

TECHNISCHE UNIVERSITÄT MÜNCHEN

SUMMARY OF THE LECTURE MA4800

Foundations in Data Analysis

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1 Linear Algebra Review

- We work on $\mathbb{K} \in \{\mathbb{R}, \mathbb{C}\}$.
- $A^H = \overline{(A^T)}$.
- A Hermitian matrix A satisfies $A = A^H$.
- $A^{(i)}$ are rows and $A_{(j)}$ are the columns.
- $A^{(i)} = (a_{ij})_{j \in J}$ and $A_{(j)} = (a_{ij})_{i \in I} = (A^T)^{(j)}$
- The matrix-vector product between $A \in \mathbb{K}^{I \times J}$ and $x \in \mathbb{K}^J$ results in the vector in $Ax \in \mathbb{K}^I$ with entries

1.1 Matrices

$$(Ax)_i = \sum_{j \in J} a_{ij} x_j.$$

1.2 Matrix multiplication

The matrix-matrix product between $A \in \mathbb{K}^{I \times J}$ and $B \in \mathbb{K}^{J \times L}$ yields the matrix in $\mathbb{K}^{I \times L}$ with entries

$$(AB)_{i\ell} = \sum_{j \in J} A_{ij} B_{j\ell}.$$

2 The Singular Value Decomposition

2.1 The leading singular vector

2.2 Principal components

2.3 Further singular vectors

2.4 Best k-rank approximation

2.5 The power method

Lemma 2.1 *Let $x \in \mathbb{R}^d$ be a unit d -dimensional vector of components $x = (x_1, \dots, x_d)$ with respect to the canonical basis and picked uniformly at random from the sphere $\{x : \|x\|_2 = 1\}$. The probability that $|x_1| \geq \alpha > 0$ is at least $1 - C\alpha\sqrt{d}$ for some absolute constant.*

Proof

We want the probability of y picked uniformly at random from

$$B^d(1) = \{y \in \mathbb{R}^d, \|y\|_2 \leq 1\}$$

satisfies $|y_1| > \alpha$. In other words, we are looking for the fraction of $B^d(1)$ that satisfies $|y_1| > \alpha$. This corresponds to

$$V_\alpha := \text{Vol}(B^d(1) \cap \{y : |y_1| \leq \alpha\})$$

$$= \int_{y \in B^d(1) \cap \{y : |y_1| \leq \alpha\}} 1 dy$$

$$\begin{aligned}
&= \int_{-\alpha}^{\alpha} \left(\int_{\mathbb{R}^{d-1}} 1_{y_2^2 + \dots + y_d^2 \leq 1 - y_1^2} dy_2 \dots dy_d \right) dy_1 \\
&= \int_{-\alpha}^{\alpha} \text{Vol} \left(B^{d-1} \left(\sqrt{1 - y_1^2} \right) \right) dy_1
\end{aligned}$$

Replacing $\text{Vol} \left(B^{d-1} \left(\sqrt{1 - y_1^2} \right) \right)$ with $(\sqrt{1 - y_1^2})^{d-1} \text{Vol} (B^{d-1}(1))$ since the volume the unit ball with a factor proportional to radius in the power of $d - 1$.

$$\begin{aligned}
&= \int_{-\alpha}^{\alpha} (\sqrt{1 - y_1^2})^{d-1} \text{Vol} (B^{d-1}(1)) dy_1 \\
&= \text{Vol} (B^{d-1}(1)) \int_{-\alpha}^{\alpha} (1 - y_1^2)^{(d-1)/2} dy_1
\end{aligned}$$

In the integral part, $\int_{-\alpha}^{\alpha} (1 - y_1^2)^{(d-1)/2} dy_1$, notice that $(1 - y_1^2)^{(d-1)/2} < 1$ in the whole integration domain. Thus we can write

$$\begin{aligned}
&= \text{Vol} (B^{d-1}(1)) \int_{-\alpha}^{\alpha} (1 - y_1^2)^{(d-1)/2} dy_1 \\
&\leq \text{Vol} (B^{d-1}(1)) \int_{-\alpha}^{\alpha} 1 dy_1 \\
&= 2\alpha \text{Vol} (B^{d-1}(1))
\end{aligned}$$

Recall that volume of unit ball in d dimensions is asymptotically

$$V_1 = \frac{1}{\sqrt{d\pi}} \left(\frac{2\pi e}{d} \right)^{d/2}$$

Hence the probability $p = \Pr(\alpha \leq |y_1|)$ we are interested in satisfies asymptotically

$$p = \frac{V_{\alpha}}{V_1} \propto \frac{2\alpha \frac{1}{\sqrt{(d-1)\pi}} \left(\frac{2\pi e}{d-1} \right)^{(d-1)/2}}{\frac{1}{\sqrt{d\pi}} \left(\frac{2\pi e}{d} \right)^{d/2}} = \frac{2\alpha \frac{1}{\sqrt{(d-1)\pi}} \left(\frac{2\pi e}{d-1} \right)^{(d-1)/2}}{\frac{1}{\sqrt{d\pi}} \left(\frac{2\pi e}{d} \right)^{(d-1)/2} \left(\frac{2\pi e}{d} \right)^{1/2}}$$

We simplify the last term

$$\begin{aligned}
&= 2\alpha * \left(\frac{d}{d-1} \right)^{1/2} * \left(\frac{d}{d-1} \right)^{(d-1)/2} * \left(\frac{d}{2\pi e} \right)^{1/2} \\
&= 2\alpha * \left(\frac{d}{\sqrt{2\pi e(d-1)}} \right) * \left(\frac{d}{d-1} \right)^{(d-1)/2}
\end{aligned}$$

Since $\frac{d}{d-1} = 1 + \frac{1}{d-1}$

$$= 2\alpha * \left(\frac{d}{\sqrt{2\pi e(d-1)}} \right) * \left(1 + \frac{1}{d-1} \right)^{(d-1)/2}$$

We modify the power of the same term, to show it as

$$= 2\alpha * \left(\frac{d}{\sqrt{2\pi e(d-1)}} \right) * \left(\left(1 + \frac{1}{d-1} \right)^{(d-1)} \right)^{1/2}$$

Recall that

$$e = \lim_{n \rightarrow \infty} (1 + 1/n)^n$$

Thus this term is bounded with \sqrt{e}

$$\leq 2\alpha * \left(\frac{d}{\sqrt{2\pi e(d-1)}} \right) * \sqrt{e}$$

We reformulate as

$$= \alpha\sqrt{d} \sqrt{\frac{2d}{\pi(d-1)}}$$

Since $\sqrt{\frac{d}{d-1}} \leq 2$ for $d \geq 2$

$$\leq \frac{2\sqrt{2}}{\pi} \alpha\sqrt{d}$$

Given that all of this only holds asymptotically; we might need another multiplicative constant to make it hold in general. Hence the constant C in the theorem.

$$p \leq C\alpha\sqrt{d}$$

This bounds the probability $p = \Pr(\alpha \leq |y_1|) \leq C\alpha\sqrt{d}$. Considering the probability of the complement event the bounds $1 - \Pr(\alpha > |y_1|) \leq C\alpha\sqrt{d}$ can be stated as

$$1 - C\alpha\sqrt{d} \leq \Pr(\alpha > |y_1|).$$

Remark 2.1 Notice that in the previous result essentially shows also that, independently of the dimension d , the $x_1 = \langle x, u_1 \rangle$ component of a random unit vector x with respect to any orthonormal basis $\{u_1, \dots, u_d\}^1$ is bounded away from zero with overwhelming probability.

Remark 2.2 Consider the isometric mapping $(a, b) \rightarrow a + bi$ from \mathbb{R}^2 to \mathbb{C} . The previous result extends to random unit vectors in \mathbb{C}^d simplify by modifying the statement as follows: The probability that, for a randomly chosen unit vector $z \in \mathbb{C}^d$, $|z_1| \geq \alpha > 0$ holds is at least $1 - C\alpha\sqrt{2d} = 1 - C'\alpha\sqrt{d}$.

It is important to note that remark 2.1 and remark 2.2 holds with any orthonormal basis by rotating it to coincide with the canonical basis.

Theorem 2.2 Let $A \in \mathbb{K}^{I \times J}$ and $x \in \mathbb{K}^I$. Let V be the space spanned by the left singular vectors of A corresponding to singular values greater than $(1 - \epsilon)\sigma_1$. Let $m \in \Omega\left(\frac{\ln(d/\epsilon)}{\epsilon}\right)$. Let w^* be the unit vector after m iterations of the power method, namely,

$$w^* = \frac{(AA^H)^m x}{\|(AA^H)^m x\|_2} \quad (1)$$

The probability that w^* has a component of at most l , where $l \in O\left(\frac{\epsilon}{\alpha d}\right)$, orthogonal to V is at least $1 - C\alpha\sqrt{d}$ i.e. $1 - C\alpha\sqrt{d} < \Pr(\|Proj_{V^\perp}(w^*)\|_2 < l)$.

Proof

Let the SVD of A be given by

$$A = \sum_{k=1}^r \sigma_k u_k v_k^H$$

If the rank of A is less than $n = |I|$ we complete the orthonormal set of vectors $\{u_1, \dots, u_r\}$ into a full orthogonal basis $\{u_1, \dots, u_n\}$ of the n -dimensional space. We can expand x in the terms of this basis as

$$x = \sum_{k=1}^n \langle x, u_k \rangle u_k$$

We set $\sigma_k = 0$ for $k > r$ so that we can write A as

$$A = \sum_{k=1}^n \sigma_k u_k v_k^H$$

It follows that

$$(AA^H)^m x = \sum_{k=1}^n \sigma_k^{2m} u_k u_k^H x = \sum_{k=1}^n \sigma_k^{2m} u_k \langle x, u_k \rangle$$

By lemma 2.1, remark 2.1 and remark 2.2 one has $|\langle x_1, u_1 \rangle| \geq \alpha > 0$ with probability at least $1 - C\alpha\sqrt{d}$. We choose r_ϵ such that $\sigma_1, \dots, \sigma_{r_\epsilon}$ are the singular values of A that are greater or equal to $(1 - \epsilon)\sigma_1$ and $\sigma_{r_\epsilon+1}, \dots, \sigma_n$ are those that are less than $(1 - \epsilon)\sigma_1$. Notice that $V = \text{span}\{\sigma_1, \dots, \sigma_{r_\epsilon}\}$ and $V^\perp = \text{span}\{\sigma_{r_\epsilon+1}, \dots, \sigma_n\}$. The component of w^* orthogonal to V^\perp is $\text{Proj}_{V^\perp}(w^*)$ which can be written as

$$\text{Proj}_{V^\perp}(w^*) = \frac{\text{Proj}_{V^\perp}((AA^H)^m x)}{\|(AA^H)^m x\|_2} \quad (2)$$

We find denominator of equation 2 by Pythagoras-Fourier theorem

$$\|(AA^H)^m x\|_2^2 = \sum_{k=1}^n \sigma_k^{4m} |\langle x, u_k \rangle|^2 \quad (3)$$

$$\sum_{k=1}^n \sigma_k^{4m} |\langle x, u_k \rangle|^2 \geq \sigma_1^{4m} |\langle x, u_1 \rangle|^2 \geq \sigma_1^{4m} \alpha^2 \quad (4)$$

with probability at least $1 - C\alpha\sqrt{d}$. To find the nominator of equation 2, we check component of $(AA^H)^m x$ that is orthogonal to $V = \text{span}\{u_1, \dots, u_{r_\epsilon}\}$, namely,

$$\text{Proj}_{V^\perp}((AA^H)^m x) = \sum_{k=1}^n \sigma_k^{2m} |\langle x, u_k \rangle|_2 = \sum_{k=r_\epsilon+1}^n \sigma_k^{2m} |\langle x, u_k \rangle|_2 \quad (5)$$

$$\text{Proj}_{V^\perp}((AA^H)^m x) \leq (1 - \epsilon)^{2m} \sigma_1^{2m} \sum_{k=r_\epsilon+1}^n |\langle x, u_k \rangle|_2 \leq (1 - \epsilon)^{2m} \sigma_1^{2m} \quad (6)$$

since $\sum_{k=r_\epsilon+1}^n \|\langle x, u_k \rangle\|_2^2 = 1$ and $(1 - \epsilon)\sigma_1 > \sigma_k$ for $r_\epsilon < k$.

By using 3 and 5 we find squared norm of the component of w^* orthogonal to V , that is $\|\text{Proj}_{V^\perp}(w^*)\|_2^2$, as

$$\|\text{Proj}_{V^\perp}(w^*)\|_2^2 = \frac{\sum_{k=r_\epsilon+1}^n \sigma_k^{4m} |\langle x, u_k \rangle|^2}{\sum_{k=1}^n \sigma_k^{4m} |\langle x, u_k \rangle|^2}$$

We bound this term by using the relations 4 and 6

$$\|\text{Proj}_{V^\perp}(w^*)\|_2^2 = \frac{\sum_{k=r_\epsilon+1}^n \sigma_k^{4m} |\langle x, u_k \rangle|^2}{\sum_{k=1}^n \sigma_k^{4m} |\langle x, u_k \rangle|^2} \leq \frac{(1 - \epsilon)^{4m} \sigma_1^{4m}}{\alpha^2 \sigma_1^{4m}} = \frac{(1 - \epsilon)^{4m}}{\alpha^2}$$

Thus, by taking the square root we have

$$\|\text{Proj}_{V^\perp}(w^*)\|_2 \leq \frac{(1 - \epsilon)^{2m}}{\alpha}$$

In terms of *Big O* notation we have

$$\|\text{Proj}_{V^\perp}(w^*)\|_2 \in \mathcal{O}\left(\frac{(1-\epsilon)^{2m}}{\alpha}\right)$$

Notice that $1 - \epsilon$ is a linear approximation of $e^{-\epsilon}$. Similarly, $(1 - \epsilon)^{2m}$ approximates $e^{-2m\epsilon}$ for small ϵ . Using this approximation,

$$\|\text{Proj}_{V^\perp}(w^*)\|_2 \in \mathcal{O}\left(\frac{e^{-2m\epsilon}}{\alpha}\right)$$

Recall that $m \in \Omega\left(\frac{\ln(d/\epsilon)}{\epsilon}\right)$. This is another way of saying there exists m_0 and some constant $c > 0$ such that $m \geq c \frac{\ln(d/\epsilon)}{\epsilon}$ for all $m > m_0$. Similarly, this also means there exists m_0 and some constant $c > 0$ such that $-\frac{m}{c} \leq -\frac{\ln(d/\epsilon)}{\epsilon}$ for all $m > m_0$. Since exponentiation is a non-decreasing function $e^{-\frac{m}{c}} \leq e^{-\frac{\ln(d/\epsilon)}{\epsilon}} = e^{\frac{\ln(\epsilon/d)}{\epsilon}} = (\epsilon/d)^{1/\epsilon}$. We have

$$e^{-\frac{m}{c}} \leq (\epsilon/d)^{1/\epsilon}$$

for some constant $c > 0$ and $m > m_0$. We take the power of ϵ of both sides

$$e^{-\frac{m\epsilon}{c}} \leq \frac{\epsilon}{d}$$

Let $c_1 = c/2$

$$e^{-\frac{2m\epsilon}{c_1}} \leq \frac{\epsilon}{d}$$

For some constant $e^{-1/c_1} > 0$ and all $m > m_0$. We divide both sides with α

$$\frac{e^{-\frac{2m\epsilon}{c_1}}}{\alpha} \leq \frac{\epsilon}{\alpha d}$$

$$\alpha^{-1} e^{-\frac{2m\epsilon}{c_1}} \leq \frac{\epsilon}{\alpha d}$$

Which means

$$\frac{e^{-2m\epsilon}}{\alpha} \in \mathcal{O}\left(\frac{\epsilon}{\alpha d}\right)$$

Consequently

$$\|\text{Proj}_{V^\perp}(w^*)\|_2 \in \mathcal{O}\left(\frac{\epsilon}{\alpha d}\right)$$