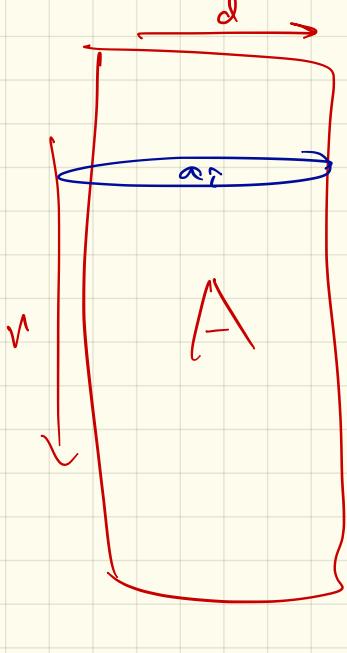


# Random Projections

Input  $\xrightarrow{d}$   $n$  data points  $\{q_1, q_2, \dots, q_n\}$   
 $q_i \in \mathbb{R}^d$



SVD / PCA  $\rightarrow A \in \mathbb{R}^{n \times d}$   
 ↳ Best rank-k subspace

$$[A] = [U] \begin{bmatrix} S \\ 0 \end{bmatrix} [V]^T$$

$U_{12} \leftarrow \underset{\text{rank}(U) \leq k}{\arg \min} \sum_{a_i \in A} \|q_i - T_{U_{12}}(q_i)\|^2$

$$|\Lambda| = n$$

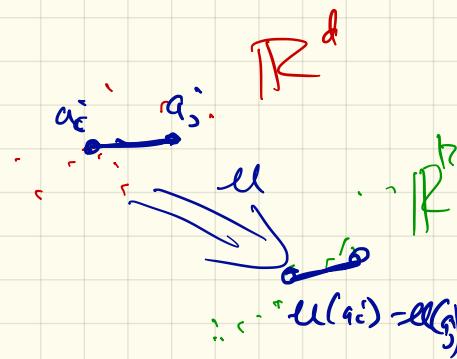
Define some map  $u: \mathbb{R}^d \rightarrow \mathbb{R}$

$\forall a_i, a_j \in \Lambda$  error parameter  $\varepsilon > 0$   
 $\approx 0.10$

$$(1-\varepsilon) \|a_i - a_j\| \leq \|u(a_i) - u(a_j)\| \leq (1+\varepsilon) \|a_i - a_j\|$$

with prob.  $\geq 1 - \delta$

$$t \approx \left(\frac{1}{\varepsilon^2}\right) \log \frac{n}{\delta}$$



How to construct  $\mu: \mathbb{R}^d \rightarrow \mathbb{R}^k$

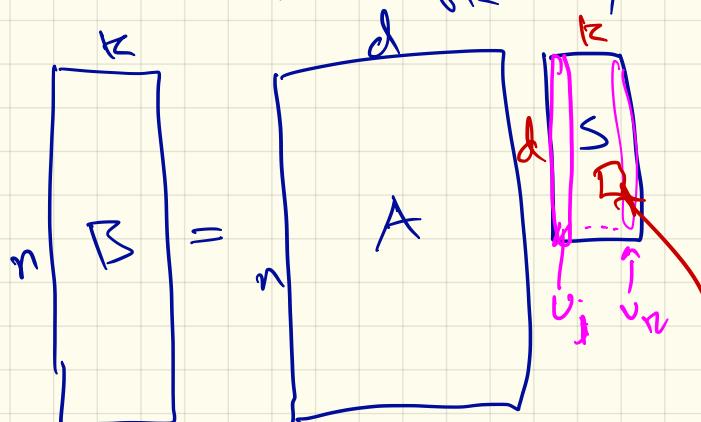
$$\mu(a) = (\mu_1(a), \mu_2(a), \dots, \mu_k(a))$$

Focus on  $\mu_j(a)$

$$\mu_j(a) \xrightarrow{\frac{\sqrt{d}}{\sqrt{k}}} \langle a, u_j \rangle$$

$$\text{so } u_j \in \text{Unif}(\mathbb{S}^{d-1})$$

$$\|u_j\| = 1$$



$$g_j \sim N_d(0, I)$$

$$u_j = \frac{g_j}{\|g_j\|}$$

$$s_{ij} \sim \frac{\sqrt{d}}{\sqrt{k}} N_1(0, 1)$$

Box-Multivar

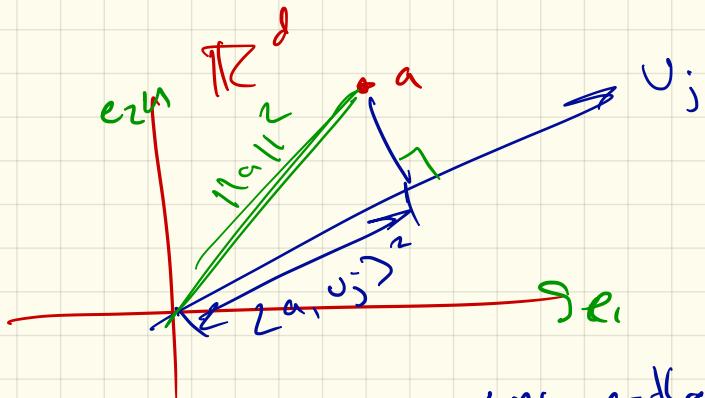
$$E\left[\|u(a)\|^2\right] = \|a\|^2$$

$$E\left[\frac{\partial}{\partial x_j} \langle u_j, a \rangle^2\right]$$

$$E\left[d \langle u_j, a \rangle^2\right] = \|a\|^2$$

*jth coord*  
↓

$$l_j = (v, \sigma_+, t, \sigma_-)$$



$$\|a\|^2 = \sum_{j=1}^d \langle a, e_j \rangle^2$$

Pythagorean

any orthonormal basis

$$U = [u_1, u_2, \dots, u_j, \dots, u_d]$$

# Oblivious Supspace Embedding

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$$\boxed{B} = \boxed{A} \boxed{S}$$

$$\forall x \in \mathbb{R}^d$$

$$t = O\left(\frac{d}{\epsilon^2} \log \frac{1}{\delta}\right)$$

$$(1-\epsilon) \|Ax\| \leq \|Bx\| \leq (1+\epsilon) \|Ax\|$$


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Count-Sample

