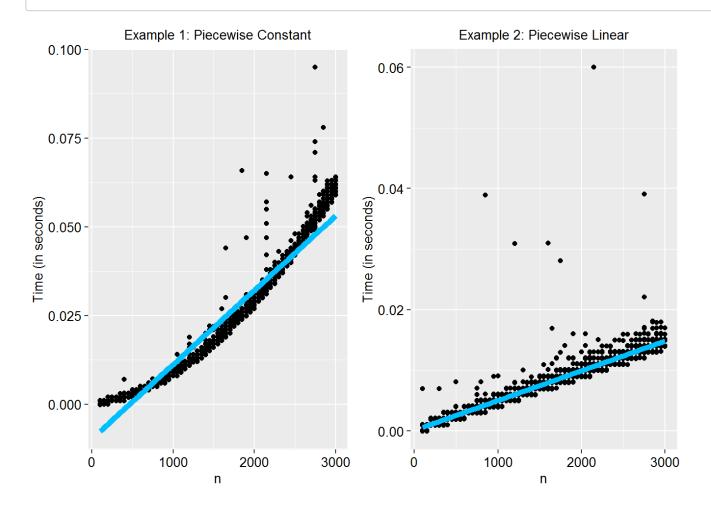
2 Figure 7: Scatterplot of runtime versus sample size

```
nlist \leftarrow seq(100, 3000, 50)
runtime <- array(0, c(length(nlist), 100, 2))
for(i in seq_along(nlist)){
    n <- nlist[i]
    for(seed in 1:100) {
         # Piecewise constant case
         # Toy Piecewise constant/linear: one knot only
         q=0; sigma=0.1; nknot=1
         data = ToyEx(n=n, q=q, sigma=sigma, seed=seed)
         start_time <- Sys.time()</pre>
         amias (data$y, D type="tf0", k=nknot)
          end_time <- Sys. time()</pre>
         runtime[i, seed, 1] <- end_time - start_time</pre>
         q=1; sigma=0.1; nknot=1
          data = ToyEx(n=n, q=q, sigma=sigma, seed=seed)
         start_time <- Sys.time()</pre>
         amias(data$y, D_type="tfq", q=q, k=nknot)
         end_time <- Sys.time()</pre>
         runtime[i, seed, 2] <- end_time - start_time</pre>
  }
save.image("runtime_amias.RData")
# load("runtime_amias.RData")
t1 <- as.numeric(t(runtime[,,1]))</pre>
data <- cbind(rep(nlist, each=100), t1)
data <- as. data. frame (data)
\texttt{colnames}\,(\texttt{data}) \, \leftarrow \, \texttt{c}\,(\texttt{"n", "time"})
library (ggplot2)
p1 \leftarrow ggplot(data = data, aes(x=n, y=time)) + geom_jitter(height = 0.0001, width=0) +
    geom_smooth(method=lm, se= FALSE, size=2, color = "deepskyblue") +
     ggtitle("Example 1: Piecewise Constant") +
    labs(x="n", y = "Time (in seconds)") +
    theme(axis.text.x = element_text(size = 10, color="black"),
                   axis.title.x = element_text(size = 10, color="black"),
                   axis.text.y = element_text(size = 10, color="black"),
                    axis.title.y = element_text(size = 10, color="black"),
                   plot.title = element_text(size = 10, hjust = 0.5))
t1 \leftarrow as. numeric(t(runtime[,,2]))
t1[which.max(t1)] \leftarrow NA
```

```
data <- cbind(rep(nlist, each=100), t1)
data <- as.data.frame(data)
colnames(data) <- c("n", "time")
library(ggplot2)
p2 <- ggplot(data = data, aes( x= n, y= time)) + geom_jitter(height = 0.0001, width=0) +
geom_smooth(method=lm, se= FALSE, size=2, color = "deepskyblue") +
ggtitle("Example 2: Piecewise Linear") +
labs(x="n", y = "Time (in seconds)") +
theme (axis.text. x = element_text(size = 10, color="black"),
    axis.title. x = element_text(size = 10, color="black"),
    axis.text. y = element_text(size = 10, color="black"),
    axis.title. y = element_text(size = 10, color="black"),
    plot.title = element_text(size = 10, hjust = 0.5))</pre>
# png("figs/amias_time_n.png", pointsize=6, width=850, height=400, res=120)
grid.arrange(p1, p2, ncol=2)
```

Warning: Removed 1 rows containing non-finite values (stat_smooth).

Warning: Removed 1 rows containing missing values (geom_point).

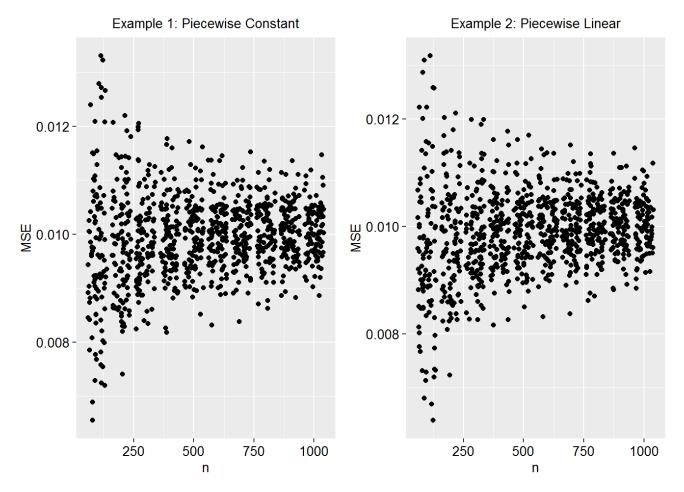


3 Figure 8: gitterplot of MSE versus sample size in Example 1 and Example 2

```
nlist \leftarrow seq(100, 1000, 100)
mse \leftarrow array(0, c(length(nlist), 100, 2))
for(i in seq_along(nlist)){
  n <- nlist[i]
 print(n)
  for(seed in 1:100) {
    # Piecewise constant case
    sigma=0.1; q=0; nknot = 1
    data = ToyEx(n=n, q=q, sigma=sigma, seed=seed)
    resL0 <- amias(data$y, D_type="tf0", k=nknot)</pre>
    mse[i, seed, 1] <- mean((as.numeric(data$y)-resL0$alpha)^2)</pre>
    # Piecewise linear case
    sigma=0.1; q=1; nknot = 1
    data = ToyEx(n=n, q=q, sigma=sigma, seed=seed)
    resL0 <- amias(data$y, D_type="tfq", q=q, k=nknot)
    mse[i, seed, 2] <- mean((as.numeric(data$y)-resL0$alpha)^2)</pre>
```

```
## [1] 100
## [1] 200
## [1] 300
## [1] 400
## [1] 500
## [1] 600
## [1] 700
## [1] 800
## [1] 900
## [1] 1000
```

```
save.image("amias_mse_n_single.RData")
load("amias_mse_n_single.RData")
title <- c("Example 1: Piecewise Constant", "Example 2: Piecewise Linear")
p <- 1ist()
for(j in 1:2){
 t1 <- as. numeric(t(mse[,,j]))
  data <- cbind(rep(nlist, each=100), t1)
  data <- as. data. frame (data)
 colnames(data) <- c("n", "time")</pre>
 library (ggplot2)
 \texttt{p[[j]]} \leftarrow \texttt{ggplot(data = data, aes(x=n, y=time)) + geom\_jitter()+}
    ggtitle(title[j]) +
    labs(x="n", y = "MSE") +
    theme(axis.text.x = element_text(size = 10, color="black"),
          axis.title.x = element_text(size = 10, color="black"),
          axis.text.y = element_text(size = 10, color="black"),
          axis.title.y = element_text(size = 10, color="black"),
          plot.title = element_text(size = 10, hjust = 0.5))
# png("figs/amias_mse_n_1.png", pointsize=6, width=850, height=400, res=120)
grid.arrange(p[[1]], p[[2]], ncol=2)
```



4 Figure 9: Gitterplot of MSE versus sample size

```
nlist <- seq(100, 1000, 100)
mse \leftarrow array(0, c(length(nlist), 100, 4))
for(i in seq_along(nlist)){
      n <- nlist[i]
      print(n)
      for(seed in 1:100) {
            # Piecewise constant case
            sigma=0.1; q=0; nknot=2;
            data = SimuEx(n=n, sigma=sigma, q=q, nknot=nknot, seed=seed)
            # Empty set initialization
            resL0 <- amias (data$y, D_type="tf0", k=nknot)
            mse[i, seed, 1] <- mean((as.numeric(data$y)-resL0$alpha)^2)</pre>
            # Random set initialization
            resL0 <- amias(data$y, D_type="tf0", k=nknot, A = sample(n-q-1, nknot))
            mse[i, seed, 2] <- mean((as.numeric(data$y)-resL0$alpha)^2)</pre>
            # Stepwise random set initialization
            resLO_1 <- amias(data$y, D_type="tf0", k=1)
            resL0 \leftarrow amias(data\$y, D_type="tf0", k=nknot, A = c(resL0_1\$A, sample(setdiff(1:(n-q-1), resL0_1\$A), 1) + (resL0_1\$A) + (resL0_
)))
            mse[i, seed, 3] <- mean((as.numeric(data$y)-resL0$alpha)^2)</pre>
            # Warm start initialization
            resL0_1 \leftarrow amias(data\$y, D_type="tf0", k=1)
            resL0 <- amias(data$y, D_type="tf0", k=nknot, A = c(resL0_1$A, which.max(abs(resL0_1$u))))
            mse[i, seed, 4] <- mean((as.numeric(data$y)-resL0$alpha)^2)</pre>
```

```
## [1] 100

## [1] 200

## [1] 300

## [1] 400

## [1] 500

## [1] 600

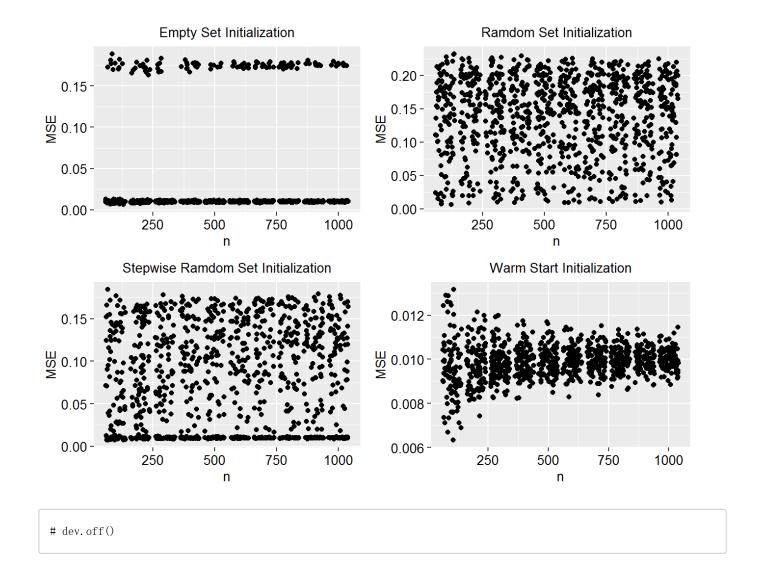
## [1] 700

## [1] 800

## [1] 900

## [1] 1000
```

```
save.image("amias_mse_n.RData")
load ("amias_mse_n.RData")
title <- c("Empty Set Initialization", "Ramdom Set Initialization", "Stepwise Ramdom Set Initialization", "War
m Start Initialization")
p <- list()
for(j in 1:4){
 t1 <- as. numeric(t(mse[,,j]))
  data <- cbind (rep(nlist, each=100), t1)
  data <- as.data.frame(data)
  colnames(data) <- c("n", "time")</pre>
 library (ggplot2)
 p[[j]] \leftarrow ggplot(data = data, aes(x=n, y=time)) + geom_jitter()+
    ggtitle(title[j]) +
    labs(x="n", y = "MSE") +
    theme(axis.text.x = element_text(size = 10, color="black"),
          axis.title.x = element_text(size = 10, color="black"),
          axis.text.y = element_text(size = 10, color="black"),
          axis.title.y = element_text(size = 10, color="black"),
          plot.title = element_text(size = 10, hjust = 0.5))
}
# png("figs/amias_mse_n.png", pointsize=6, width=850, height=800, res=120)
grid.arrange(p[[1]], p[[2]], p[[3]], p[[4]], nco1=2)
```



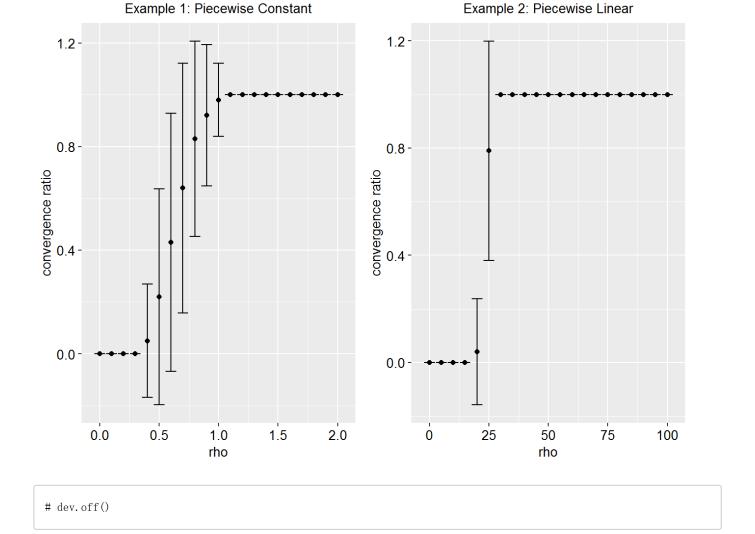
5 Figure 10: Error bars of the proportion of convergence against the mixing parameter ρ

```
# amias rho vs convergence
c1 <- c2 <- c()
for(i in 1:100){
       if (i%%20==0) print(i)
       data1 = SimuEx(n=100, sigma=0.1, q=0, nknot=1, seed=i)
       data2 = SimuEx(n=100, sigma=0.1, q=1, nknot=1, seed=i)
      rho1 < -seq(0, 2, 0.1)
       rho2 <- seq (0, data2$n, 5)
       r1 <- r2 <- c()
       for (rho in rho1) {
               res1 = amias\_R (y = data1\$y, D = DiffMat (data1\$n, data1\$nknot), A = c(), k = data1\$nknot, rho = rho, q = 
   data1$q)
              if (res1$iter==20) {
                    r1 <- c (r1, 0)
              }else{
                      r1 <- c (r1, 1)
               }
       for(rho in rho2) {
               res2 = amias_R(y = data2$y, D = DiffMat(data2$n, data2$nknot), A = c(), k = data2$nknot, rho = rho, q =
   data2$q)
               if(res2$iter==20){
                   r2 <- c (r2, 0)
             }else{
                    r2 <- c (r2, 1)
              }
      }
       c1 \leftarrow rbind(c1, r1)
       c2 <- rbind(c2, r2)
## [1] 20
## [1] 40
```

```
p1 = convsteps(t(c1), rho1, "Example 1: Piecewise Constant", 'rho', 'convergence ratio')
p2 = convsteps(t(c2), rho2, "Example 2: Piecewise Linear", 'rho', 'convergence ratio')
save.image("amias_rho_conv.RData")

# png("figs/amias_rho_conv.png", pointsize=8, width=850, height=400, res=120)
ggarrange(p1, p2, ncol = 2)
```

[1] 60 ## [1] 80 ## [1] 100



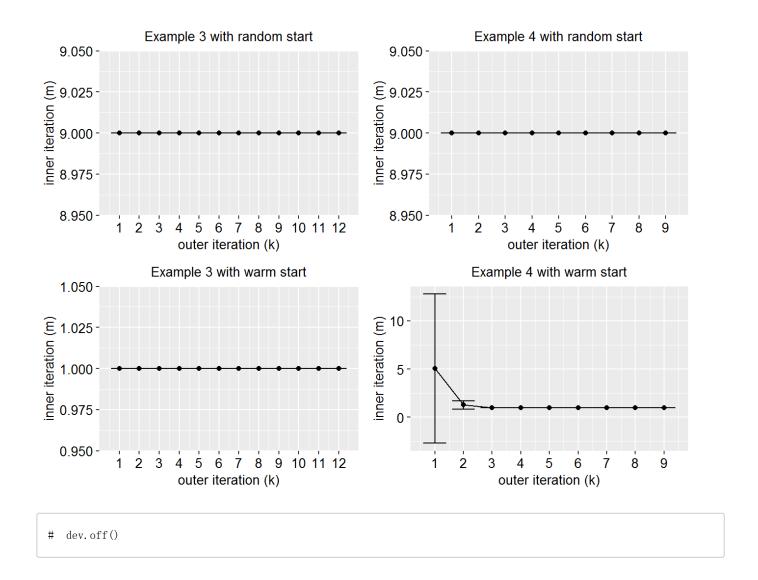
6 Figure 11: Error bars of inner iterations against the outer iterations in the sequential AMIAs algorithm

```
library (ggpubr)
c3 <- c4 <- c5 <- c6 <- c()
for(i in 1:100) {
       if (i%%20==0) print(i)
        data3 = SimuEx(n=300, sigma=0.1, q=0, nknot=8, seed=i)
        data4 = SimuEx(n=300, sigma=0.1, q=1, nknot=5, seed=i)
      res1 = ramias(data3, kmax = data3$nknot+4)
      res2 = ramias(data4, kmax = data4$nknot+4)
      res3 = samias_R (y = data3\$y, D = DiffMat (data3\$n, data3\$q+1), kmax = data3\$nknot+4, rho = data3\$n**(data3\$n, data3\$q+1), kmax = data3\$nknot+4, rho = data3\$nkno
q+1, q = data3q
        res4 = samias_R(y = data4$y, D = DiffMat(data4$n, data4$q+1), kmax = data4$nknot+4, rho = data4$n**(data4
q+1, q = data4q
        c3 = rbind(c3, res1$iters)
      c4 = rbind(c4, res2$iters)
      c5 = rbind(c5, res3\$iters)
      c6 = rbind(c6, res4$iters)
```

```
## [1] 20
## [1] 40
## [1] 60
## [1] 80
## [1] 100
```

```
p1 = inout_loop(c3, "Example 3 with random start", 'outer iteration (k)', 'inner iteration (m)')
p2 = inout_loop(c4, "Example 4 with random start", 'outer iteration (k)', 'inner iteration (m)')
p3 = inout_loop(c5, "Example 3 with warm start", 'outer iteration (k)', 'inner iteration (m)')
p4 = inout_loop(c6, "Example 4 with warm start", 'outer iteration (k)', 'inner iteration (m)')
save. image("samias_conv_inout.RData")

# png("figs/samias_conv_inout.png", pointsize=8, width=850, height=850, res=120)
ggarrange(p1, p2, p3, p4, nrow=2, ncol = 2)
```



7 Figure 12: Error bars of MSE over cardinality parameters

```
c3 <- c4 <- c5 <- c6 <- c()
for(i in 1:100) {
    if (i%20=0) print(i)
    data3 = SimuEx(n=300, sigma=0.1, q=0, nknot=8, seed=i)
    data4 = SimuEx(n=300, sigma=0.1, q=1, nknot=5, seed=i)
    res1 = ramias(data3, kmax = data3$nknot+12)
    res2 = ramias(data4, kmax = data4$nknot+15)
    res3 = samias_R(y = data3$y, D = DiffMat(data3$n, data3$q+1), kmax = data3$nknot+12, rho = data3$n**(data3$q+1), q = data3$q)
    res4 = samias_R(y = data4$y, D = DiffMat(data4$n, data4$q+1), kmax = data4$nknot+15, rho = data4$n**(data4$q+1), q = data4$q)
    c3 = rbind(c3, res1$mse)
    c4 = rbind(c4, res2$mse)
    c5 = rbind(c5, res3$mse)
    c6 = rbind(c6, res4$mse)
}
```

```
## [1] 20

## [1] 40

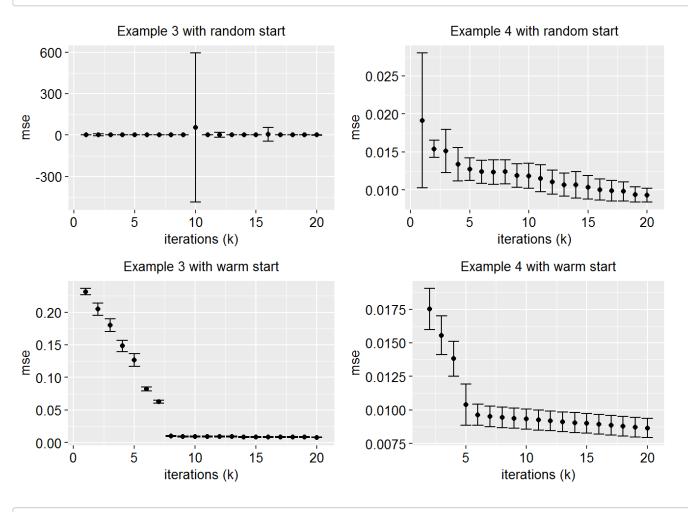
## [1] 60

## [1] 80

## [1] 100
```

```
p1 = convsteps(t(c3), 1:(data3$nknot+12), "Example 3 with random start", "iterations (k)", "mse")
p2 = convsteps(t(c4), 1:(data4$nknot+15), "Example 4 with random start", "iterations (k)", "mse")
p3 = convsteps(t(c5), 1:(data3$nknot+12), "Example 3 with warm start", "iterations (k)", "mse")
p4 = convsteps(t(c6[,-1]), 2:(data4$nknot+15), "Example 4 with warm start", "iterations (k)", "mse") # cut fi
rst col

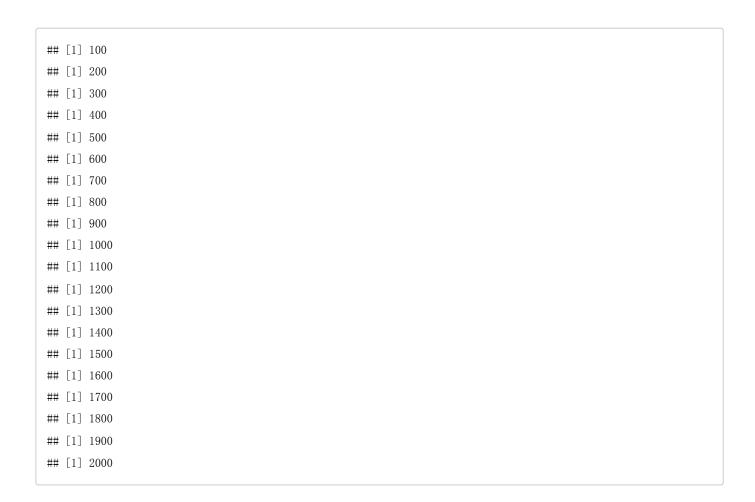
save.image("samias_mse_conv.RData")
# png("figs/samias_mse_conv.png", pointsize=8, width=850, height=850, res=120)
ggarrange(p1, p2, p3, p4, nrow=2, ncol = 2)
```



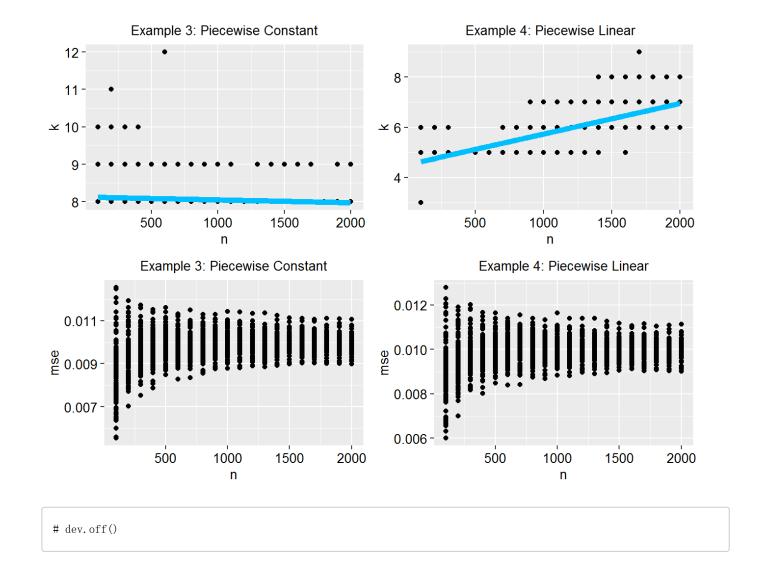
dev.off()

8 Figure 13: Scatterplot of number of the dtected knots against sample size

```
nlist < -seq(100, 2000, 100)
re \langle -\text{ re_o} \langle -\text{ re_u} \langle -\text{ kest } \langle -\text{ mse } \langle -\text{ array } (0, \text{ c } (\text{length } (\text{nlist}), 100, \text{ 2}))
for(n in nlist){
  print(n)
  for(seed in 1:100) {
    par (mfrow=c(1, 2), mar=c(3, 3, 3, 3))
    # Piecewise constant case
    sigma=0.1; q=0; nknot=8;
    data = SimuEx(n=n, sigma=sigma, q=q, nknot=nknot, seed=seed)
    resL0 = samias(as.numeric(data$y), D_type="tf0", kmax=nknot+4)
    mse[n/100, seed, 1] \leftarrow mean((as.numeric(data$y)-resL0$alpha)^2)
    kest[n/100, seed, 1] \leftarrow resL0\$kopt
    re[n/100, seed, 1] \leftarrow nknot==resL0$kopt
    re_o[n/100, seed, 1] \leftarrow nknot < resL0$kopt
    re_u[n/100, seed, 1] \leftarrow nknot > resL0  kopt
    sigma=0.1; q=1; nknot=5;
    data = SimuEx(n=n, sigma=sigma, q=q, nknot=nknot, seed=seed)
    resLO = samias(as.numeric(data$y), D_type="tfq", q=q, kmax=nknot+4, adjust = TRUE)
    mse[n/100, seed, 2] \leftarrow mean((as.numeric(data\$y)-resL0\$alpha)^2)
    kest[n/100, seed, 2] \leftarrow resL0$kopt
    re[n/100, seed, 2] \leftarrow nknot==resL0$kopt
    re_o[n/100, seed, 2] <- nknot<resL0$kopt
    re\_u[n/100, seed, 2] <- nknot>resL0\$kopt
  }
```



```
save.image("smias_k_n.RData")
load ("smias_k_n. RData")
library(ggplot2)
title <- c ("Example 3: Piecewise Constant", "Example 4: Piecewise Linear")
p <- 1ist()
for(j in 1:2){
t1 <- as. numeric(t(kest[,,j]))
data <- cbind(rep(nlist, each=100), t1)
data <- as.data.frame(data)
colnames (data) <- c ("n", "k")
p[[j]] \leftarrow ggplot(data = data, aes(x=n, y=k)) + geom point() +
      geom_smooth(method=lm, se= FALSE, size=2, color = "deepskyblue") + ggtitle(title[j]) +
      theme(axis.text.x = element_text(size = 10, color="black"),
            axis.title.x = element_text(size = 10, color="black"),
            axis.text.y = element_text(size = 10, color="black"),
            axis.title.y = element_text(size = 10, color="black"),
            plot.title = element_text(size = 10, hjust = 0.5))
}
p2 <- 1ist()
for(j in 1:2){
 t1 \leftarrow as.numeric(t(mse[,,j]))
 data <- cbind (rep (nlist, each=100), t1)
  data <- as. data. frame (data)
  colnames(data) <- c("n", "mse")</pre>
  \texttt{p2[[j]]} \leftarrow \texttt{ggplot(data = data, aes(x=n, y=mse)) + geom\_point() +} \\
       ggtitle(title[j]) +
      theme(axis.text.x = element_text(size = 10, color="black"),
            axis.title.x = element_text(size = 10, color="black"),
            axis.text.y = element_text(size = 10, color="black"),
            axis.title.y = element_text(size = 10, color="black"),
            plot.title = element_text(size = 10, hjust = 0.5))
# png("figs/samias_k_n.png", pointsize=8, width=850, height=850, res=120)
grid.arrange(p[[1]], p[[2]], p2[[1]], p2[[2]], ncol=2)
```



9 Figure 14: Early stoppint rule effect on the convergence of sequential AMIAS algorithm

```
# samias eps vs convergence
c1 = c2 = c()
for(i in 1:100) {
 if(i%%20==0) print(i)
  data3 = SimuEx(n=300, sigma=0.1, q=0, nknot=8, seed=i)
  data4 = SimuEx(n=300, sigma=0.1, q=1, nknot=5, seed=i)
 r1 <- r2 <- c()
  sig1 \leftarrow median(abs(diff(data3\$y, diff=1)))/(qnorm(3/4)*sqrt(choose(2, 1)))
  sig2 \leftarrow median(abs(diff(data4\$y, diff=2)))/(qnorm(3/4)*sqrt(choose(4,2)))
  eps1 \leftarrow c(0, seq(0.9, 1, 0.02))*(sig1^2)
  eps2 \leftarrow c(0, seq(0.9, 1, 0.02))*(sig2^2)
  for (eps in eps1) {
    res1 = samias_R(y = data3$y, D = DiffMat(data3$n, data3$q+1), kmax = data3$nknot+4, rho = data3$n**(data
3$q+1), q = data3$q, eps=eps)
    r1 <- c(r1, res1$outiter)
  for (eps in eps2) {
   res2 = samias_R(y = data4$y, D = DiffMat(data4$n, data4$q+1), kmax = data4$nknot+4, rho = data4$n**(data
4$q+1), q = data4$q, eps=eps)
    r2 \leftarrow c (r2, res2\$outiter)
 c1 <- rbind(c1, r1)
 c2 <- rbind(c2, r2)
## [1] 20
## [1] 40
## [1] 60
## [1] 80
## [1] 100
```

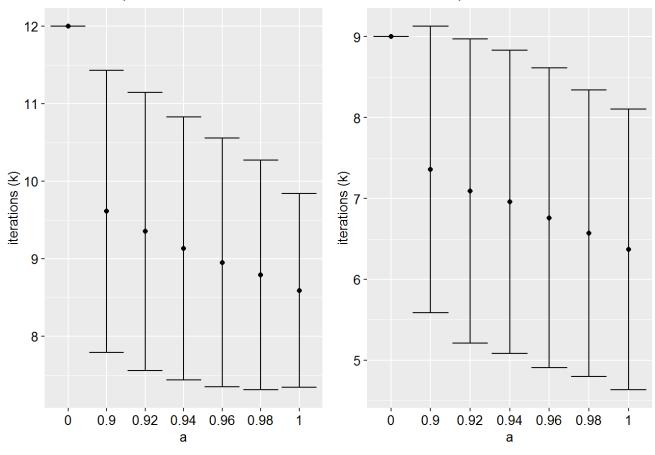
```
epsl <- c(0, seq(0.9,1,0.02))
pl = convsteps(t(c1), as.character(epsl), "Example 3: Piecewise Constant", "a", "iterations (k)")
p2 = convsteps(t(c2), as.character(epsl), "Example 4: Piecewise Linear", "a", "iterations (k)")
save.image("samias_eps_conv.RData")

load("samias_eps_conv.RData")

# png("figs/samias_eps_conv.png", pointsize=8, width=850, height=400, res=120)
ggarrange(p1, p2, ncol = 2)</pre>
```

Example 3: Piecewise Constant

Example 4: Piecewise Linear



dev.off()