# Johnson-Lindenstrauss Lemma and Restricted Isometry Property

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#### **Outline**

Introduction

Johnson-Lindenstrauss (JL) Lemma

**Concentration Inequality** 

Restricted Isometry Property (RIP)

Connection between JL Lemma and RIP

#### Johnson-Lindenstrauss Lemma 1

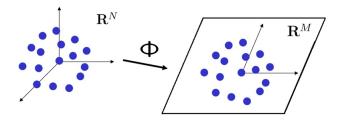


Figure: Mapping from high dimensional space on lower dimensional space

#### Johnson-Lindenstrauss Lemma 2

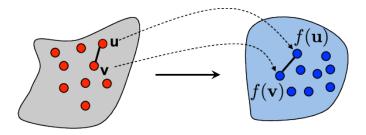


Figure: Mapping from high dimensional space on lower dimension by keeping relative distances

#### Johnson-Lindenstrauss Lemma 3

**Lemma 4** (Johnson-Lindenstrauss) (Lemma 4.1 in (Baraniuk et al., 2008))

Let  $\epsilon \in (0,1)$  is given and  $N \in \mathbb{N}$ . For every set  $Q \in \mathbb{R}^N$ , if n is a positive integer such that  $n \geq n_0 := O(\ln(\#Q)/\epsilon^2)$ , there exists a Lipschtz mapping  $f : \mathbb{R}^N \to \mathbb{R}^n$  such that

$$(1 - \epsilon) \|u - v\|_2^2 \le \|f(u) - f(v)\|_2^2 \le (1 + \epsilon) \|u - v\|_2^2,$$
 (1)

for all  $u, v \in Q$ .

# **Concentration Inequality**

#### **Definition 2**

Let  $\Phi \in \mathbb{R}^{n \times N}$  be a random matrix and for any  $x \in \mathbb{R}^N$  expectation of random variable  $\|\Phi(x)\|_2^2$  is  $\|x\|_2^2$ , that is

$$\mathbb{E}[\|\Phi(x)\|_2^2] = \|x\|_2^2,$$

where  $n, N \in \mathbb{N}$ . We say that random matrix  $\Phi$  satisfies concentration inequality if the random variable  $\|\Phi(x)\|_2^2$  is strongly concentrated about its expected value, that is

$$\mathbb{P}(\|\Phi(x)\|_{2}^{2} - \|x\|_{2}^{2}) \ge \epsilon \|x\|_{2}^{2} \le 2e^{-nc(\epsilon)}, \quad 0 \le \epsilon \le 1,$$

where  $\epsilon$  is a constant and  $c(\epsilon)$  is also a constant depends only on  $\epsilon$  such that for all  $\epsilon \in (0,1)$ ,  $c(\epsilon) \geq 0$ .

# **Concentration Inequality: Examples 1**

Entries  $\Phi_{i,j}$  of  $\Phi$  are independent samples from Normal distribution

$$\{\Phi_{i,j}\}_{i\in[n],j\in[N]} \stackrel{i.i.d}{\sim} \mathcal{N}(0,\frac{1}{n}).$$

# **Concentration Inequality: Examples 2**

Entries  $\Phi_{i,j}$  of  $\Phi$  are independent samples from following distribution

$$X := \begin{cases} +1/\sqrt{n} & \text{with probability } 1/2, \\ -1/\sqrt{n} & \text{with probability } 1/2. \end{cases}$$

# **Concentration Inequality: Examples 3**

Entries  $\Phi_{i,j}$  of  $\Phi$  are independent samples from following distribution

$$X := \begin{cases} +\sqrt{3}/\sqrt{n} & \text{with probability } 1/6, \\ 0 & \text{with probability } 2/3, \\ -\sqrt{3}/\sqrt{n} & \text{with probability } 1/6. \end{cases}$$

In all the above examples the constant  $c(\epsilon)$  can be chosen as  $\epsilon^2/4 - \epsilon^3/6$ .

# Restricted Isometry Property 1

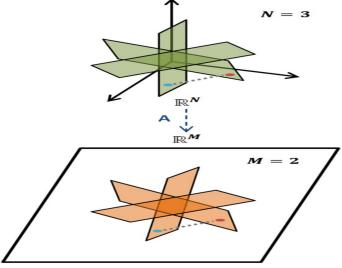


Figure: RIP mapping

# **Restricted Isometry Property 2**

#### **Definition 3**

We say that a matrix  $\Phi$  satisfies Restricted Isometry Property (RIP) of order k and level  $\delta \in (0,1)$  (concisely,  $(k,\delta)-R$ IP) if

$$(1-\delta) \|x\|_2^2 \le \|\Phi(x)\|_2^2 \le (1+\delta) \|x\|_2^2 \quad \forall k-sparse \ x \in \mathbb{R}^N.$$
 (2)

The restricted isometry constant  $\delta_k$  is defined as the smallest value of  $\delta$  for which (2) holds.

# JL Lemma implies RIP

**Theorem 4** (Theorem 5.2 in (Baraniuk et al., 2008)) Let  $n, N \in \mathbb{N}$ ,  $\delta \in (0, 1)$  and  $\Phi \in \mathbb{R}^{n \times N}$  be a random matrix which satisfies concentration inequality (see Definition 2), then  $\exists c_1, c_2 > 0$  depending only on  $\delta$  such that  $\Phi$  satisfies Restricted Isometry Property of order  $k \leq c_1 n/\log(N/k)$  and level  $\delta$  with probability  $\geq 1 - 2e^{-c_2 n}$ .

#### Proof idea of Theorem 4

1. For each index set T with #T = k there is finite set of points  $\mathcal{Q}_T \subseteq X_T$  such that  $\|q\|_2 = 1$ ,  $\forall q \in \mathcal{Q}_T$  and  $\forall x \in X_T$  with  $\|x\|_2 = 1$  we have

$$\min_{q \in \mathcal{Q}_T} \|x - q\|_2 \le \delta/4. \tag{3}$$

such that  $\#Q_T \leq (12/\delta)^k$ .

2. Using 1. prove

$$(1 - \delta) \|x\|_{2} \le \|\Phi(x)\|_{2} \le (1 + \delta) \|x\|_{2}, \quad \forall x \in X_{T},$$
 (4)

with high probability

3. Using union bound with 2. prove RIP

$$(1 - \delta) \|x\|_2 \le \|\Phi(x)\|_2 \le (1 + \delta) \|x\|_2 \quad \forall k - \text{sparse } x \in \mathbb{R}^N,$$
 (5)

with high probability.

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# RIP implies JL Lemma (1)

- 1. Matrices that satisfy RIP are not random!
- 2. How can we randomize a matrix?

# RIP implies JL Lemma (2)

#### **Definition 5**

We call  $\xi \in \mathbb{R}^N$  as a Rademacher sequence if it is uniformly distributed on  $\{-1, +1\}^N$ , where  $N \in \mathbb{N}$ .

For a matrix  $\Phi$  we might consider random matrix  $\Phi D_{\xi}$ , where  $\xi$  is a Rademacher sequence and  $D_{\xi}$  is a diagonal matrix with entries from  $\xi$ 

# RIP implies JL Lemma (2)

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# RIP implies JL Lemma (3)

**Theorem 6** (Theorem 3.1 in (Krahmer and Ward, 2010)) Let  $\eta > 0, \epsilon \in (0,1), n, N \in \mathbb{N}$  be given and the set  $E \subset \mathbb{R}^N$  has cardinality p := #E. Suppose the matrix  $\Phi \in \mathbb{R}^{n \times N}$  satisfies Restricted Isometry Property of order  $k \geq 40\log\frac{4p}{\eta}$  and level  $\delta \leq \frac{\epsilon}{4}$ . Taking  $\xi \in \mathbb{R}^N$  as a Rademacher sequence we have

$$(1 - \epsilon) \|x\|_2^2 \le \|\Phi D_{\xi} x\|_2^2 \le (1 + \epsilon) \|x\|_2^2, \tag{6}$$

for all  $x \in E$ , with probability

$$\geq 1 - \eta$$
.

1 Split matrices by column wise blocks

$$\|\Phi D_{\xi} x\|_{2}^{2} = \|\Phi D_{x} \xi\|_{2}^{2}$$

$$= \sum_{J=1}^{R} \|\Phi_{(J)} D_{(x_{(J)})} \xi_{(J)}\|_{2}^{2}$$

$$+ 2 \xi_{(1)}^{*} D_{x_{(1)}} \Phi_{(1)}^{*} \Phi_{(-1)} D_{x_{(-1)}} \xi_{(-1)}$$

$$+ \sum_{J,L=2,J\neq L}^{R} \langle \Phi_{(J)} D_{x_{(J)}} \xi_{(J)}, \Phi_{(L)} D_{x_{(L)}} \xi_{(L)} \rangle.$$
(7)

- 2 Get bounds for every item in the right hand side of equation (7)
  - 2.1 Deterministic bound on the first item

$$(1 - \epsilon/4) \|x\|_{2}^{2} \le \sum_{J=1}^{R} \|\Phi_{(J)} D_{x_{(J)}} \xi_{(J)}\|_{2}^{2} \le (1 + \epsilon/4) \|x\|_{2}^{2}.$$
 (8)

2.2 Probabilistic bound on the second item

$$\mathbb{P}(\exists x \in E : |2\xi_{(1)}^* D_{x_{(1)}} \Phi_{(1)}^* \Phi_{(-1)} D_{x_{(-1)}} \xi_{(-1)}| \ge 0.2\epsilon) \le \frac{\eta}{2}$$
 (9)

$$\mathbb{P}(\exists x \in E : |\sum_{J,L=2,J\neq L}^{R} \langle \Phi_{(J)} D_{x_{(J)}} \xi_{(J)}, \Phi_{(L)} D_{x_{(L)}} \xi_{(L)} \rangle| \ge 0.55\epsilon) \le \eta/2$$
(10)

- 2 Get bounds for every item in the right hand side of equation (7)
  - 2.1 Deterministic bound on the first item

$$(1 - \epsilon/4) \|x\|_2^2 \le \sum_{I=1}^R \|\Phi_{(I)} D_{x_{(J)}} \xi_{(J)}\|_2^2 \le (1 + \epsilon/4) \|x\|_2^2.$$
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$$\mathbb{P}(\exists x \in E : |\sum_{J,L=2,J\neq L}^{R} \langle \Phi_{(J)} D_{X_{(J)}} \xi_{(J)}, \Phi_{(L)} D_{X_{(L)}} \xi_{(L)} \rangle| \ge 0.55\epsilon) \le \eta/2$$
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$$\mathbb{P}(\exists x \in E : |\sum_{J,L=2,J\neq L}^{K} \langle \Phi_{(J)} D_{x_{(J)}} \xi_{(J)}, \Phi_{(L)} D_{x_{(L)}} \xi_{(L)} \rangle| \ge 0.55\epsilon) \le \eta/2$$
(10)

3 Putting everything together

$$\mathbb{P}(\exists x \in E : |\|\Phi D_{\xi} x\|_{2}^{2} / \|x\|_{2}^{2} - 1| \ge \epsilon) \le \eta. \tag{11}$$

Thank you!

#### references

Baraniuk, R., Davenport, M., DeVore, R., and Wakin, M. (2008). A simple proof of the restricted isometry property for random matrices. *Constructive Approximation*, 28(3):253–263.

Krahmer, F. and Ward, R. (2010). New and improved Johnson-Lindenstrauss embeddings via the Restricted Isometry Property. *arXiv e-prints*, page arXiv:1009.0744.