

Sparse Low Rank Approximation in a Stream

Let $\mathbf{A} \in \mathbb{R}^{n \times d}$. Recall the usual low rank approximation problem:

$$\min_{\mathbf{X}} \|\mathbf{A} - \mathbf{X}\|_F^2. \tag{1}$$

where \mathbf{X} ranges over rank k matrices. However, \mathbf{X} returned by the above minimization problem can be dense, and it is often desirable to find a *sparse* \mathbf{X} . In the *sparse low rank approximation* problem, we seek to minimize Equation (1) over \mathbf{X} which can be written as the sum of k rank 1 matrices, each of which is supported on a $s \times s$ submatrix. We write $\mathcal{S}_{s,k}$ for the set of such matrices.

It can be shown that solving the above problem exactly is NP hard! For polynomial time algorithms, we allow for the following relaxations:

- We only seek $(1 + \varepsilon)$ -approximations
- We allow for the matrix \mathbf{X} to be supported on an $O(sk/\varepsilon) \times O(sk/\varepsilon)$ submatrix

That is, we seek a rank k matrix \mathbf{Y} supported on a $O(sk/\varepsilon) \times O(sk/\varepsilon)$ submatrix such that

$$\|\mathbf{A} - \mathbf{Y}\|_F^2 \leq (1 + \varepsilon) \min_{\mathbf{X} \in \mathcal{S}_{s,k}} \|\mathbf{A} - \mathbf{X}\|_F^2$$

Our approach is as follows:

- Identify $O(sk/\varepsilon)$ heavy rows and columns
- Obtain an entry-wise estimate $\hat{\mathbf{A}}$ to \mathbf{A} , and restrict it to the heavy rows and columns
- Output a rank k approximation of $\hat{\mathbf{A}}$ on the restriction

Let \mathbf{B} an optimal matrix in $\mathcal{S}_{s,k}$, let W be an $sk \times sk$ support containing \mathbf{B} , and let X be the part of W lying in columns of squared norm at most τ .

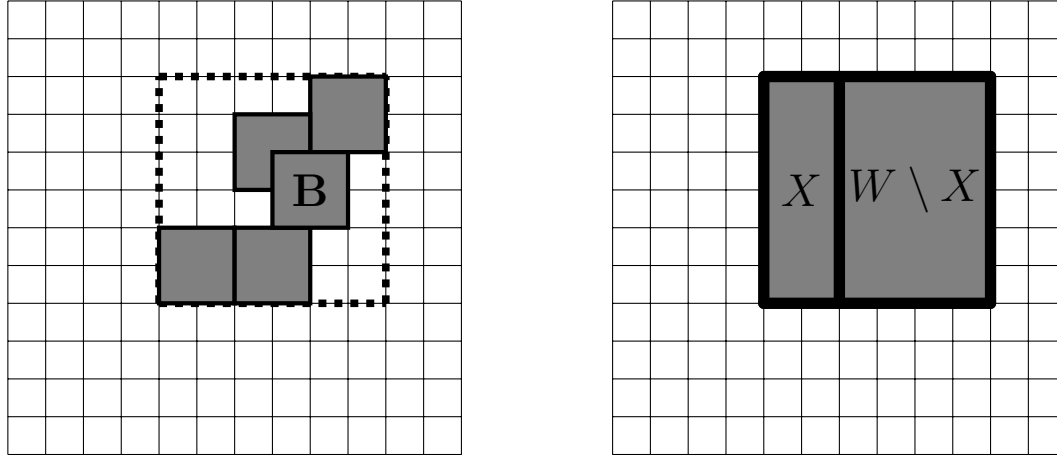


Figure 1: The supports $W, X, W \setminus X$ to be used in the proof.

Exercise. Show that

$$\|\mathbf{A} - (\mathbf{A} |_{\mathcal{W}})_k\|_F^2 \leq \|\mathbf{A} - \mathbf{B}\|_F^2.$$

Exercise. Show that

$$\begin{aligned} \|\mathbf{A} - (\mathbf{A} \mid_W)_k\|_F^2 &= \|\mathbf{A} - (\mathbf{A} \mid_{W \setminus X})_k\|_F^2 - \|\mathbf{A} \mid_X\|_F^2 \\ &\quad + \|\mathbf{A} \mid_W - (\mathbf{A} \mid_W)_k\|_F^2 - \|\mathbf{A} \mid_{W \setminus X} - (\mathbf{A} \mid_{W \setminus X})_k\|_F^2 \end{aligned}$$

Exercise. Show that

$$\|\mathbf{A}|_{W \setminus X} - (\mathbf{A}|_{W \setminus X})_k\|_F^2 \leq \|\mathbf{A}|_W - (\mathbf{A}|_W)_k\|_F^2$$

Exercise. Conclude that there exists a rank k matrix \mathbf{C} supported on columns of squared norm at least τ such that

$$\|\mathbf{A} - \mathbf{C}\|_F^2 \leq \min_{\mathbf{X} \in \mathcal{S}_{s,k}} \|\mathbf{A} - \mathbf{X}\| + sk\tau$$

Fact. By a simple modification of the analysis of **CountSketch**, one can identify the rows $i \in [n]$ such that

$$\|\mathbf{e}_i^\top \mathbf{A}\|_2^2 \geq \alpha \left\| \mathbf{A}_{[1/\alpha],*} \right\|_F^2$$

with $O(\alpha^{-1} \log^2 n)$ bits of space.

Exercise. Show how to identify a $O(sk/\varepsilon) \times O(sk/\varepsilon)$ submatrix $S \times T$ that contains a rank k matrix \mathbf{C} such that

$$\|\mathbf{A} - \mathbf{C}\|_F^2 \leq (1 + \varepsilon) \min_{\mathbf{X} \in \mathcal{S}_{s,k}} \|\mathbf{A} - \mathbf{X}\|_F^2$$

using $O(\varepsilon^{-1}sk \log^2 n)$ bits of space.

Exercise. Show how to recover a matrix $\hat{\mathbf{A}}$ such that

$$\left\| \hat{\mathbf{A}}|_{S \times T} - \mathbf{A}|_{S \times T} \right\|_{\infty}^2 \leq \frac{\varepsilon^4}{s^2 k^2} \cdot \min_{\mathbf{X} \in \mathcal{S}_{s,k}} \|\mathbf{A} - \mathbf{X}\|_F^2$$

using $O(\varepsilon^{-4} s^2 k^2 \log n)$ bits of space. Conclude that

$$\left\| \hat{\mathbf{A}}|_{S \times T} - \mathbf{A}|_{S \times T} \right\|_F^2 \leq \varepsilon^2 \cdot \min_{\mathbf{X} \in \mathcal{S}_{s,k}} \|\mathbf{A} - \mathbf{X}\|_F^2$$

Exercise. Show that

$$\left\| \mathbf{A} - (\hat{\mathbf{A}} |_{S \times T})_k \right\|_F^2 \leq (1 + \varepsilon) \min_{\mathbf{X} \in \mathcal{S}_{s,k}} \|\mathbf{A} - \mathbf{X}\|_F^2$$

