

Assignment 5

Randomization in Numerical Linear Algebra (PCMI)

1. Recall the definition of the edge-incidence matrix \mathbf{B} and the weight matrix \mathbf{W} (positive weights only) from slide 151. Recall that the Laplacian matrix of a graph can be expressed as $\mathbf{L} = \mathbf{B}^T \mathbf{W} \mathbf{B}$, and recall that the effective resistances of the edges of the graph are defined to be the diagonal entries of the matrix $\mathbf{B} \mathbf{L}^\dagger \mathbf{B}^T$. Prove that the effective resistances are equal (up to rescaling) to the row leverage scores of the matrix $\mathbf{W}^{1/2} \mathbf{B}$.

2. Let \mathbf{A} and $\tilde{\mathbf{A}}$ be two matrices with the same dimensions. Prove:

$$\left\| \mathbf{A} - \left(\tilde{\mathbf{A}} \right)_k \right\|_2 \leq \|\mathbf{A} - \mathbf{A}_k\|_2 + 2 \left\| \mathbf{A} - \tilde{\mathbf{A}} \right\|_2.$$

In the above, \mathbf{A}_k is the best rank- k approximation to \mathbf{A} (similarly for $\tilde{\mathbf{A}}$).

3. We will prove a bound for the spectral norm of the matrix $\mathbf{A} - \mathcal{S}_\Omega(\mathbf{A})$, where $\mathcal{S}_\Omega(\mathbf{A})$ is constructed using the element-wise sampling algorithm of slide 165 and the hybrid sampling probabilities of slide 175.

1. Let $\mathbf{M}_t = \frac{\mathbf{A}_{i_t j_t}}{p_{i_t j_t}} \mathbf{e}_{i_t} \mathbf{e}_{j_t}^T - \mathbf{A}$. Prove:

$$\mathbf{A} - \mathcal{S}_\Omega(\mathbf{A}) = -\frac{1}{s} \sum_{t=1}^s \mathbf{M}_t.$$

Also compute $\mathbf{E}[\mathbf{M}_t]$.

2. Compute the expectation $\mathbf{E}[\mathbf{M}_t \mathbf{M}_t^T]$.

3. Use the sampling probabilities of slide 175 to bound $\left\| \mathbf{E}[\mathbf{M}_t \mathbf{M}_t^T] \right\|_2$ and $\max_{i,j} \left\| \frac{\mathbf{A}_{ij}}{p_{ij}} \mathbf{e}_i \mathbf{e}_j^T - \mathbf{A} \right\|_2$.

4. Apply the matrix-Bernstein inequality of slide 177 to bound the spectral norm of the matrix $\mathbf{A} - \mathcal{S}_\Omega(\mathbf{A})$ with high probability.