simulation_week7

Tingyu Zhu

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```
library(tidyverse)
library(purrr)
library(ggplot2)
library(here)
devtools::load_all()
set.seed(1222)
```

Inspection about the bandwidth. Compare the performance under different methods of selecting the bandwidth.

Bulid functions that can simulate over different sample sizes & kernels & bandwidths.

```
kde_est_big <- function(x, n, ker, h, grid){</pre>
 # x: large population
  # n: sample sizes
 # ker: name of kernels
  # h: bandwidths
  # grid grid points
  1 <- data.frame(n=n, ker=ker, h=h) %>%
    mutate( x = map(.x=n, ~sample(x, .x, replace = FALSE)))
 1$n <- NULL
 pmap(1, KDE_est, grid=grid)
sim_big_made <- function(x, ns, kes, h, grid, true_f){</pre>
  # generates every combination of parameters (ns:kers:hs)
  simulation_params_big <- list(</pre>
 n = ns,
 kernel_type = kers,
 h = hs)
  est_big <- cross_df(simulation_params_big)</pre>
  est_big <- est_big %>%
 mutate(
```

```
f_ests = kde_est_big(x, n, kernel_type, h, grid),
  f_true = map(1:length(n), ~true_f)
) %>%
mutate(
  MADE = map2_dbl(.x = f_ests, .y = f_true, ~made(.x,.y))
)
return(est_big$MADE)
}
```

Simulation

```
x \leftarrow rt(5000, df=15)
ns <- c(100, 500)
kers <- c("normal", "epanech", "uniform", "biweight", "triweight")</pre>
hs <-c(0.1,0.5,1)
grid <- seq(-5,15, 0.1)
true_f <- dt(grid, df=15)</pre>
n.sim <- 50
sim_rlt <- map(1:n.sim, ~sim_big_made(x, ns, kers, hs, grid,true_f))</pre>
made_mat_big <- matrix(unlist(sim_rlt), nrow=length(ns)*length(hs)*length(kers))</pre>
made_comapre <- cross_df(list(n = ns, kernel_type = kers, h = hs)) %>%
  mutate(made_mean=apply(made_mat_big, 1, mean),
         made_sd=apply(made_mat_big, 1, sd))
rownames (made mat big) <- paste (made comapre$n,
                             made_comapre$kernel_type, made_comapre$h, sep = ", ")
write_rds(made_mat_big , here("results", "week7-sim-1.rds"))
made_comapre
```

```
## # A tibble: 30 x 5
##
        n kernel_type h made_mean made_sd
##
     <dbl> <chr>
                  <dbl>
                                     <dbl>
                             <dbl>
## 1 100 normal
                      0.1 0.0147 0.00220
                      0.1 0.00698 0.00112
## 2
       500 normal
## 3
       100 epanech
                      0.1 0.0220 0.00230
## 4
       500 epanech
                      0.1 0.00990 0.00117
## 5
       100 uniform
                       0.1 0.0202 0.00235
                       0.1 0.00911 0.00115
## 6
       500 uniform
## 7
       100 biweight
                       0.1 0.0243 0.00277
## 8
       500 biweight
                       0.1 0.0108 0.00116
## 9
       100 triweight
                       0.1 0.0251 0.00252
                       0.1 0.0120 0.00127
## 10
       500 triweight
## # ... with 20 more rows
```

Types of bandwidth selectors

1. Normal reference bandwidth selector

The normal reference bandwidth selector is defined by:

$$\hat{h}_{opt} = \left\{ \begin{array}{l} 1.06 sn^{-1/5} \text{ for the Gaussian kernel} \\ 2.34 sn^{-1/5} \text{ for the Epanechnikov kernel} \end{array} \right.$$

The normal reference bandwidth selector is only a simple rule of thumb. It is a good selector when the data are nearly Gaussian distributed. However, it can lead to over-smooth when the underlying distribution is asymmetric or multi-modal.

2. plug-in bandwidth selector

There are quite a few important techniques for selecting the bandwidth such as cross-validation (CV) and plug-in bandwidth selectors.

Function dpik() in the package KernSmooth in R selects a bandwidth for estimating the kernel density estimation using the plug-in method.

```
library("KernSmooth")

## Warning: package 'KernSmooth' was built under R version 4.0.2

## KernSmooth 2.23 loaded
## Copyright M. P. Wand 1997-2009

nx <- 5000
x <- rt(nx, df=15)

(h_opt_Gau <- 1.06*sd(x)*nx^(-0.2))

## [1] 0.2038434

(h_opt_Epa <- 2.34*sd(x)*nx^(-0.2))

## [1] 0.4499939

(h_plug_Gau <- dpik(x,kernel = "normal") )

## [1] 0.1835798

(h_plug_Epa <- dpik(x,kernel = "epanech"))

## [1] 0.4064099</pre>
```

```
(h_ran_Gau <- 0.5*nx^(-0.2))
## [1] 0.09102821
(h_ran_Epa <- 0.5*nx^(-0.2))
## [1] 0.09102821
ns < -c(4000)
kers <- c("normal", "epanech")</pre>
hs <- c(h_opt_Gau,h_plug_Gau,h_ran_Gau, h_opt_Epa, h_plug_Epa,h_ran_Epa)
grid \leftarrow seq(-5,15, 0.1)
true_f <- dt(grid, df=15)</pre>
n.sim < -50
sim_rlt_bdwt <- map(1:n.sim, ~sim_big_made(x, ns, kers, hs, grid,true_f))</pre>
made_mat_bdwt <- matrix(unlist(sim_rlt_bdwt), nrow=length(ns)*length(hs)*length(kers))</pre>
made_comapre_bdwt <- cross_df(list(n = ns, kernel_type = kers, h = hs)) %>%
  mutate(made_mean=apply(made_mat_bdwt, 1, mean),
         made_sd=apply(made_mat_bdwt, 1, sd))
rownames(made_mat_bdwt) <- paste(made_comapre_bdwt$n,</pre>
                            made comapre bdwt$kernel type,
                            round(made_comapre_bdwt$h,4), sep = ", ")
write_rds(made_mat_bdwt , here("results", "week7-sim-2.rds"))
made_comapre_bdwt
## # A tibble: 12 x 5
##
          n kernel_type
                             h made_mean made_sd
##
      <dbl> <chr>
                         <dbl>
                                    <dbl>
                                             <dbl>
                        0.204
                                 0.00181 0.000212
##
   1 4000 normal
## 2 4000 epanech
                        0.204
                                 0.00316 0.000267
## 3 4000 normal
                        0.184
                                 0.00199 0.000194
## 4 4000 epanech
                        0.184
                                 0.00334 0.000170
## 5 4000 normal
                        0.0910
                                 0.00320 0.000244
## 6 4000 epanech
                        0.0910
                                 0.00456 0.000290
## 7 4000 normal
                        0.450
                                 0.00415 0.000256
## 8 4000 epanech
                        0.450
                                 0.00184 0.000202
## 9 4000 normal
                        0.406
                                 0.00342 0.000273
## 10 4000 epanech
                        0.406
                                 0.00183 0.000201
## 11 4000 normal
                                 0.00314 0.000228
                        0.0910
## 12 4000 epanech
                        0.0910
                                 0.00466 0.000244
```

Results show that under this setting, the optimal rule of thumb bandwidths and plug-in bandwidths have the smallest MADE under their corresponding kernels.