Homework-1 Solution

October 14, 2018

1 Nearest Neighbors and the Curse of Dimensionality

(a)

$$\mathbf{X}, \mathbf{Y} \sim \mathbf{U}[0, 1]$$

 $\mathbf{Z} = (\mathbf{X} - \mathbf{Y})^2$

Then, we have

$$\mathbb{E}\left[\mathbf{Z}\right] = \mathbb{E}\left[(\mathbf{X} - \mathbf{Y})^{2}\right]$$

$$= \mathbb{E}\left[\mathbf{X}^{2} + \mathbf{Y}^{2} - 2\mathbf{X}\mathbf{Y}\right]$$

$$= \mathbb{E}\left[\mathbf{X}^{2}\right] + \mathbb{E}\left[\mathbf{Y}^{2}\right] - \mathbb{E}\left[2\mathbf{X}\mathbf{Y}\right]$$
(1)

and

$$Var [\mathbf{Z}] = \mathbb{E}[\mathbf{Z}^{2}] - \mathbb{E}[\mathbf{Z}]^{2}$$

$$= \mathbb{E}[\mathbf{X}^{4} - 4\mathbf{X}^{3}\mathbf{Y} + 6\mathbf{X}^{2}\mathbf{Y}^{2} - 4\mathbf{X}\mathbf{Y}^{3} + \mathbf{Y}^{4}] + \mathbb{E}[\mathbf{X}^{2} + \mathbf{Y}^{2} - 2\mathbf{X}\mathbf{Y}]^{2}$$

$$= \mathbb{E}[\mathbf{X}^{4}] - 4\mathbb{E}[\mathbf{X}^{3}] \mathbb{E}[\mathbf{Y}] + 6\mathbb{E}[\mathbf{X}^{2}] \mathbb{E}[\mathbf{Y}^{2}] - 4\mathbb{E}[\mathbf{X}] \mathbb{E}[\mathbf{Y}^{3}] + \mathbb{E}[\mathbf{Y}^{4}]$$

$$+ (\mathbb{E}[\mathbf{X}^{2}] + \mathbb{E}[\mathbf{Y}^{2}] - 2\mathbb{E}[\mathbf{X}] \mathbb{E}[\mathbf{Y}])^{2}.$$
(2)

Given that both \mathbf{X} and \mathbf{Y} are sampled uniformly form unit interval, we can get the n-th moment of \mathbf{X} and \mathbf{Y} as follows,

$$\mathbb{E}\left[\mathbf{X}\right] = \mathbb{E}\left[\mathbf{Y}\right] = \int_0^1 x \, dx = \frac{1}{2}$$

$$\mathbb{E}\left[\mathbf{X}^2\right] = \mathbb{E}\left[\mathbf{Y}^2\right] = \int_0^1 x^2 \, dx = \frac{1}{3}$$

$$\mathbb{E}\left[\mathbf{X}^3\right] = \mathbb{E}\left[\mathbf{Y}^3\right] = \int_0^1 x^3 \, dx = \frac{1}{4}$$

$$\mathbb{E}\left[\mathbf{X}^4\right] = \mathbb{E}\left[\mathbf{Y}^4\right] = \int_0^1 x^4 \, dx = \frac{1}{5}$$

Plugging the above results of moment into Eqn. (1) and Eqn. (2), we have

$$\mathbb{E}\left[\mathbf{Z}\right] = \frac{1}{6}$$

$$\operatorname{Var}\left[\mathbf{Z}\right] = \frac{7}{180}$$
(3)

(b) Since $\mathbf{X}_1, \mathbf{X}_2, ..., \mathbf{X}_d$ and $\mathbf{Y}_1, \mathbf{Y}_2, ..., \mathbf{Y}_d$ are independently sampled from uniform distribution, we have

$$\mathbb{E}\left[\mathbf{R}\right] = \sum_{1}^{d} \mathbb{E}\left[\mathbf{Z}_{i}\right] = d\mathbb{E}\left[\mathbf{Z}\right]$$

$$\operatorname{Var}\left[\mathbf{R}\right] = \sum_{1}^{d} \operatorname{Var}\left[\mathbf{Z}_{i}\right] = d\operatorname{Var}\left[\mathbf{Z}\right]$$
(4)

(c) From part (b), we get the expectation and variance of \mathbf{R} , it is then straightforward to get the standard deviation as follows:

$$Std\left[\mathbf{R}\right] = \sqrt{d} Std\left[\mathbf{Z}\right]. \tag{5}$$

Now, we conclude that the expectation increases linearly along with the dimension d while the standard deviation increases with the rate \sqrt{d} . That's why in high dimensions, "most points are far away, and approximately the same distance".