

Theorem 0.1. *Let D be a negative type matrix. Let $LRO_k(S)$ denote the optimal low-rank approximation of S in the Frobenius norm. Then in $O(n \text{poly} \frac{k}{\varepsilon})$ time, we can compute matrix D' of rank k such that:*

$$\|D - D'\|_F \leq (1 + \varepsilon) \|D - LRO_{k-2}(D)\|_F \quad (1)$$

To do this, we will use Musco and Woodruff and a careful use of Cauchy's Interlacing Theorem.

Theorem 0.2. *(Musco and Woodruff) Let M be a positive semi-definite matrix. Then we can compute a matrix M' of rank k in $O(n \text{poly} \frac{k}{\varepsilon})$ time such that:*

$$\|M - M'\|_F \leq (1 + \varepsilon) \|M - LRO_k(M)\|_F \quad (2)$$

Cauchy's Interlacing theorem, as stated in lemma 3.4 of <https://arxiv.org/pdf/1408.4421v1.pdf>:

Theorem 0.3. *(Optimal Low Rank Approximation of Matrices): Let μ_1, \dots, μ_n be eigenvectors of symmetric matrix S , where $|\mu_1| \leq \dots \leq |\mu_n|$. Then:*

$$S - LRO_k(S) = l_2(\mu_{k+1}, \dots, \mu_n)$$

We state the above theorem without proof. Here, l_2 represents the l_2 norm.

Theorem 0.4. *(Cauchy's Interlacing Theorem) Let λ_i^X denote the i^{th} largest eigenvalues of X for any matrix X . If A is a symmetric matrix and v is a vector, then the eigenvalues of $A + vv^T$ satisfy:*

$$\lambda_i^A \leq \lambda_i^{A+vv^T} \leq \lambda_{i+1}^A \quad (3)$$

$$\lambda_i^{A+vv^T} \leq \lambda_{i+1}^A \leq \lambda_{i+1}^{A+vv^T} \quad (4)$$

Corollary 0.4.1.

$$\lambda_{i-1}^A \leq \lambda_i^{A+vv^T-ww^T} \leq \lambda_{i+1}^A \quad (5)$$

$$\lambda_{i-1}^{A+vv^T-ww^T} \leq \lambda_i^A \leq \lambda_{i+1}^{A+vv^T-ww^T} \quad (6)$$

This follows from Theorem 0.4.

Definition 0.5. *Let $M := [d_{ij} - d_{0i} - d_{0j}]$, where d_{ij} is the ij entry of D . Let $A_1 = [d_{0i}]$, and $A_2 = [d_{0j}]$, where A_1 and A_2 are n by n matrices. By construction, $D = M + A_1 + A_2$, and sampling from M can be done quickly by taking three samples from D .*

Let v be the vector with entries $(1 + d_{0i})/2$, and w be the vector with entries $(1 - d_{0i})/2$.

Lemma 0.6. $A1 + A2 = vv^T - ww^T$

(Proof omitted. This is straightforward computation I think.)

Lemma 0.7. $D = M + vv^T - ww^T$

Proof. This follows from lemma 0.6. □

Corollary 0.7.1. *Using D and M as defined in 0.5:*

$$\lambda_{i-1}^M \leq \lambda_i^D \leq \lambda_{i+1}^M \quad (7)$$

$$\lambda_{i-1}^D \leq \lambda_i^M \leq \lambda_{i+1}^D \quad (8)$$

Proof. This follows from corollary 0.4.1 and Lemma 0.6 □

Now, we list basic facts about negative type distances D .

Theorem 0.8.

$$\begin{aligned} \lambda_1^D &\leq \dots \lambda_{n-1}^D \leq 0 < \lambda_n^D \\ \lambda_n^D &\geq |\lambda_1^D| \geq \dots |\lambda_{n-1}^D| \geq 0 \end{aligned}$$

The second chain of inequalities holds since D has at most one positive eigenvalue (known, I can show you how to prove this offline). Also, the trace of D is 0, which means the positive eigenvalue is equal to the negative sum of the negative eigenvalues.

Theorem 0.9. M is negative-semidefinite, and so:

$$\begin{aligned} \lambda_1^M &\leq \dots \leq \lambda_n^M = 0 \\ |\lambda_1^M| &\geq \dots |\lambda_n^M| \geq 0 \end{aligned}$$

That M is negative-semidefinite, is well established in the literature.

Theorem 0.10. *For all $i < n$:*

$$|\lambda_{i-1}^M| \geq |\lambda_i^D| \geq |\lambda_{i+1}^M| \quad (9)$$

For all $2 \leq i < n - 1$:

$$|\lambda_{i-1}^D| \geq |\lambda_i^M| \geq |\lambda_{i+1}^D| \quad (10)$$

Proof. This follows from Corollary 0.7.1. □

Theorem 0.11. *Let M' be the output of Woodruff and Musco, applied to M , where M' is rank $k - 2$. Then let $D' = M' + vv^T - ww^T$. Recall that $D = M + vv^T - ww^T$. Then*

$$\begin{aligned} \|D - D'\|_F &= \|M - M'\| \leq (1 + \varepsilon) \|M - LRO_{k-2}(M)\| \\ &= (1 + \varepsilon) \cdot l_2(\lambda_{k-1}^M, \dots, \lambda_n^M) \end{aligned}$$

Theorem 0.12. *For $j \geq 1$,*

$$\|D - LRO_j(D)\|_F = l_2(\lambda_j^D, \dots, \lambda_{n-1}^D)$$

Theorem 0.13.

$$l_2(\lambda_{k-1}^M, \dots, \lambda_n^M) \leq l_2(\lambda_{k-2}^D, \dots, \lambda_{n-1}^D) = \|D - LRO_{k-2}(D)\|_F$$

Proof. The first inequality follows from Theorem 0.10. □

Theorem 0.14.

$$\|D - D'\|_F \leq (1 + \varepsilon) \|D - LRO_{k-2}(D)\|_F$$

As desired. Note that I think we can actually bound $\|D - D'\|_F$ with $\|D - LRO_{k-1}(D)\|_F$ with more careful eigenvalue bounding, but whatever. Note that D' can be computed quickly (M' can be obtained quickly by sampling from D , and v and w can be computed in $O(n)$ time.)