

University of Edinburgh
INFR11156: Algorithmic Foundations of Data Science (2019)

Solutions 3

Solutions

Problem 1: Project the volume of a d -dimensional ball of radius \sqrt{d} onto a line through the centre. For large d , give an intuitive argument that the projected volume should behave like a Gaussian.

Solution: Projecting the volume of a d -dimensional ball corresponds to assigning a function $f : [-\sqrt{d}, \sqrt{d}]$ such that $f(x)$ is the volume of a $(d-1)$ -dimensional ball with radius $\sqrt{d-x^2}$. This function is 0 on the endpoints of its domain, reaches its maximal value for $x = 0$ and is an even function, i.e., $f(-x) = f(x)$. We will show that, as x is increasing on the interval $[0, \sqrt{d}]$, the function f decreases exponentially in x^2 .

We will assume that $d-1 = 2k$, for some natural number k , as the other case is done similarly. Let $V_n(R)$ denote the volume of a ball in n dimensions of radius R . Then, we know that

$$\begin{aligned} f(x) &= V_{d-1}(\sqrt{d-x^2}) = \frac{\pi^k}{k!} \cdot (\sqrt{d-x^2})^{d-1} \\ &= \frac{\pi^k}{k!} \cdot d^{(d-1)/2} \cdot \left(1 - \frac{x^2}{d}\right)^{(d-1)/2} \\ &\approx \frac{\pi^k}{k!} \cdot d^{(d-1)/2} \cdot e^{-\frac{x^2}{2}}. \end{aligned}$$

Problem 2: Consider a nonorthogonal basis $\mathbf{e}_1, \mathbf{e}_2, \dots, \mathbf{e}_d$. The \mathbf{e}_i are a set of linearly independent unit vectors that span the space.

1. Prove that the representation of any vector in this basis is unique;
2. Calculate the squared length of $\mathbf{z} = \frac{\sqrt{2}}{2}\mathbf{e}_1 + \mathbf{e}_2$ where $\mathbf{e}_1 = (1, 0)$ and $\mathbf{e}_2 = \left(-\frac{\sqrt{2}}{2}, \frac{\sqrt{2}}{2}\right)$;
3. If $\mathbf{y} = \sum_{i=1}^d a_i \mathbf{e}_i$ and $\mathbf{z} = \sum_{i=1}^d b_i \mathbf{e}_i$, with $0 < a_i < b_i$ for all $1 \leq i \leq d$, is it necessarily true that the length of \mathbf{z} is greater than the length of \mathbf{y} ? If yes give a proof of the statement, if no find a counterexample;
4. Consider the basis $\mathbf{e}_1 = (1, 0)$ and $\mathbf{e}_2 = \left(-\frac{\sqrt{2}}{2}, \frac{\sqrt{2}}{2}\right)$.
 - (a) What is the representation of the vector $\mathbf{v} = (0, 1)$ in the basis $(\mathbf{e}_1, \mathbf{e}_2)$? I.e. find scalars a, b such that $\mathbf{v} = a\mathbf{e}_1 + b\mathbf{e}_2$.
 - (b) What is the representation of the vector $\left(\frac{\sqrt{2}}{2}, \frac{\sqrt{2}}{2}\right)$ in the basis $(\mathbf{e}_1, \mathbf{e}_2)$?
 - (c) What is the representation of the vector $(1, 2)$ in the basis $(\mathbf{e}_1, \mathbf{e}_2)$?

Solution:

1. Let \mathbf{v} be an arbitrary vector. Since the vectors $\{\mathbf{e}_i\}$'s form a basis, they span the entire space. So there exists a representation $\mathbf{v} = \sum_{i=1}^d \alpha_i \mathbf{e}_i$. Suppose \mathbf{v} can also be represented as $\sum_{i=1}^d \beta_i \mathbf{e}_i$. Then it holds that

$$\sum_{i=1}^d \alpha_i \mathbf{e}_i = \sum_{i=1}^d \beta_i \mathbf{e}_i$$

which gives us that $\sum_{i=1}^d (\alpha_i - \beta_i) \mathbf{e}_i = \mathbf{0}$. Since the vectors are linearly independent, we must have that $\alpha_i = \beta_i$ for all i . We conclude that the two representations are the same.

2. We have $\mathbf{z} = \frac{\sqrt{2}}{2}\mathbf{e}_1 + \mathbf{e}_2 = \begin{pmatrix} \frac{\sqrt{2}}{2} \\ 0 \end{pmatrix} + \begin{pmatrix} -\frac{\sqrt{2}}{2} \\ \frac{\sqrt{2}}{2} \end{pmatrix} = \begin{pmatrix} 0 \\ \frac{\sqrt{2}}{2} \end{pmatrix}$. Therefore the length of \mathbf{z} is $\|\mathbf{z}\| = \frac{\sqrt{2}}{2}$.
3. We will show that the answer is false. Let $\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_d$ be the standard basis, i.e. \mathbf{u}_i is the vector with 1 in position i and 0 everywhere else. We can rewrite the vectors \mathbf{e}_i , for all i , as follows:

$$\mathbf{e}_i = \sum_{j=1}^d \alpha_{ij} \mathbf{u}_j.$$

Substituting in the representations of \mathbf{y} and \mathbf{z} we see that

$$\begin{aligned} \mathbf{y} &= \sum_{i=1}^d a_i \left(\sum_{j=1}^d \alpha_{ij} \mathbf{u}_j \right) = \sum_{j=1}^d \left(\sum_{i=1}^d a_i \alpha_{ij} \right) \mathbf{u}_j, \\ \mathbf{z} &= \sum_{i=1}^d b_i \left(\sum_{j=1}^d \alpha_{ij} \mathbf{u}_j \right) = \sum_{j=1}^d \left(\sum_{i=1}^d b_i \alpha_{ij} \right) \mathbf{u}_j. \end{aligned}$$

Moreover, we see that the two norms can be expressed as

$$\begin{aligned} \|\mathbf{y}\|^2 &= \sum_{j=1}^d \left(\sum_{i=1}^d a_i \alpha_{ij} \right)^2 \\ \|\mathbf{z}\|^2 &= \sum_{j=1}^d \left(\sum_{i=1}^d b_i \alpha_{ij} \right)^2. \end{aligned}$$

Since the two lengths are positive real numbers, it is sufficient to compare their squared norms. We have that

$$\|\mathbf{z}\|^2 - \|\mathbf{y}\|^2 = \sum_{j=1}^d \left(\left(\sum_{i=1}^d b_i \alpha_{ij} \right)^2 - \left(\sum_{i=1}^d a_i \alpha_{ij} \right)^2 \right) = \sum_{j=1}^d \left(\left(\sum_{i=1}^d \alpha_{ij} (a_i + b_i) \right) \left(\sum_{i=1}^d \alpha_{ij} (b_i - a_i) \right) \right).$$

For certain choices of the numbers α_{ij} , $0 < a_i < b_i$, the right hand side can be negative. For example, take

$$\alpha = \frac{1}{\sqrt{5}} \begin{pmatrix} 2 & 1 \\ -1 & -2 \end{pmatrix}, \quad a_1 = 0.9, \quad a_2 = 0.1, \quad b_1 = b_2 = 1.$$

4. Suppose $\mathbf{v} = a\mathbf{e}_1 + b\mathbf{e}_2$. We substitute the values for the three vectors and solve for a and b as follows:

(a)

$$\begin{pmatrix} 0 \\ 1 \end{pmatrix} = a \begin{pmatrix} 1 \\ 0 \end{pmatrix} + b \begin{pmatrix} -\frac{\sqrt{2}}{2} \\ \frac{\sqrt{2}}{2} \end{pmatrix} \Leftrightarrow \begin{cases} a - \frac{\sqrt{2}}{2}b = 0 \\ \frac{\sqrt{2}}{2}b = 1 \end{cases} \Leftrightarrow \begin{cases} a = 1 \\ b = \sqrt{2} \end{cases}$$

(b) $a = \sqrt{2}, b = 1$.

(c) $a = 3, b = 2\sqrt{2}$.

Problem 3: Compute the right-singular vectors v_i , the left-singular vectors u_i , the singular values σ_i and hence find the *Singular value decomposition* of

$$1. A = \begin{pmatrix} 1 & 1 \\ 0 & 3 \\ 3 & 0 \end{pmatrix};$$

$$2. A = \begin{pmatrix} 0 & 2 \\ 2 & 0 \\ 1 & 3 \\ 3 & 1 \end{pmatrix}.$$

Solution: Throughout the solution we will make use of the following lemma.

Lemma 1. Let a and b be two real numbers satisfying $a^2 + b^2 = 1$ and $a \geq 0$. The product ab is maximised when $a = b = \frac{\sqrt{2}}{2}$.

Proof. Using the initial conditions we can rewrite $a = \sqrt{1 - b^2}$. Hence maximising the product ab reduces to maximising the function $f(x) = x\sqrt{1 - x^2}$. A point x_0 maximises $f(x)$ if $x_0 \geq 0$ and $f'(x_0) = 0$. We have that

$$f'(x) = \sqrt{1 - x^2} + \frac{-x^2}{\sqrt{1 - x^2}} = \frac{1 - 2x^2}{\sqrt{1 - x^2}}.$$

We conclude that $x_0 = \frac{\sqrt{2}}{2}$ which gives $a = b = \frac{\sqrt{2}}{2}$. □

1. For finding the first right-singular vector v_1 , we look at any vector $v = \begin{pmatrix} a \\ b \end{pmatrix}$ such that $\|v\| = 1$ and v maximises $\|Av\|$. Without loss of generality we can also assume that $a \geq 0$. Firstly, note that maximising $\|Av\|$ is equivalent to maximising $\|Av\|^2$. We also have that:

$$\|Av\|^2 = \left\| \begin{pmatrix} a+b \\ 3b \\ 3a \end{pmatrix} \right\|^2 = (a+b)^2 + 9b^2 + 9a^2.$$

Since $\|v\| = 1$, we have that $a^2 + b^2 = 1$. Therefore $\|Av\|^2 = 10(a^2 + b^2) + 2ab = 10 + 2ab$. We see that $\|Av\|^2$ is maximised if and only if ab is maximised. Using Lemma 1 that happens when $a = b = \frac{1}{\sqrt{2}}$. So the first right-singular vector $v_1 = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ 1 \end{pmatrix}$ and the first singular value is $\sigma_1 = \|Av_1\| = \sqrt{11}$. For the first left-singular vector u_1 we compute

$$u_1 = \frac{1}{\sigma_1} Av_1 = \frac{1}{\sqrt{22}} \begin{pmatrix} 2 \\ 3 \\ 3 \end{pmatrix}.$$

For the second right-singular vector v_2 , we look at vectors $v = \begin{pmatrix} a' \\ b' \end{pmatrix}$ such that $\|v\| = 1$, $v \perp v_1$ and v maximises $\|Av\|$. Without loss of generality we can assume $a' \geq 0$. Since $v \perp v_1$ this implies that $a' + b' = 0$. Solving $a'^2 + b'^2 = 1$ gives us that $a' = \frac{1}{\sqrt{2}}$. Hence $v_2 = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ -1 \end{pmatrix}$. Moreover, the second singular value is $\sigma_2 = \|Av_2\| = 3$. The second left-singular vector u_2 is computed by

$$u_2 = \frac{1}{\sigma_2} Av_2 = \frac{1}{\sqrt{2}} \begin{pmatrix} 0 \\ -1 \\ 1 \end{pmatrix}.$$

The singular value decomposition of A is

$$A = UDV^T = \begin{pmatrix} \frac{2}{\sqrt{22}} & 0 \\ \frac{3}{\sqrt{22}} & \frac{-1}{\sqrt{2}} \\ \frac{3}{\sqrt{22}} & \frac{1}{\sqrt{2}} \end{pmatrix} \begin{pmatrix} \sqrt{11} & 0 \\ 0 & 3 \end{pmatrix} \begin{pmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \end{pmatrix}.$$

2. Again, for finding v_1 we look at any vector $v = \begin{pmatrix} a \\ b \end{pmatrix}$ such that $\|v\| = 1$ and v maximises $\|Av\|$. Without loss of generality we can assume $a \geq 0$.

$$\|Av\|^2 = \left\| \begin{pmatrix} 2b \\ 2a \\ a+3b \\ 3a+3b \end{pmatrix} \right\|^2 = 4a^2 + 4b^2 + (a+3b)^2 + (3a+b)^2 = 14(a^2 + b^2) + 12ab.$$

Using that $\|v\| = 1$ we have that $\|Av\|^2 = 14 + 12ab$ which, by Lemma 1, is maximised for $a = b = \frac{1}{\sqrt{2}}$. Therefore we have that $v_1 = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ 1 \end{pmatrix}$ and $\sigma_1 = \sqrt{20}$. The first left-singular vector u_1 is given by

$$u_1 = \frac{1}{\sigma_1} Av_1 = \frac{1}{2\sqrt{10}} \begin{pmatrix} 2 \\ 2 \\ 4 \\ 4 \end{pmatrix} = \frac{1}{\sqrt{10}} \begin{pmatrix} 1 \\ 1 \\ 2 \\ 2 \end{pmatrix}.$$

A similar reasoning to the previous part tells us that $v_2 = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ -1 \end{pmatrix}$ and $\sigma_2 = \|Av_2\| = \sqrt{8}$. We also have that

$$u_2 = \frac{1}{\sigma_2} Av_2 = \frac{1}{4} \begin{pmatrix} -2 \\ 2 \\ -2 \\ 2 \end{pmatrix} = \frac{1}{2} \begin{pmatrix} -1 \\ 1 \\ -1 \\ 1 \end{pmatrix}.$$

Hence, the singular value decomposition of A is

$$A = UDV^T = \begin{pmatrix} \frac{1}{\sqrt{10}} & -\frac{1}{2} \\ \frac{1}{\sqrt{10}} & \frac{1}{2} \\ \frac{2}{\sqrt{10}} & -\frac{1}{2} \\ \frac{2}{\sqrt{10}} & \frac{1}{2} \end{pmatrix} \begin{pmatrix} \sqrt{20} & 0 \\ 0 & \sqrt{8} \end{pmatrix} \begin{pmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \end{pmatrix}.$$