

Tuesday, March 31

- James Enouen (USC)
  - ↳ Estimating Shapley values (my research interest)
  - ↳ Thursday 12pm in Roberts North 15

Today

Suyog Soti (Databricks) !!  
😊

↳ Fairview → CU Boulder → Google → Databricks

↳ real load balancing

Thursday, April 2

## Power Method

Motivation: Compute  $\lambda_k$  in  $\tilde{O}(ndk)$  time instead of  $O(nd^2)$  time

$$X^T X = \sum_{i=1}^d v_i v_i^T \sigma_i^2 = A = \sum_{i=1}^d v_i v_i^T \lambda_i$$

Goal: Compute  $z$  s.t.  $\|z - Av\|_2 \leq \epsilon$  in  $O\left(nd \frac{\log d/\epsilon}{\gamma}\right)$  time

$$z^{(0)} \sim \mathcal{N}(0, I_d)$$

$$z^{(0)} = c_1^{(0)} v_1 + c_2^{(0)} v_2 + \dots + c_d^{(0)} v_d$$

since  $v_1, \dots, v_d$  is a basis

for  $t=1, \dots, m$

$$n^{(t)} = \|z^{(t)}\|_2$$

$$z^{(t+1)} = A z^{(t)} / n^{(t)}$$

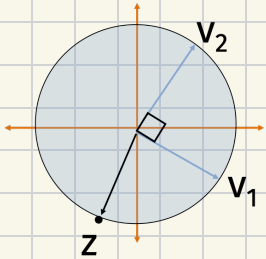
Return  $z^{(m)}$

where

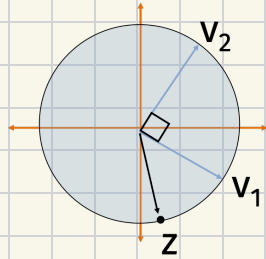
$$m = \frac{\log(d^{3.5}/\epsilon)}{\gamma}$$

# Geometric Intuition

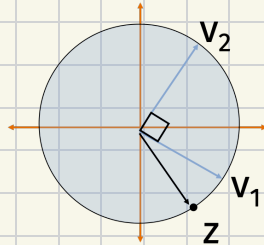
0 iterations



1 iterations



2 iterations



$$z^{(t)} = \frac{1}{\prod_{\eta}^{(t)}} A^m z^{(0)} =$$

$$z^{(q)} = \frac{1}{\prod_t \eta^{(t)}} A^m z^{(0)}$$

Fact:  $\left| \frac{c_j^{(0)}}{c_1^{(0)}} \right| \leq d^3$  whp

Fact:  $(1-\gamma)^{1/\gamma} \leq 1/e$

$$\gamma = \frac{\lambda_1 - \lambda_2}{\lambda_1} \leq \frac{\lambda_1 - \lambda_j}{\lambda_1}$$

$$\left| \frac{c_j^{(q)}}{c_1^{(q)}} \right| = \left| \frac{c_j^{(0)}}{c_1^{(0)}} \right| \left( \frac{\lambda_j}{\lambda_1} \right)^q \leq d^3 \left( 1 - \frac{\lambda_1 - \lambda_j}{\lambda_1} \right)^m \leq d^3 \left( (1-\gamma)^{1/\gamma} \right)^{\log(d^3 \sqrt{d/\epsilon})} \leq \sqrt{\epsilon/d}$$

Since  $\sum_{i=1}^d (c_i^{(q)})^2 = 1 = (c_1^{(q)})^2 \left( 1 + \sum_{i=2}^d \left| \frac{c_i^{(q)}}{c_1^{(q)}} \right|^2 \right) \leq (c_1^{(q)})^2 (1 + \epsilon/2)$

$\Rightarrow \left| c_1^{(q)} \right| \geq 1 - \epsilon$  then  $\| z^{(q)} - v_1 \|_2^2 = 2 - 2 \langle z^{(q)}, v_1 \rangle \stackrel{\text{wlog}}{\leq} 2 - 2(1-\epsilon) \leq 2\epsilon$

↑  
hiding  
some constants

## Krylov Subspace

Power method just uses last vector. Can we do better?

$$K = \{z^{(0)}, A z^{(0)}, A^2 z^{(0)}, \dots, A^{m-1} z^{(0)}\}$$

## Lanczos Method

$z^{(0)} \sim N(0, I)$  initial start

Compute  $Q \in \mathbb{R}^{d \times m}$ , an orthonormal basis for  $K$

Form  $T = Q^T A Q$  ( $T$  is tridiagonal, prove for hw)

Let  $z$  be top eigenvector of  $T$

Return  $y^* = Q z$

Analysis:

1.  $y^* = Qz$  is the best vector in  $K$  i.e.,

$$\begin{aligned} y^* &= \underset{y}{\operatorname{argmin}} \|X - Xy y^T\|_F^2 \\ &= \underset{y}{\operatorname{argmin}} \|X\|_F^2 - \|Xy y^T\|_F^2 = \end{aligned}$$

2. There is a very "good" vector  $w$  in  $K$ . All  $w$  in  $K$

$$\begin{aligned} w &= \alpha_1 z^{(0)} + \alpha_2 A z^{(0)} + \dots + \alpha_m A^{m-1} z^{(0)} \\ &= p(A) z^{(0)} \\ &= \sum_{i=1}^d p(\lambda_i) v_i v_i^T z^{(0)} \\ &= \sum_{i=1}^d p(\lambda_i) v_i c_i^{(0)} \end{aligned}$$

Goal: Find  $p(\lambda)$  s.t.  $p(\lambda_j)$  small for  $j \neq i$  but  $p(\lambda_i)$  large

Answer: Chebyshev polynomial!

Claim: There is degree  $m = O(\sqrt{\frac{1}{\epsilon}} \log \frac{d}{\epsilon})$

Chebyshev  $p$  where  $p(1) = 1$  but

$$|p(\lambda)| \leq \epsilon \quad \text{for} \quad 0 \leq \lambda \leq 1 - \epsilon$$

## Efficient Lanczos

$$T = \begin{bmatrix} T_{11} & T_{12} & 0 & 0 & 0 \\ T_{21} & T_{22} & T_{23} & 0 & 0 \\ 0 & T_{32} & T_{33} & T_{34} & 0 \\ 0 & 0 & T_{43} & T_{44} & T_{45} \\ 0 & 0 & 0 & T_{54} & T_{55} \end{bmatrix}$$

Goal: Build  $Q$  and  $T$  efficiently

$$k = \{z^{(0)}, Az^{(0)}, A^2 z^{(0)}, \dots, A^{m-1} z^{(0)}\}$$

$q_1, \dots, q_t, q_{t+1}, \dots, q_m$  are orthonormal cols. of  $Q$

$q_t$  in the span of  $z^{(0)}, Az^{(0)}, \dots, A^{t-1} z^{(0)}$

Since  $A$  symmetric,  
 $T = Q^T A Q$  is tridiagonal

## Naive (Gram-Schmidt)

$$q_1 = z^{(0)}$$

for  $t=1, \dots, m-1$ :

$$q_t = A^{t-1} z^{(0)} - \sum_{i=1}^{t-1} \langle A^{t-1} z^{(0)}, q_i \rangle q_i$$

$$q_t = q_t / \|q_t\|_2$$

## Clever

$$q_t = z^{(0)}$$

for  $t=1, \dots, m-1$ :

$$q_t = A^{t-1} z^{(0)} - \langle q_{t-1}, A^{t-1} z^{(0)} \rangle q_{t-1} - \langle q_{t-2}, A^{t-1} z^{(0)} \rangle q_{t-2}$$

$$q_t = q_t / \|q_t\|_2$$

Why is this enough?

Hint: Inspect  $AQ = QT$