

Recap:

ℓ_2 regression:

$$\underset{x}{\operatorname{argmin}} \quad \|Ax - b\|_2^2$$

$A: n \times d$ matrix
 $b \in \mathbb{R}^n, n \gg d$

$O(nd^2)$ via normal eqn

relaxed problem

$$\|Ax - b\|_2 \leq (1+\varepsilon) \|Ax^* - b\|_2$$

failure prob. δ

① Sample $r \times n$ matrix S rccn

② Compute SA, Sb

③ Solve $\underset{x}{\operatorname{argmin}} \underbrace{\|(SA)x - (Sb)\|_2}_\text{smaller regression}$

ℓ_2 -subspace embedding:

$(1 \pm \varepsilon)$ ℓ_2 subspace embedding for col-space of $n \times d$ matrix A

is a matrix S s.t. $\forall x \in \mathbb{R}^d \quad \|Sx\|_2^2 = (1 \pm \varepsilon) \|Ax\|_2^2$

U : orthonormal basis for cols of A

$$\|S^T y\|_2^2 = (1 \pm \varepsilon) \|Uy\|_2^2 = (1 \pm \varepsilon) \|y\|_2^2$$

operator norm $\|I_d - U^T S^T S U\|_2 \leq \varepsilon$

Oblivious ℓ_2 -subspace embedding

distrb^n on $r \times n$ matrices S

w. prob $1-\delta$ for any fixed $n \times d$ matrix A

$S \sim \Pi$ is a $(1 \pm \varepsilon)$ ℓ_2 subspace embedding for A

Def: Random matrix $S \in \mathbb{R}^{k \times n}$ is a JL-transform $JLT(\varepsilon, \delta, f)$

if w. prob $1-\delta$ for any f element subset $V \subset \mathbb{R}^n$,

$$\forall v, v' \in V, \quad |\langle Sv, Sv' \rangle - \langle v, v' \rangle| \leq \varepsilon \|v\|_2 \|v'\|_2$$

Thm: $S = \frac{1}{\sqrt{k}} R \in \mathbb{R}^{k \times n}$ R_{ij} indept normal, $k = \lceil 2 \left(\frac{\log(f/\delta)}{\varepsilon^2} \right) \rceil$
then S is a $JLT(\varepsilon, \delta, f)$

Subspace Embedding Property

$$T = \{y \in \mathbb{R}^n \mid y = Ax \text{ for some } x \in \mathbb{R}^d, \|y\|_2 = 1\}$$

Need: $\|\text{Syl}\|_2^2 = (1 \pm \varepsilon) \|y\|_2^2 = 1 \pm \varepsilon \quad \forall y \in T$

ε -net: finite subset $N \subset T$ so that $\forall y \in T$

$$\exists w \in N, \|y - w\|_2 \leq \frac{1}{2}$$

If $\forall w, w' \in N, \langle sw, sw' \rangle = \langle w, w' \rangle \pm \varepsilon$

then get $\|\text{Syl}\|_2^2 = 1 \pm o(\varepsilon)$

Lemma: $\exists r$ -net N of T , $|N| \leq (1 + \frac{2}{r})^d$

Pf: $t = \text{rank}(A)$

$$T = \{y \in \mathbb{R}^n \mid y = Ux, x \in \mathbb{R}^t, \|y\|_2 = 1\}$$

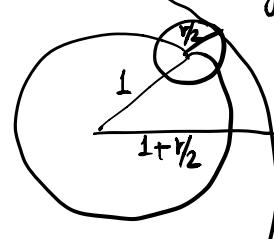
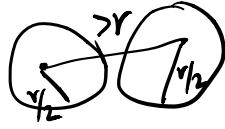
U : orthonormal basis for A

(Intuition) Net for S^{t-1} will give net for T
image of S^{t-1} under U

Choose maximal set N' of points in S^{t-1}

so that no 2 points within dist r of each other.

Balls of radius $r/2$ centered at points in N' are disjoint



$$|N'| \leq \frac{\left(1 + \frac{r}{2}\right)^t}{\left(\frac{r}{2}\right)^t} = \left(1 + \frac{2}{r}\right)^t$$

$$N = \{y \in \mathbb{R}^n \mid y = Ux \text{ for some } x \in N'\}$$

If $\exists Ux \in T$ s.t. $\forall y = Uz \in N, z \in N'$

$$\|y - Ux\|_2 = \|Uz - Ux\|_2 > r$$

$\Rightarrow x \in S^{t-1}$, s.t. $\forall z \in N', \|x - z\|_2 > r$ contradiction

Set $V = N$, $f = 5^d$ in JLT theorem

$$S = \frac{1}{\sqrt{K}} R \quad R: K \times n \quad R_{ij} : \text{indep std. normal}$$

$$K = \Theta\left(\left(d + \log(1/\delta)\right)/\epsilon^2\right)$$

Oblivious ℓ_2 -subspace embedding

Optimal # rows : d/ϵ^2

bottleneck : time to compute $S A$

$$\downarrow$$

$$nnz(A) \quad O\left(\frac{d}{\epsilon^2} \cdot nnz(A)\right)$$

[Dasgupta, Kumar, Sarlos] [Kane, Nelson]

$$\frac{\log(f/\delta)}{\epsilon} \text{ non-zero entries per col}$$

$$O\left(nnz(A) \cdot \frac{d}{\epsilon}\right)$$

Near-optimal.

[Clarkson-Woodruff] $S.A$ in $O(nnz(A))$

Sparse Embedding Matrix

S : Count Sketch matrix $r \times n$

For each of n cols S_{x_i}

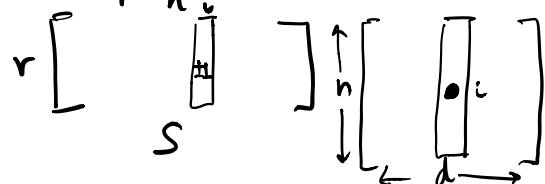
choose uniform random row $h(i) \in \{1, 2, \dots, r\}$

choose random ele of $\{\pm 1\}$: $\sigma(i)$

$$S_{h(i), i} = \sigma(i)$$

$$S_{j,i} = 0 \quad \forall j \neq h(i)$$

Can compute $S.A$ in $O(nnz(A))$



Thm

S : Count Sketch matrix
 $r = O\left(\frac{d^2}{\varepsilon^2 \delta}\right)$ rows
 For any fixed A , S is $(\pm \varepsilon)$ - ℓ_2 subspace embedding
 for A w. prob $1-\delta$

h : pairwise indept
 σ : 4-wise indept.

Proof Sketch:

$$\Pr\left[\|I_d - u^T S^T S u\|_2 \geq \varepsilon\right] = \Pr\left[\|I_d - u^T S^T S u\|_2^\ell \geq \varepsilon^\ell\right] \text{ (even)} \\ \leq \frac{1}{\varepsilon^\ell} E\left[\|I_d - u^T S^T S u\|_2^\ell\right] \\ \leq \frac{1}{\varepsilon^\ell} E\left[\text{tr}((I_d - u^T S^T S u)^\ell)\right]$$

Simpler Proof [Nguyen]

Approx. Matrix Multiplication

Defn C is ε -approx matrix product of A, B if
 $\|A^T B - C\|_F \leq \varepsilon \|A\|_F \|B\|_F$

Maintain sketch SA, SB, S : $r \times n$ matrix

Want $E[A^T S^T S B] = A^T B$

Defn: Distrib "D" on $S \in \mathbb{R}^{k \times d}$ satisfies
 (ε, δ, k) -JL moment property if
 $\forall x \in \mathbb{R}^d, \|x\|_2 = 1 \quad E\left[\left(\|Sx\|_2^2 - 1\right)^\ell\right] \leq \varepsilon^\ell \delta$

Defn: For scalar random var X
 $\|X\|_p = E[|X|^p]^{1/p}$
 $\|\cdot\|_p$ is a metric: $\|(X+Y)\|_p \leq \|(X)\|_p + \|Y\|_p$

Lemma: Let $\ell \geq 2$, $\varepsilon, s \in (0, 1/2)$
D: distribution that satisfies (ε, s, ℓ) -JL moment property
For A, B with d rows

$$P_{S \sim D} \left[\|A^T S^T S B - A^T B\|_F > 3\varepsilon \|A\|_F \|B\|_F \right] \leq s$$

Pf: For unit vectors $x, y \in \mathbb{R}^d$

$$\begin{aligned} \langle Sx, Sy \rangle &= \frac{1}{2} (\|Sx\|_2^2 + \|Sy\|_2^2 - \|S(x-y)\|_2^2) \\ \|\langle Sx, Sy \rangle - \langle x, y \rangle\|_2 &= \frac{1}{2} \left\| \left(\|Sx\|_2^2 - 1 \right) + \left(\|Sy\|_2^2 - 1 \right) - \left(\|S(x-y)\|_2^2 - \|x-y\|_2^2 \right) \right\|_2 \\ &\leq \frac{1}{2} \left(\|\|Sx\|_2^2 - 1\|_2 + \|\|Sy\|_2^2 - 1\|_2 + \|\|S(x-y)\|_2^2 - \|x-y\|_2^2\|_2 \right) \\ &\leq \frac{1}{2} (\varepsilon s^{\ell/2} + \varepsilon s^{\ell/2} + \|x-y\|_2 \leq s^{\ell/2}) \\ &\leq 3\varepsilon s^{\ell/2} \end{aligned}$$

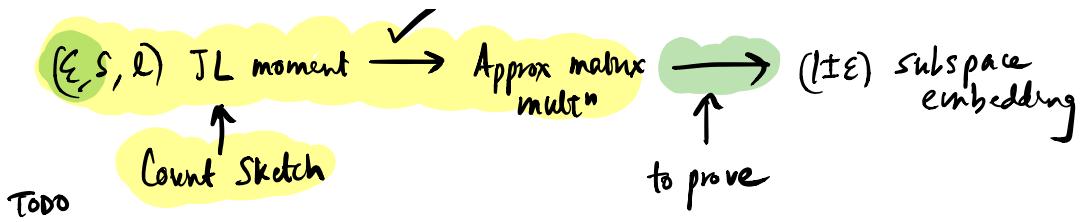
For arbitrary x, y

$$\|\langle Sx, Sy \rangle - \langle x, y \rangle\|_2 \leq 3\varepsilon s^{\ell/2} \|x\|_2 \|y\|_2$$

$(i, j)^{\text{th}}$ entry of $A^T B$ is (A^i, B^j)

$$\begin{aligned} \|\|A^T S^T S B - A^T B\|_F^2\|_{\ell/2} &\leq \sum_{i,j} \|(\langle S A^i, S B^j \rangle - \langle A^i, B^j \rangle)^2\|_{\ell/2} \\ &\leq (3\varepsilon s^{\ell/2})^2 \sum_{i,j} \|A^i\|_2^2 \|B^j\|_2^2 \\ &= (3\varepsilon s^{\ell/2})^2 \|A\|_F^2 \|B\|_F^2 \\ \mathbb{E}[\|A^T S^T S B - A^T B\|_F^\ell] &\leq (3\varepsilon s^{\ell/2})^\ell \|A\|_F^\ell \|B\|_F^\ell \end{aligned}$$

$$\begin{aligned} P[\|A^T S^T S B - A^T B\|_F^\ell > (3\varepsilon)^\ell \|A\|_F^\ell \|B\|_F^\ell] \\ &\leq \frac{1}{(3\varepsilon \|A\|_F \|B\|_F)^\ell} \mathbb{E}[\|A^T S^T S B - A^T B\|_F^\ell] \\ &\leq s \end{aligned}$$



S : Count Sketch with $\frac{2}{\epsilon^2 \delta}$ rows.

Claim: S satisfies $(\epsilon, \delta, 2)$ JL moment property

$$h: [d] \rightarrow [r] \quad \text{2-wise indept}$$

$$\sigma: [d] \rightarrow \{\pm 1\} \quad \text{4-wise indept.}$$

Pf: For any unit vector $x \in \mathbb{R}^d$

$$\mathbb{E}_S[(\|Sx\|_2^2 - 1)^2] = \mathbb{E}_S[\|Sx\|_2^4 - 2\mathbb{E}_S[\|Sx\|_2^2] + 1] \stackrel{\text{to prove}}{\leq} \epsilon^2 \delta$$

$$\begin{aligned} \mathbb{E}[\|Sx\|_2^2] &= \sum_{i \in [r]} \mathbb{E}\left[\left(\sum_{j \in [d]} I_{h(j)=i} x_j \sigma(j)\right)^2\right] \\ &= \sum_{i \in [r]} \sum_{j, j' \in [d]} x_j x_{j'} \mathbb{E}[I_{h(j)=i} I_{h(j')=i}] \mathbb{E}[\sigma(j) \sigma(j')] \\ &= \sum_{i \in [r]} \sum_{j \in [d]} \frac{x_j^2}{r} \\ &= \|x\|_2^2 = 1 \end{aligned}$$

$$\mathbb{E}[\|Sx\|_2^4] \leq 1 + \frac{2}{r}$$

$$\text{Need } r \geq \frac{2}{\epsilon^2 \delta}$$

Proof: (Count Sketch gives ℓ_2 -subspace embedding)

S satisfies $(\epsilon, \delta, 2)$ JL moment property

U orthonormal basis for cols of A

$$P\left[\|U^T S^T S U - \underbrace{\|U^T U\|_F}_{I_d} \geq 3\epsilon \underbrace{\|U\|_F \|U\|_F}_d\right] \leq \delta$$

$$\text{Apply with } \varepsilon' = \varepsilon/d \quad r = \frac{2}{(\varepsilon')^2 s} = \frac{2d^2}{\varepsilon^2 s}$$

$$P[\|u^T S^T S u - I_d\|_F > 3\varepsilon] \leq \delta$$

Woodruff: Sketching As a Tool for Numerical Linear Algebra