

Recap:

l_2 regression:

$$\operatorname{argmin}_x \|Ax - b\|_2^2$$

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

$O(nd^2)$ via normal eqⁿ

relaxed
problem

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

failure prob. δ

- ① Sample $r \times n$ matrix S $r < n$
- ② Compute SA, Sb
- ③ Solve $\operatorname{argmin}_x \|(SA)x - (Sb)\|_2$
smaller regression

l_2 -subspace embedding:

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

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

U : orthonormal basis for cols of A

$$\|SUy\|_2^2 = (1 \pm \epsilon) \|Uy\|_2^2 = (1 \pm \epsilon) \|y\|_2^2$$

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

Oblivious l_2 -subspace embedding

distribⁿ on $r \times n$ matrices S

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

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

Defⁿ Random matrix $S \in \mathbb{R}^{r \times n}$ is a JL-transform $JLT(\epsilon, \delta, f)$

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

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

Thm: $S = \frac{1}{\sqrt{K}} R \in \mathbb{R}^{r \times n}$ Rij indpt normal, $K = \Omega\left(\frac{\log(f/\delta)}{\epsilon^2}\right)$
then S is a $JLT(\epsilon, \delta, f)$

Subspace Embedding Property

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

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

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

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

$$\text{If } \forall w, w' \in N, \langle Sw, Sw' \rangle = \langle w, w' \rangle \pm \epsilon$$

$$\text{then } \forall y \in T \quad \|Sy\|_2^2 = 1 \pm O(\epsilon)$$

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

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

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

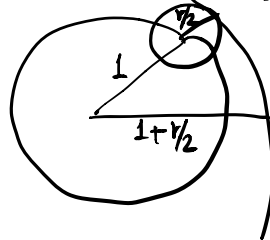
U : orthonormal basis for A

(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 γ of each other.

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



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

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

If $\exists uz \in T$ st. $\forall y = Ax \in N, z \in N'$

$$\|y - uz\|_2 = \|Ax - uz\|_2 > \gamma$$

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

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

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

$$k = \Theta \left(\frac{d + \log(1/\delta)}{\epsilon^2} \right)$$

Oblivious l_2 -subspace embedding

Optimal # rows: d/ϵ^2

bottleneck: time to compute SA $O(kdn) = O(nd^2/\epsilon^2)$
 \downarrow
 $\text{nnz}(A)$ $O\left(\frac{d}{\epsilon^2} \cdot \text{nnz}(A)\right)$

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

$$\frac{\log(f/\delta)}{\epsilon} \text{ non-zero entries per col} \quad O(\text{nnz}(A) \cdot \frac{d}{\epsilon})$$

near-optimal.

[Clarkson-Woodruff] S.A in $O(\text{nnz}(A))$

Sparse Embedding Matrix

S : Count Sketch matrix $r \times n$

For each of n cols S_{xi}

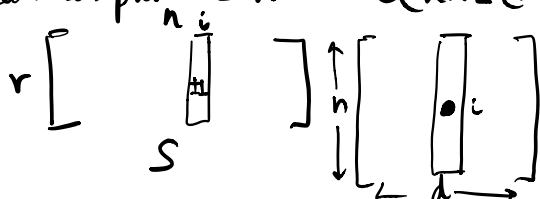
choose uniform random row $h(i) \in \{1, 2, \dots, r\}$

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

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

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

Can compute SA in $O(\text{nnz}(A))$



Thm S : Count Sketch matrix

$r = O\left(\frac{d^2}{\epsilon^2 \delta}\right)$ rows

For any fixed A , S is $(\pm\epsilon)$ - l_2 subspace embedding for A w. prob $1-\delta$

h : pairwise indpt

σ : 4-wise indpt.

Proof Sketch:

$$\begin{aligned} \Pr[\|I_d - U^T S^T S U\|_2 \geq \epsilon] &= \Pr[\|I_d - U^T S^T S U\|_2^l \geq \epsilon^l] \text{ i.i.d.} \\ &\leq \frac{1}{\epsilon^l} E[\|I_d - U^T S^T S U\|_2^l] \\ &\leq \frac{1}{\epsilon^l} E[\text{tr}((I_d - U^T S^T S U)^l)] \end{aligned}$$

Simpler Proof [Nguyen]

Approx. Matrix Multiplication

Defn C is ϵ -approx matrix product of A, B if

$$\|A^T B - C\|_F \leq \epsilon \|A\|_F \|B\|_F$$

Mountain sketch S_A, S_B, S $r \times n$ matrix

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

Defn: Distribⁿ D on $S \in \mathbb{R}^{k \times d}$ satisfies

(ϵ, δ, l) -JL moment property if

$$\forall x \in \mathbb{R}^d, \|x\|_2 = 1 \quad E[(\|Sx\|_2^2 - 1)^l] \leq \epsilon^l \delta$$

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 $l \geq 2$, $\epsilon, \delta \in (0, 1/2)$
 D : distribⁿ that satisfies (ϵ, δ, l) -JL moment property
 For A, B with d rows

$$P_{S \sim D} [\|A^T S^T S B - A^T B\|_F > 3\epsilon \|A\|_F \|B\|_F] \leq \delta$$

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

$$\langle Sx, Sy \rangle = \frac{1}{2} (\|Sx\|_2^2 + \|Sy\|_2^2 - \|S(x-y)\|_2^2)$$

$$\| \langle Sx, Sy \rangle - \langle x, y \rangle \|_l = \frac{1}{2} \| (\|Sx\|_2^2 - 1) + (\|Sy\|_2^2 - 1) - (\|S(x-y)\|_2^2 - \|x-y\|_2^2) \|_l$$

$$\leq \frac{1}{2} (\| \|Sx\|_2^2 - 1 \|_l + \| \|Sy\|_2^2 - 1 \|_l + \| \|S(x-y)\|_2^2 - \|x-y\|_2^2 \|_l)$$

$$\leq \frac{1}{2} (\epsilon \delta^{1/l} + \epsilon \delta^{1/l} + \|x-y\|_2^2 \epsilon \delta^{1/l})$$

$$\leq 3\epsilon \delta^{1/l}$$

For arbitrary x, y

$$\| \langle Sx, Sy \rangle - \langle x, y \rangle \|_l \leq 3\epsilon \delta^{1/l} \|x\|_2 \|y\|_2$$

(i, j) th entry of $A^T B$ is (A^i, B^j)
 \uparrow i th col \uparrow j th col

$$\| \|A^T S^T S B - A^T B\|_F^2 \|_l \leq \sum_{i,j} \| (\langle SA^i, SB^j \rangle - \langle A^i, B^j \rangle)^2 \|_l$$

$$\leq (3\epsilon \delta^{1/l})^2 \sum_{i,j} \|A^i\|_2^2 \|B^j\|_2^2$$

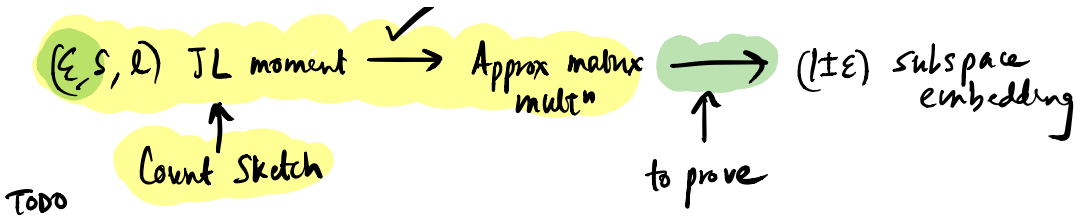
$$= (3\epsilon \delta^{1/l})^2 \|A\|_F^2 \|B\|_F^2$$

$$\mathbb{E} [\|A^T S^T S B - A^T B\|_F^l] \leq (3\epsilon \delta^{1/l})^l \|A\|_F^l \|B\|_F^l$$

$$P [\|A^T S^T S B - A^T B\|_F^l > (3\epsilon)^l \|A\|_F^l \|B\|_F^l]$$

$$\leq \frac{1}{(3\epsilon \|A\|_F \|B\|_F)^l} \mathbb{E} [\|A^T S^T S B - A^T B\|_F^l]$$

$$\leq \delta$$



Todo

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

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

$h: [d] \rightarrow [r]$ 2-wise indep
 $\sigma: [d] \rightarrow \{\pm 1\}$ 4-wise indep.

Pf: For any unit vector $x \in \mathbb{R}^d$
 $E_S[(\|Sx\|_2^2 - 1)^2] = E_S[\|Sx\|_2^4 - 2E_S[\|Sx\|_2^2] + 1] \stackrel{\text{to prove}}{\leq} \epsilon^2 \delta$

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

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

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

Proof: (Count Sketch gives l_2 -subspace embedding)

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

U orthonormal basis for cols of A

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

Apply with $\epsilon' = \epsilon/d$ $r = \frac{2}{(\epsilon')^2} \delta = \frac{2d^2}{\epsilon^2 \delta}$

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

$$\Rightarrow P[\|u^T S^T S u - I_d\|_2 > 3\epsilon] \leq \delta$$

Woodruff: Sketching As a Tool for Numerical Linear Algebra