Exercise of Supervised Learning: Curse of Dimensionality

Yawei Li

yawei.li@stat.uni-muenchen.de

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Exercise 1

Consider a random vector $X = (X_1, \dots, X_p)^T \sim \mathcal{N}(0, I)$, i.e., a multivariate normally distributed vector with mean vector zero and covariance matrix beging the identity matrix of dimension $p \times p$. In this case, the coordinates X_1, \dots, X_p are i.i.d. each with distribution $\mathcal{N}(0, 1)$.

(a) Show that $\mathbb{E}[||X||_2^2] = p$ and $\operatorname{Var}(||X||_2^2) = 2p$, where $||\cdot||_2$ is the Euclidean norm. Hint: $\mathbb{E}_{Y \sim \mathcal{N}(0,1)}(Y^4) = 3$.

Note that

$$||X||_2^2 = \sum_{i=1}^p X_i^2.$$

$$\mathbb{E}[||X||_2^2] = \mathbb{E}\left[\sum_{i=1}^p X_i^2\right]$$

Note that

$$||X||_2^2 = \sum_{i=1}^p X_i^2.$$

$$\mathbb{E}[||X||_2^2] = \mathbb{E}\left[\sum_{i=1}^p X_i^2\right]$$
$$= \sum_{i=1}^p \mathbb{E}[X_i^2]$$

Note that

$$||X||_2^2 = \sum_{i=1}^p X_i^2.$$

$$\mathbb{E}[||X||_2^2] = \mathbb{E}\left[\sum_{i=1}^{\rho} X_i^2\right]$$

$$= \sum_{i=1}^{\rho} \mathbb{E}[X_i^2]$$

$$= \sum_{i=1}^{\rho} (\underbrace{\mathbb{E}[X_i]^2}_{-0} + \operatorname{Var}(X_i))$$

Note that

$$||X||_2^2 = \sum_{i=1}^p X_i^2.$$

$$\mathbb{E}[||X||_2^2] = \mathbb{E}\left[\sum_{i=1}^{p} X_i^2\right]$$

$$= \sum_{i=1}^{p} \mathbb{E}[X_i^2]$$

$$= \sum_{i=1}^{p} (\underbrace{\mathbb{E}[X_i]^2}_{=0} + \operatorname{Var}(X_i))$$

$$= \sum_{i=1}^{p} 1$$

Note that

$$||X||_2^2 = \sum_{i=1}^p X_i^2.$$

$$\mathbb{E}[||X||_2^2] = \mathbb{E}\left[\sum_{i=1}^{\rho} X_i^2\right]$$

$$= \sum_{i=1}^{\rho} \mathbb{E}[X_i^2]$$

$$= \sum_{i=1}^{\rho} (\underbrace{\mathbb{E}[X_i]^2}_{=0} + \operatorname{Var}(X_i))$$

$$= \sum_{i=1}^{\rho} 1$$

$$= \rho.$$

Solution to Question 1 (a): Continued

$$Var(||X||_{2}^{2}) = Var(\sum_{i=1}^{p} X_{i}^{2})$$

$$= \sum_{i=1}^{p} Var(X_{i}^{2})$$

$$= \sum_{i=1}^{p} (\underbrace{\mathbb{E}[X_{i}^{4}]}_{=3} - \underbrace{\mathbb{E}[X_{i}^{2}]^{2}}_{=1}) \qquad \triangleright$$

$$= \sum_{i=1}^{p} (3-1)$$

$$= 2p.$$

Exercise 1 (b)

(b) Use (a) to infer that $|\mathbb{E}[||X||_2 - \sqrt{p}]| \leq \frac{1}{\sqrt{p}}$ by using the following steps:

(i) Write
$$||X||_2 - \sqrt{p} = \underbrace{\frac{||X||_2 - p}{2\sqrt{p}}}_{:=(1)} - \underbrace{\frac{(||X||_2^2 - p)^2}{2\sqrt{p}(||X||_2 + \sqrt{p})^2}}_{:=(2)}$$
.

- (ii) Compute $\mathbb{E}[(1)]$.
- (iii) Note that $0 \leq \mathbb{E}[(2)] \leq \frac{\mathrm{Var}(||X||_2^2)}{2\rho^{3/2}}$ holds due to $||X||_2 \geq 0$.
- (iv) Put (i)- (iii) together.

Step (i): Write
$$||X||_2 - \sqrt{p} = \underbrace{\frac{||X||_2 - p}{2\sqrt{p}}}_{:=(1)} - \underbrace{\frac{(||X||_2^2 - p)^2}{2\sqrt{p}(||X||_2 + \sqrt{p})^2}}_{:=(2)}.$$

Step (ii)

$$\mathbb{E}[(1)] = \mathbb{E}\left[\frac{||X||_2^2 - p}{2\sqrt{p}}\right]$$

$$= \frac{1}{2\sqrt{p}}\left(\underbrace{\mathbb{E}[||X||_2^2]}_{=p \text{ (from Question(a))}} - p\right)$$

$$= 0.$$

Step (i): Write
$$||X||_2 - \sqrt{p} = \underbrace{\frac{||X||_2 - p}{2\sqrt{p}}}_{:=(1)} - \underbrace{\frac{(||X||_2^2 - p)^2}{2\sqrt{p}(||X||_2 + \sqrt{p})^2}}_{:=(2)}.$$

Step (ii):

$$\mathbb{E}[(1)] = \mathbb{E}\left[\frac{||X||_2^2 - \rho}{2\sqrt{\rho}}\right]$$

$$= \frac{1}{2\sqrt{\rho}}(\underbrace{\mathbb{E}[||X||_2^2]}_{=\rho \text{ (from Question(a))}} - \rho)$$

$$= 0.$$

Step (iii): Prove that $0 \le \mathbb{E}[(2)] \le \frac{\mathrm{Var}(||X||_2^2)}{2p^{3/2}}$ holds due to $||X||_2 \ge 0$.

$$\mathbb{E}[(2)] = \mathbb{E}\left[\frac{(||X||_2^2 - p)^2}{2\sqrt{p}(||X||_2 + \sqrt{p})^2}\right] \ge 0, \quad \text{since all the terms are non-negative.}$$

Besides, since $||X||_2 \ge 0$, it follows that

$$(2) \le \frac{(||X||_2^2 - \rho)^2}{2\rho^{3/2}}$$

$$\Rightarrow \quad \mathbb{E}[(2)] \le \mathbb{E}\left[\frac{(||X||_2^2 - \rho)^2}{2\rho^{3/2}}\right] = \frac{1}{2\rho^{3/2}} \cdot \mathbb{E}\left[(||X||_2^2 - \mathbb{E}[||X||_2^2])^2\right] = \frac{\operatorname{Var}(||X||_2^2)}{2\rho^{3/2}}$$

$$= \frac{2\rho}{2\rho^{3/2}} = \frac{1}{\sqrt{\rho}},$$

where we utilize the lemma from (a) that $\mathbb{E}[||X||_2^2]=p$ and $\mathrm{Var}(||X||_2^2)=2p$

Step (iii): Prove that $0 \le \mathbb{E}[(2)] \le \frac{\operatorname{Var}(||X||_2^2)}{2p^{3/2}}$ holds due to $||X||_2 \ge 0$.

$$\mathbb{E}[(2)] = \mathbb{E}\left[\frac{(||X||_2^2 - \rho)^2}{2\sqrt{\rho}(||X||_2 + \sqrt{\rho})^2}\right] \ge 0, \qquad \text{since all the terms are non-negative}.$$

Besides, since $||X||_2 \ge 0$, it follows that

$$(2) \le \frac{(||X||_2^2 - p)^2}{2p^{3/2}}$$

$$\Rightarrow \quad \mathbb{E}[(2)] \le \mathbb{E}\left[\frac{(||X||_2^2 - p)^2}{2p^{3/2}}\right] = \frac{1}{2p^{3/2}} \cdot \mathbb{E}\left[(||X||_2^2 - \mathbb{E}[||X||_2^2])^2\right] = \frac{\operatorname{Var}(||X||_2^2)}{2p^{3/2}}$$

$$= \frac{2p}{2p^{3/2}} = \frac{1}{\sqrt{p}},$$

where we utilize the lemma from (a) that $\mathbb{E}[||X||_2^2]=p$ and $\mathrm{Var}(||X||_2^2)=2p$.

Step (iii): Prove that $0 \le \mathbb{E}[(2)] \le \frac{\operatorname{Var}(||X||_2^2)}{2p^{3/2}}$ holds due to $||X||_2 \ge 0$.

$$\mathbb{E}[(2)] = \mathbb{E}\left[\frac{(||X||_2^2 - \rho)^2}{2\sqrt{\rho}(||X||_2 + \sqrt{\rho})^2}\right] \ge 0, \qquad \text{since all the terms are non-negative}.$$

Besides, since $||X||_2 \ge 0$, it follows that

$$(2) \leq \frac{(||X||_2^2 - \rho)^2}{2\rho^{3/2}}$$

$$\Rightarrow \quad \mathbb{E}[(2)] \leq \mathbb{E}\left[\frac{(||X||_2^2 - \rho)^2}{2\rho^{3/2}}\right] = \frac{1}{2\rho^{3/2}} \cdot \mathbb{E}\left[(||X||_2^2 - \mathbb{E}[||X||_2^2])^2\right] = \frac{\operatorname{Var}(||X||_2^2)}{2\rho^{3/2}}$$

$$= \frac{2\rho}{2\rho^{3/2}} = \frac{1}{\sqrt{\rho}},$$

where we utilize the lemma from (a) that $\mathbb{E}[||X||_2^2] = p$ and $\operatorname{Var}(||X||_2^2) = 2p$.

Step (iv): Putting everything together:

$$|\mathbb{E}[||X||_2 - \sqrt{\rho}]| = |\underbrace{\mathbb{E}[(1)]}_{=0} - \underbrace{\mathbb{E}[(2)]}_{>0}| = \mathbb{E}[(2)] \le \frac{1}{\sqrt{\rho}}.$$

Exercise 1 (c)

- (c) Use (b) to infer that $Var(||X||_2) \le 2$ by using the following steps:
 - (i) Write $Var(||X||_2) = Var(||X||_2 \sqrt{p})$.
 - (ii) For any random variable Y it holds that $\operatorname{Var}(Y) \leq \mathbb{E}[Y^2]$.
- (iii) If you encounter the term $E[||X||_2]$ write it as $\mathbb{E}[\underbrace{||X||_2 \sqrt{p}}_{=(*)} + \sqrt{p}]$ and use (b) for (*).

Step (i): Write
$$Var(||X||_2) = Var(||X||_2 - \sqrt{p})$$
.

It holds because variance does not change by constant shifts.

Step (ii): For any random variable Y it holds that $Var(Y) \leq \mathbb{E}[Y^2]$.

It holds because $\operatorname{Var}(Y) + \mathbb{E}[Y]^2 = \mathbb{E}[Y^2]$ and $\mathbb{E}[Y]^2 \geq 0$. Later we will use this inequality.

$$\begin{split} \operatorname{Var}(||X||_2) &= \operatorname{Var}(||X||_2 - \sqrt{p}) \quad \text{Step (i)} \\ &\leq \mathbb{E}[(||X||_2 - \sqrt{p})^2] \quad \text{Step (ii)} \\ &= \mathbb{E}[||X||_2^2 - 2\sqrt{p}||X||_2 + p] \\ &= \underbrace{\mathbb{E}[||X||_2^2]}_{=p} - 2\sqrt{p} \cdot \mathbb{E}[||X||_2] + p \\ &= 2p - 2\sqrt{p} \cdot \mathbb{E}[||X||_2] \\ &= 2p - 2\sqrt{p} \cdot \mathbb{E}[||X||_2 - \sqrt{p} + \sqrt{p}] \quad \text{Step (iv)} \\ &= 2p - 2p - 2\sqrt{p} \cdot \mathbb{E}[||X||_2 - \sqrt{p}] \\ &= -2\sqrt{p} \cdot \underbrace{\mathbb{E}[||X||_2 - \sqrt{p}]}_{\leq \frac{1}{\sqrt{p}} \quad \text{(from (b))}} \\ &\leq 2\sqrt{p} \cdot \frac{1}{\sqrt{p}} = 2. \end{split}$$

Question 1 (d)

Now let $X' = (X'_1, \dots, X'_p)^T \sim \mathcal{N}(0, \textbf{I})$ be another multivariate normally distributed vector with mean vector zero and covariance matrix being the identity matrix of dimension $p \times p$. Further, assume that X and X' are independent, so that $Z := \frac{X - X'}{\sqrt{2}} \sim \mathcal{N}(0, \textbf{I})$. Conclude from the previous that

$$\left|\mathbb{E}\left[||X-X'||_2-\sqrt{2p}\right]\right|\leq \frac{2}{p}$$
 and $\operatorname{Var}(||X-X'||_2)\leq 4$.

We first investigate Z. Since $Z = \frac{X - X'}{\sqrt{2}} \sim \mathcal{N}(0, I)$, it follows from (b) and (c) that

$$|\mathbb{E}[||Z||_2 - \sqrt{p}]| \le \sqrt{\frac{1}{p}},\tag{1}$$

$$\operatorname{Var}(||Z||_2) \le 2 \tag{2}$$

But the norm of 7

$$||Z||_{2} = \sqrt{\sum_{i=1}^{p} \left(\frac{X_{i} - X_{i}'}{\sqrt{2}}\right)^{2}} = \sqrt{\frac{1}{2} \sum_{i=1}^{p} (X_{i} - X_{i}')^{2}} = \sqrt{\frac{1}{2}} \sqrt{\sum_{i=1}^{p} (X_{i} - X_{i}')^{2}} = \sqrt{\frac{1}{2}} ||X - X'||_{2}.$$
(3)

It follows from (1) tha

$$\sqrt{2} \cdot |\mathbb{E}[||Z||_2 - \sqrt{p}]| \le \frac{2}{p} \Rightarrow |\mathbb{E}[\sqrt{2} \underbrace{||Z||_2}_{||X - X'||_2} - \sqrt{2p}]| \le \sqrt{\frac{2}{p}}$$

We first investigate Z. Since $Z = \frac{X - X'}{\sqrt{2}} \sim \mathcal{N}(0, I)$, it follows from (b) and (c) that

$$|\mathbb{E}[||Z||_2 - \sqrt{p}]| \le \sqrt{\frac{1}{p}},\tag{1}$$

$$Var(||Z||_2) \le 2 \tag{2}$$

But the norm of Z

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We first investigate Z. Since $Z = \frac{X - X'}{\sqrt{2}} \sim \mathcal{N}(0, I)$, it follows from (b) and (c) that

$$|\mathbb{E}[||Z||_2 - \sqrt{p}]| \le \sqrt{\frac{1}{p}},\tag{1}$$

$$Var(||Z||_2) \le 2 \tag{2}$$

But the norm of Z

$$||Z||_{2} = \sqrt{\sum_{i=1}^{p} \left(\frac{X_{i} - X_{i}'}{\sqrt{2}}\right)^{2}} = \sqrt{\frac{1}{2} \sum_{i=1}^{p} (X_{i} - X_{i}')^{2}} = \sqrt{\frac{1}{2}} \sqrt{\sum_{i=1}^{p} (X_{i} - X_{i}')^{2}} = \sqrt{\frac{1}{2}} ||X - X'||_{2}.$$
(3)

It follows from (1) that

$$\sqrt{2}\cdot |\mathbb{E}[||Z||_2 - \sqrt{p}]| \leq \frac{2}{\rho} \Rightarrow |\mathbb{E}[\sqrt{2}\underbrace{||Z||_2}_{||X-X'||_2} - \sqrt{2p}]| \leq \sqrt{\frac{2}{\rho}}.$$

Moreover, (2):
$$Var(||Z||_2) \le 2$$
 implies that

$$\begin{split} \operatorname{Var}(||Z||_2) &\leq 2 \\ \Leftrightarrow & 2\operatorname{Var}(||Z||_2) \leq 4 \\ \Leftrightarrow & \operatorname{Var}(\sqrt{2}||Z||_2) \leq 4 \quad \quad (\operatorname{Var}(aY) = a^2\operatorname{Var}(Y) \text{ for any RV } Y) \\ \Leftrightarrow & \operatorname{Var}(||X - X'||_2) \leq 4 \quad \quad \text{Using (3) that } ||Z||_2 = \sqrt{\frac{1}{2}}||X - X'||_2. \end{split}$$

Exercise 1 (e)

(e) From the cosine rule we can infer that for any $x, x' \in \mathbb{R}^p$ it holds that

$$\langle x, x' \rangle = \frac{1}{2}(||x||_2^2 + ||x'||_2^2 - ||x - x'||_2^2).$$

Use this to show that $\mathbb{E}[\langle X, X' \rangle] = 0$. Moreover, derive that $\operatorname{Var}(\langle X, X' \rangle) = p$.

Since
$$\langle x, x' \rangle = \frac{1}{2}(||x||_2^2 + ||x'||_2^2 - ||x - x'||_2^2)$$
, we can infer that
$$\mathbb{E}[\langle X, X' \rangle] = \frac{1}{2}(\mathbb{E}[||X||_2^2] + \mathbb{E}[||X'||_2^2] - \mathbb{E}[||X - X'||_2^2])$$

$$= \frac{1}{2}\left(p + p - 2 \cdot \mathbb{E}\left[\underbrace{\frac{1}{2}||X - X'||_2^2}_{=||Z||_2^2}\right]\right)$$

$$= \frac{1}{2}(p + p - 2p) = 0. \qquad \text{(From (a) we know that } \mathbb{E}[||Z||_2^2] = p\text{)}$$

$$\operatorname{Var}(\langle X, X' \rangle) = \operatorname{Var}(\sum_{i=1}^{p} X_{i} X'_{i})$$

$$= \sum_{i=1}^{p} \operatorname{Var}(X_{i} X'_{i})$$

$$= p \operatorname{Var}(X_{1} X'_{1})$$

$$= p \cdot (\mathbb{E}[X_{1}^{2}(X'_{1})^{2}] - \mathbb{E}[X_{1}(X'_{1})]^{2})$$

$$= p \cdot (\mathbb{E}[X_{1}^{2}] \cdot \mathbb{E}[(X'_{1})^{2}] - \mathbb{E}[X_{1}]^{2} \cdot \mathbb{E}[X'_{1}]^{2})$$

$$= p.$$

Exercise 1 (f)

- (f) For different dimensions p, e.g., $p \in \{1, 2, 4, 8, ..., 1024\}$, create two sets consisting of 100 i.i.d. random observations from $\mathcal{N}(0, I)$, respectively and
 - (i) compute the average Euclidean length of (one of) the sampled sets and compare it to \sqrt{p} ;
 - (ii) comptue the average Euclidean distances between the sampled sets and compare it to $\sqrt{2p}$;
- (iii) compute the average inner products between the sampled sets;
- (iv) compute in (i)-(iii) also the empirical variances of the respective terms.

Visualize your results in an appropriate manner.

Show the standard solution.