

# Machine Learning HW#5

Evgeny Marshakov

## Problem 1

- By definition

$$\hat{f}_n(x) = \frac{1}{nh} \sum_{i=1}^n K\left(\frac{x - x_i}{h}\right) \quad (1)$$

So it can be easily seen that

$$\mathbb{E}\hat{f}_n(x) = \mathbb{E}\left[\frac{1}{nh} \sum_{i=1}^n K\left(\frac{x - x_i}{h}\right)\right] = \frac{1}{h} \mathbb{E}K\left(\frac{x - x_i}{h}\right) = \frac{1}{h} \int_{x-\frac{h}{2}}^{x+\frac{h}{2}} f(y)dy \quad (2)$$

and

$$\mathbb{V}\hat{f}_n(x) = \mathbb{V}\frac{1}{nh} \sum_{i=1}^n K\left(\frac{x - x_i}{h}\right) = \frac{1}{nh^2} \mathbb{V}\left[K\left(\frac{x - x_i}{h}\right)\right] = \quad (3)$$

$$= \frac{1}{nh^2} \left( \mathbb{E}K^2\left(\frac{x - x_i}{h}\right) - \left[\mathbb{E}K\left(\frac{x - x_i}{h}\right)\right]^2 \right) \quad (4)$$

as  $K$  is an indicator function we can get rid of the square on the first summand of (4), hence we obtain

$$\frac{1}{nh^2} \left( \mathbb{E}K\left(\frac{x - x_i}{h}\right) - \left[\mathbb{E}K\left(\frac{x - x_i}{h}\right)\right]^2 \right) = \frac{1}{nh^2} \left( \int_{x-\frac{h}{2}}^{x+\frac{h}{2}} f(y)dy - \left[ \int_{x-\frac{h}{2}}^{x+\frac{h}{2}} f(y)dy \right]^2 \right) \quad (5)$$

and the result follows.

- To prove this we need to check a few properties:

1. The first one is

$$\lim_{h \rightarrow 0} \mathbb{E} \hat{f}_n(x) = f(x) \quad (6)$$

Indeed,

$$\lim_{h \rightarrow 0} \mathbb{E} \hat{f}_n(x) = \lim_{h \rightarrow 0} \frac{\int_{x-\frac{h}{2}}^{x+\frac{h}{2}} f(y) dy}{h} = \lim_{h \rightarrow 0} \frac{\left( \int_{x-\frac{h}{2}}^{x+\frac{h}{2}} f(y) dy \right)'}{h'} = \quad (7)$$

$$= \frac{1}{2} \lim_{h \rightarrow 0} \left[ f\left(x + \frac{h}{2}\right) + f\left(x - \frac{h}{2}\right) \right] = f(x) \quad (8)$$

2. The second is

$$\lim_{n, nh \rightarrow \infty, h \rightarrow 0} \mathbb{V} \hat{f}_n(x) = 0 \quad (9)$$

Indeed,

$$\lim_{n, nh \rightarrow \infty, h \rightarrow 0} \mathbb{V} \hat{f}_n(x) = \lim_{n, nh \rightarrow \infty, h \rightarrow 0} \frac{1}{nh^2} \left( \int_{x-\frac{h}{2}}^{x+\frac{h}{2}} f(y) dy - \left[ \int_{x-\frac{h}{2}}^{x+\frac{h}{2}} f(y) dy \right]^2 \right) = \quad (10)$$

$$= \lim_{n, nh \rightarrow \infty, h \rightarrow 0} \left[ \frac{1}{h} \int_{x-\frac{h}{2}}^{x+\frac{h}{2}} f(y) dy \right] \cdot \left[ \frac{1}{nh} - \frac{1}{n} \cdot \frac{1}{h} \int_{x-\frac{h}{2}}^{x+\frac{h}{2}} f(y) dy \right] \quad (11)$$

Using the first property we can simplify (11) as follows

$$\lim_{n, nh \rightarrow \infty} f(x) \left[ \frac{1}{nh} - \frac{1}{n} f(x) \right] = 0 \quad (12)$$

Now we can apply Chebyshev inequality. More precisely, we have the following expression (all conditions of Chebyshev inequality are satisfied)

$$\mathbb{P} \left( |\hat{f}_n(x) - \mathbb{E} \hat{f}_n(x)| > \epsilon \right) \leq \frac{\mathbb{V} \hat{f}_n(x)}{\epsilon^2} \rightarrow 0 \quad (13)$$

And after taking the limit with  $n, nh \rightarrow \infty, h \rightarrow 0$  we obtain

$$\mathbb{P} \left( |\hat{f}_n(x) - f(x)| > \epsilon \right) = \lim_{n, nh \rightarrow \infty, h \rightarrow 0} \mathbb{P} \left( |\hat{f}_n(x) - \mathbb{E} \hat{f}_n(x)| > \epsilon \right) \leq \lim_{n, nh \rightarrow \infty, h \rightarrow 0} \frac{\mathbb{V} \hat{f}_n(x)}{\epsilon^2} = 0 \quad (14)$$

## Problem 2