

# Homework 3

This homework is due by the end of the day on November 5th 2018. Solutions should appear as a vignette in your package called “homework-3”. If you do not want to write your solution for questions 2 - 4 in L<sup>A</sup>T<sub>E</sub>X, you may provide a solution using paper and pencil and include an image in your write-up.

1. CASL page 117, question 7.
  2. CASL page 200, question 3
  3. CASL page 200, question 4
  4. CASL page 200, question 5
  5. CASL page 200, question 6
- 

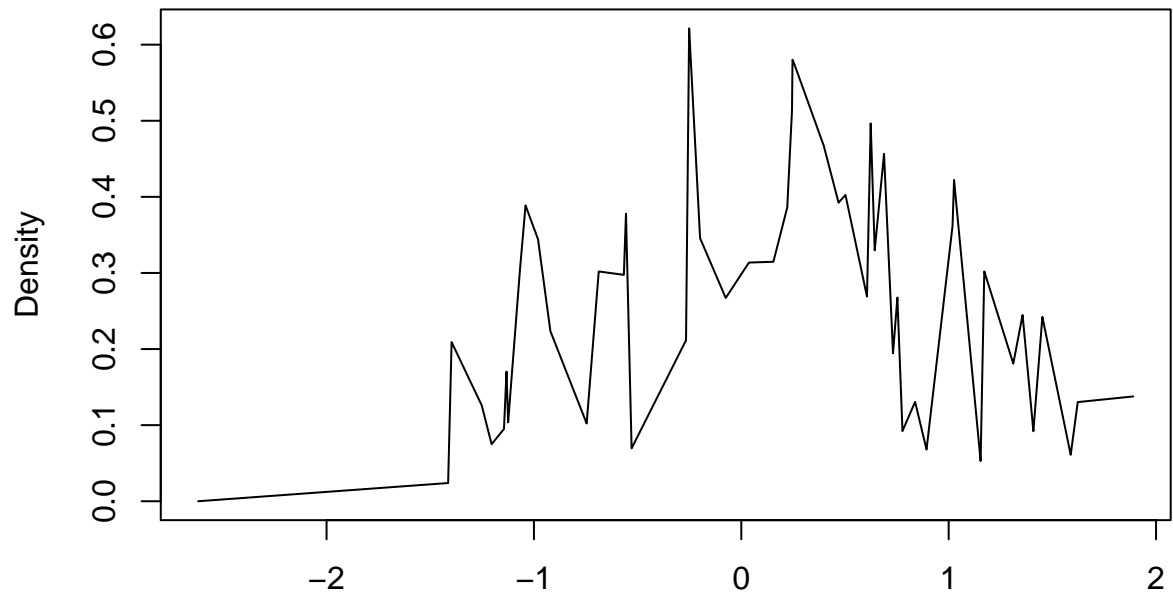
1.

```
set.seed(6909)
#write out the Epanechnikov kernel function
kernel_epan <- function(x) {
  ran <- as.numeric(abs(x) <= 1)
  val <- (3/4) * ( 1 - x^2 ) * ran
  return(val)
}

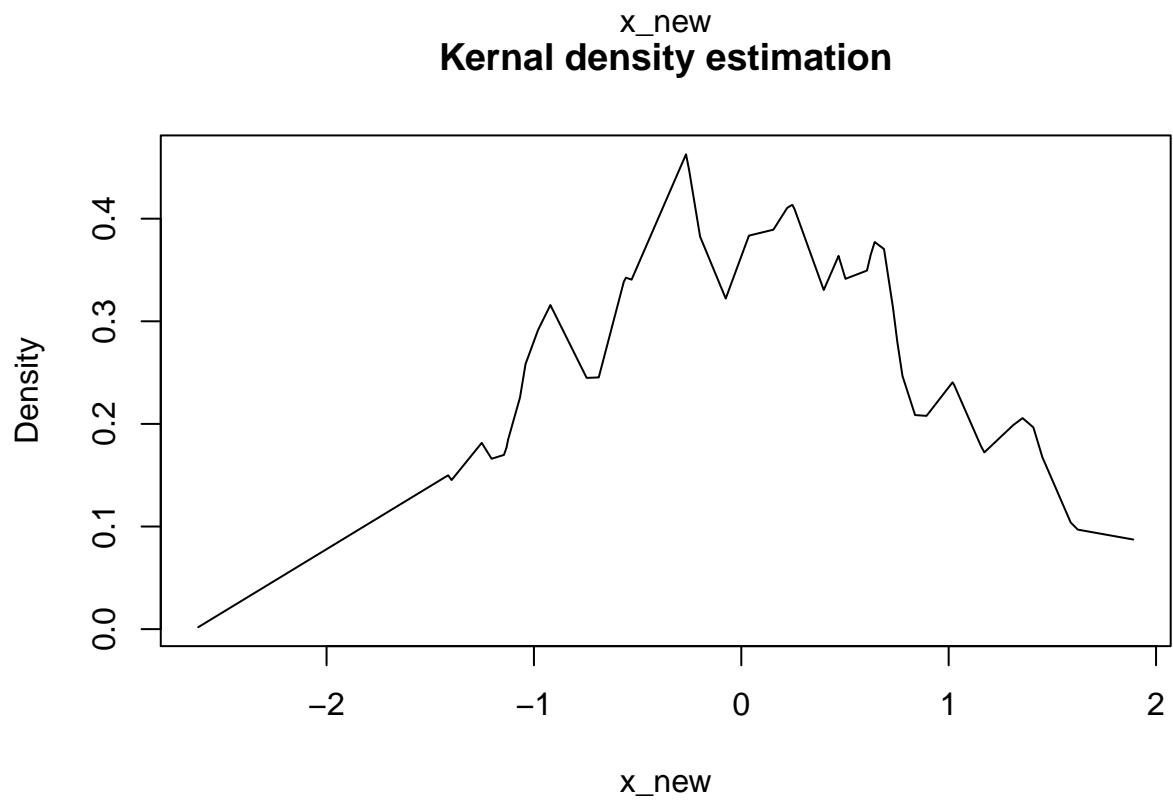
#write out the Epanechnikov density function
i = 1
h = 1
kern_density <- function(x, h, x_new){
  dst_est <- numeric()
  for (i in 1:length(x_new)){
    dst_est[i] <- mean(kernel_epan((x_new[i]-x)/h))/h
  }
  dst_est
}

#visualize different bandwidths
x <- rnorm(1000, 0, 1)
x_new <- sort(rnorm(50, 0, 1))
h = c(0.01, 0.1, 0.5, 1, 2, 3)
for (i in h){
  plot(x_new, kern_density(x,i,x_new), ylab = "Density", main = "Kernal density estimation", type = "l")
}
```

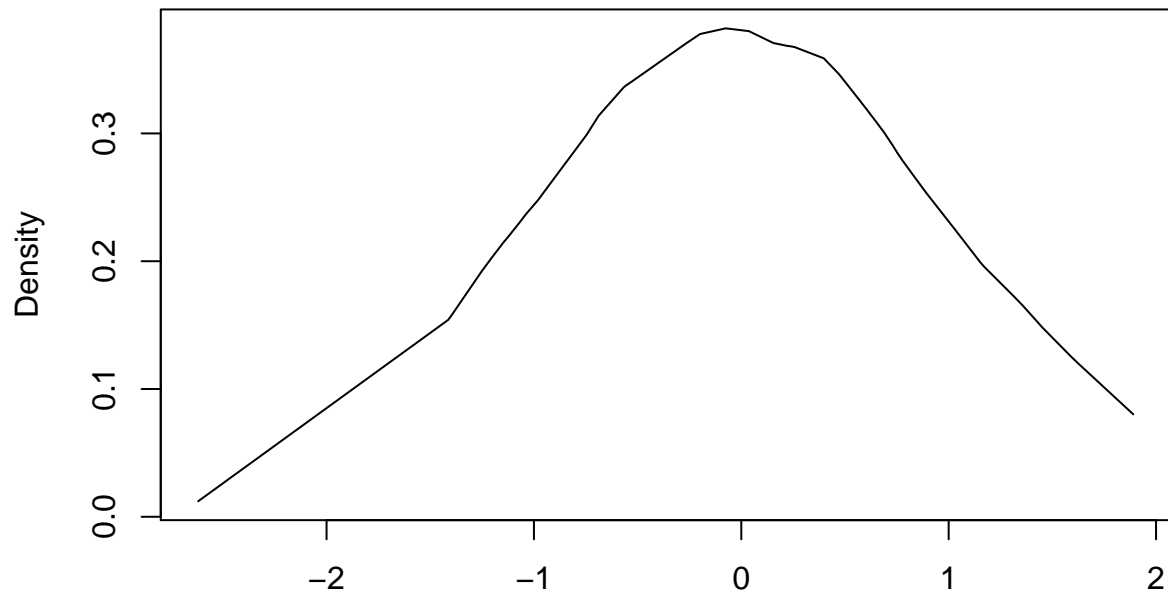
**Kernal density estimation**



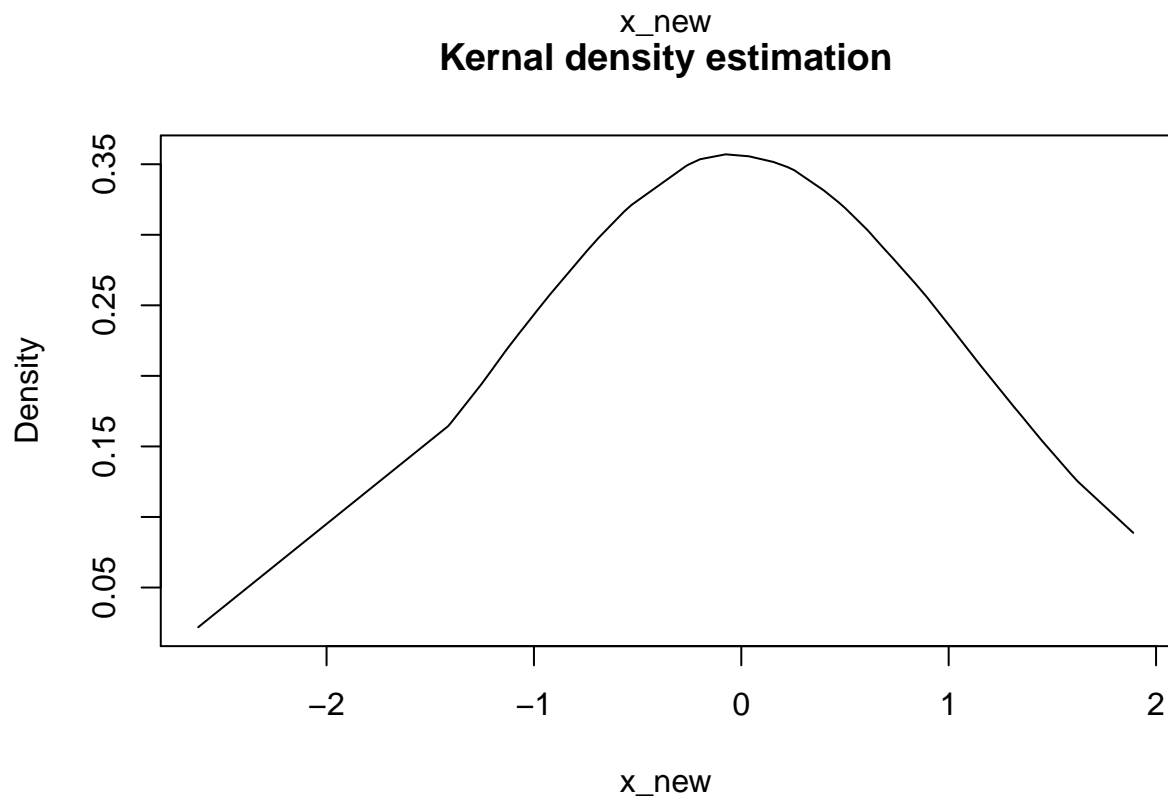
**Kernal density estimation**



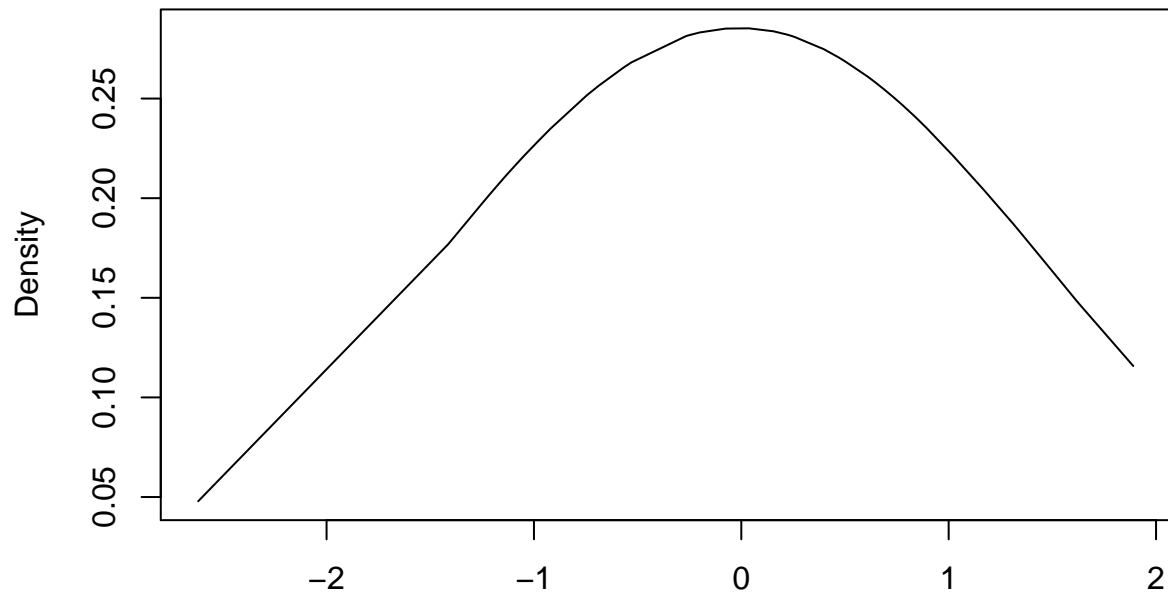
**Kernal density estimation**



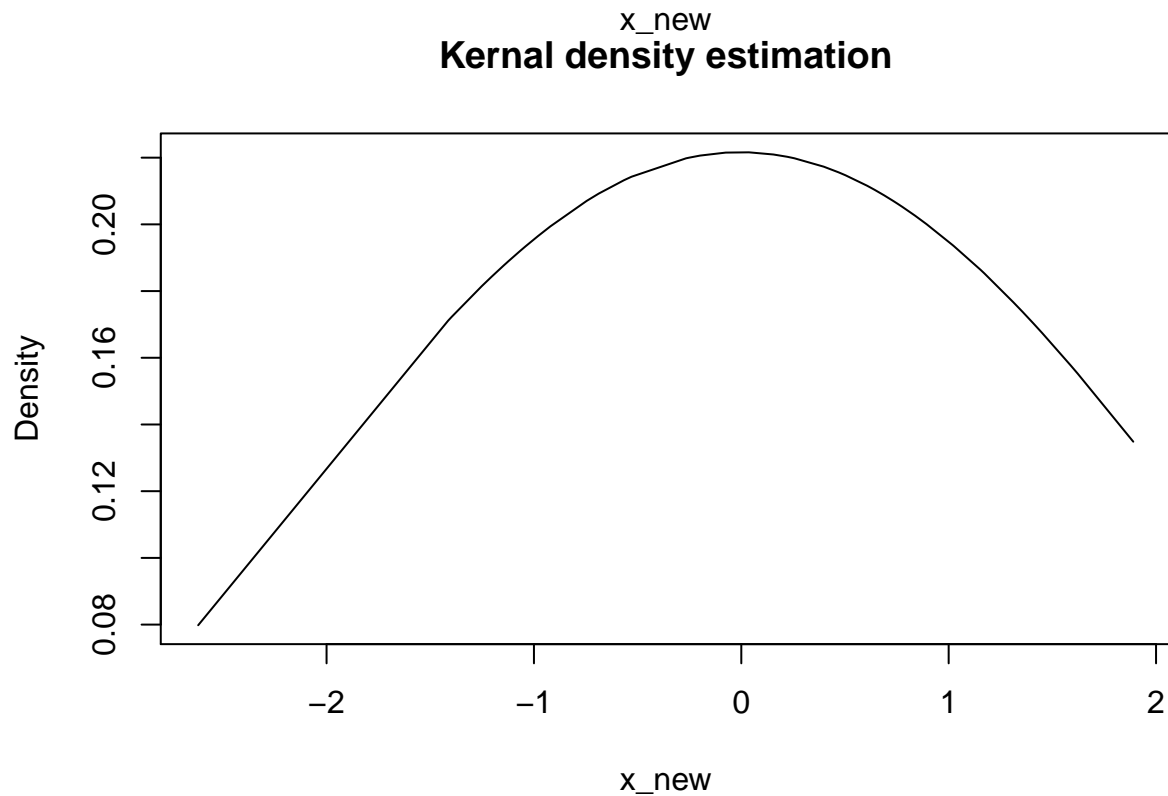
**Kernal density estimation**



### Kernal density estimation



### Kernal density estimation



From the plots above, as bandwidth increases, the estimations become smoother and closer to normal distribution. And when bandwidth increases to a certain value, the estimation remains similar no matter how bandwidth increases.

**2.**

By definition of convex function, for  $t \in [0, 1]$

$$\begin{aligned} f(tx + (1-t)y) &\leq tf(x) + (1-t)f(y) \\ g(tx + (1-t)y) &\leq tg(x) + (1-t)g(y) \end{aligned}$$

Then

$$\begin{aligned} h(tx + (1-t)y) &= f(tx + (1-t)y) + g(tx + (1-t)y) \\ &\leq tf(x) + (1-t)f(y) + tg(x) + (1-t)g(y) \\ &= t(f(x) + g(x)) + (1-t)(f(y) + g(y)) \\ &= th(x) + (1-t)h(y) \end{aligned}$$

Therefore,  $h = f + g$  is convex.

**3.**

we want to prove that  $f(x) = |x|$  is convex. For  $t \in [0, 1]$

$$\begin{aligned} f(x) &= |x| \\ f(tx + (1-t)y) &= |tx + (1-t)y| \\ &\leq |tx| + |(1-t)y| \\ &= f(tx) + f((1-t)y) \\ &= tf(x) + (1-t)f(y) \end{aligned}$$

Thus,  $f(x) = |x|$  is convex.

the other question,  $\mathcal{L}_1$  norm:

$$|x|_1 = \sum_{i=1}^n |x_i|$$

By the last question, it's also convex.

**4.**

$l_2$  norm

$$\begin{aligned} f(x) &= x^2 \\ f(tx + (1-t)y) &= t^2x^2 + (1-t)^2y^2 + 2t(1-t)xy \\ tf(x) + (1-t)f(y) &= tx^2 + (1-t)y^2 \\ f(tx + (1-t)y) - [tf(x) + (1-t)f(y)] &= t(t-1)x^2 + t(t-1)y^2 + 2t(1-t)xy \\ &= t(1-t)(-x^2 - y^2 + 2xy) \\ &= t(t-1)(x-y)^2 \\ &\leq 0 \quad \text{since } 0 \leq t \leq 1 \\ f(tx + (1-t)y) &\leq tf(x) + (1-t)f(y) \end{aligned}$$

Therefore,  $f(x) = x^2$  is convex. Objective function of elastic net is

$$\frac{1}{2n} \|y - Xb\|_2^2 + \lambda[(1-\alpha)\frac{1}{2}\|b\|_2^2 + \alpha\|b\|_1]$$

It's a sum of  $l_2$  norm and  $l_1$  norm, from the previous two exercises, it's convex

## 5.

According to textbook page 189:

```
library(glmnet)
```

```
## Loading required package: Matrix
```

```
## Loading required package: foreach
```

```
## Loaded glmnet 2.0-16
```

```
# KKT check function
```

```
#Return:
```

```
#A logical vector indicating where the KKT conditions have been violated by the variables that are curr
```

```
check_kkt <- function(y, X, b, lambda) {
```

```
  resid <- y - X %*% b
```

```
  s <- apply(X, 2, function(xj) crossprod(xj, resid)) / lambda / nrow(X)
```

```
  (b == 0) & (abs(s) >= 1)
```

```
}
```

```
##use iris as dataset
```

```
x <- scale(model.matrix(Sepal.Length ~. -1, iris))
```

```
y <- iris[,1]
```

```
#implement lasso regression
```

```
#Check which variables violate KKT condition
```

```
#Return: logical vector indicating where the KKT conditions have been violated
```

```
lasso_reg_with_screening <- function(x, y){
```

```
  m1 <- cv.glmnet(x,y,alpha=1)
```

```
  #use 1se as the criteria to choose lambda
```

```
  lambda <- m1$lambda.1se
```

```
  b <- m1$glmnet.fit$beta[, m1$lambda == lambda]
```

```
  check_kkt(y, x, b, lambda)
```

```
}
```

```
lasso_reg_with_screening(x,y)
```

```
##      Sepal.Width      Petal.Length      Petal.Width      Speciessetosa
```

```
##              FALSE              FALSE              FALSE              FALSE
```

```
## Speciesversicolor Speciesvirginica
```

```
##              FALSE              FALSE
```

We can observe that all coefficients get a false result for KKT violation, which indicates that no KKT violation