

Exercise 1:

a) We have

$$\|x_j - x^*\|_2^2 = \|A_u^\dagger(b_u - A_u x_{j-1}) + x_{j-1} - x^*\|_2^2 = \|A_u^\dagger b_u - A_u^\dagger A_u x_{j-1} + x_{j-1} - x^*\|_2^2$$

And

$$\|(I - A_u^\dagger A_u)(x_{j-1} - x^*)\|_2^2 = \|A_u^\dagger A_u x^* - A_u^\dagger A_u x_{j-1} + x_{j-1} - x^*\|_2^2$$

Since x^* is the optimal solution and since we assume the least square problem is consistent we have

$$Ax^* = b \Rightarrow A_u x^* = b_u$$

Which proves

$$\|x_j - x^*\|_2^2 = \|(I - A_u^\dagger A_u)(x_{j-1} - x^*)\|_2^2$$

□

b) We have that $A_u^\dagger A_u$ is an orthogonal projector (since it's idempotent and symmetric). Then any vector $y \in \mathbb{R}^n$ can be rewritten as $y = A_u^\dagger A_u y + (I - A_u^\dagger A_u)y$, so $(I - A_u^\dagger A_u)$ is the orthogonal complement projection matrix to $A_u^\dagger A_u$.

This implies

$$\|y\|_2^2 = \|A_u^\dagger A_u y\|_2^2 + \|(I - A_u^\dagger A_u)y\|_2^2$$

And hence

$$\|(I - A_u^\dagger A_u)y\|_2^2 = \|y\|_2^2 - \|A_u^\dagger A_u y\|_2^2$$

Then we have:

$$\frac{\sigma_{\min}^2(A)}{\beta m} \|y\|_2^2 \leq \frac{\sum_{j=1}^m \sigma_{\min}^2(A_{u_j})}{\beta m} \|y\|_2^2$$

Using the boundedness of singular values:

$$\|A_u\|_2^2 = \sigma_{\max}(A_u)^2 = \lambda_{\max}(A_u A_u^\top) \leq \beta$$

We get

$$\frac{\sum_{j=1}^m \sigma_{\min}^2(A_{u_j})}{\beta m} \|y\|_2^2 \leq \sum_{j=1}^m \frac{\sigma_{\min}^2(A_{u_j})}{\sigma_{\max}^2(A_{u_j}) m} \|y\|_2^2 = \sum_{j=1}^m \frac{1}{\kappa^2(A_{u_j}) m} \|y\|_2^2$$

Where $\kappa(A)$ denotes the consistency of A .

Using the following consistency bound:

$$\|A^\dagger A y\|_2 \geq \frac{\|y\|_2}{\kappa(A)} = \frac{\|y\|_2}{\|A\|_2 \|A^\dagger\|_2},$$

we finally get:

$$\sum_{j=1}^m \frac{1}{\kappa^2(A_{u_j})m} \|y\|_2^2 \leq \sum_{j=1}^m \frac{1}{m} \|A_{u_j}^\dagger A_{u_j} y\|_2^2 = \mathbb{E} \|A_u^\dagger A_u y\|_2^2$$

Since each u_j is chosen uniformly at random. Putting everything together we have

$$\mathbb{E} \|(I - A_u^\dagger A_u) y\|_2^2 = \|y\|_2^2 - \mathbb{E} \|A_u^\dagger A_u y\|_2^2 \leq (1 - \frac{\sigma_{\min}^2(A)}{\beta m}) \|y\|_2^2$$

Which is the desired bound! \square

Using part a) and this result (and supposing that the choice of u is independent from one iteration to another) we finally have

$$\begin{aligned} \mathbb{E} \|x_j - x^*\|_2^2 &= \mathbb{E} \|(I - A_u^\dagger A_u)(x_{j-1} - x^*)\|_2^2 \leq (1 - \frac{\sigma_{\min}^2(A)}{\beta m}) \mathbb{E} \|x_{j-1} - x^*\|_2^2 \\ &\leq (1 - \frac{\sigma_{\min}^2(A)}{\beta m})^j \|x_0 - x^*\|_2^2 \end{aligned}$$

Which completes the proof. \square

Exercise 2:

a) Let us prove that $\psi(L)$ is an orthogonal projection.

We first prove that $\psi(L)$ is idempotent:

$$\psi(L)^2 = \left((L^\dagger)^{\frac{1}{2}} L (L^\dagger)^{\frac{1}{2}} \right) \left((L^\dagger)^{\frac{1}{2}} L (L^\dagger)^{\frac{1}{2}} \right) = (L^\dagger)^{\frac{1}{2}} L L^\dagger L (L^\dagger)^{\frac{1}{2}} = (L^\dagger)^{\frac{1}{2}} L (L^\dagger)^{\frac{1}{2}} = \psi(L)$$

Let us now prove symmetry:

$$\psi(L)^\top = ((L^\dagger)^{\frac{1}{2}})^\top L^\top ((L^\dagger)^{\frac{1}{2}})^\top$$

Since G is undirected, L is symmetric.

We then have from the properties of the pseudo-inverse:

$$\begin{cases} (LL^\dagger)^\top = LL^\dagger \Rightarrow (L^\dagger)^\top L = LL^\dagger \\ (L^\dagger L)^\top = L^\dagger L \Rightarrow L(L^\dagger)^\top = L^\dagger L \end{cases}$$

Now recall that the pseudo-inverse is uniquely defined, let us check that $(L^\dagger)^\top$ satisfies all for defining properties of the pseudo-inverse of L :

$$\begin{cases} L(L^\dagger)^\top L = (LL^\dagger L)^\top = (L)^\top = L \\ (L^\dagger)^\top L (L^\dagger)^\top = (L^\dagger L L^\dagger)^\top = (L^\dagger)^\top \\ ((L^\dagger)^\top L)^\top = LL^\dagger = (L^\dagger)^\top L \\ (L(L^\dagger)^\top)^\top = L^\dagger L = L(L^\dagger)^\top \end{cases}$$

By uniqueness of the pseudo-inverse, this implies $(L^\dagger)^\top = (L^\dagger)$.

Then since L is a graph Laplacian, it is positive semi-definite (PSD), which implies that L^\dagger is also PSD, hence $(L^\dagger)^{\frac{1}{2}}$ is also symmetric (and PSD).

Thus

$$\psi(L)^\top = \psi(L)$$

Which proves that $\psi(L)$ is an orthogonal projector. \square

For the second part of the question, we start with the definition of S being an ϵ -spectral approximation of L :

$$(1 - \epsilon)L \leq S \leq (1 + \epsilon)L \quad \Leftrightarrow \quad -\epsilon L \leq S - L \leq \epsilon L$$

Now let us prove that ψ is a monotone operator (i.e. it preserves inequalities). Let $B \geq A$, ($B - A$ is PSD), we have $\psi(B) - \psi(A) = \psi(B - A)$. Since $(L^\dagger)^{\frac{1}{2}}$ is PSD, $\psi(B - A)$ must be PSD as well which proves ψ is monotone.

We then have,

$$-\epsilon L \leq S - L \leq \epsilon L \quad \Leftrightarrow \quad -\epsilon \psi(L) \leq \psi(S - L) \leq \epsilon \psi(L)$$

$$\Leftrightarrow \|\psi(S - L)\|_2 \leq \epsilon \|\psi(L)\|_2 = \epsilon,$$

since $\psi(L)$ is an orthogonal projector, $\|\psi(L)\| = 1$.

- b) In this section we shorthand $E := E_{ii} + E_{jj} - E_{ij} - E_{ji}$.

We first notice that E is a rank-one matrix.

Indeed, it has only two non-zero columns, one equal to $e_i - e_j$ the other $e_j - e_i$, where e_i denotes a zero-vector with a one in the i th coordinate. Multiplying one column by -1 gives the other column, hence they are linear multiples of each other, and the matrix is rank one.

Since $(L^\dagger)^{\frac{1}{2}}$ is non-zero, it has rank at least one, and thus $\psi(E)$ has rank one.

This means all of its eigenvalues but one are equal to zero.

The non-zero eigenvalue is equal to two, indeed:

$$E(e_i - e_j) = 2(e_i - e_j)$$

which shows E is PSD.

Using similar arguments as in part a, this implies that $\psi(E)$ is also PSD.

Since $\psi(E)$ has only one non-zero eigenvalue,

$$\text{tr}(\psi(E)) = \lambda_{\max}(\psi(E)) = \|\psi(E)\|_2.$$

We then recall that the matrix representing the graph Laplacian L for a connected graph has rank $n - 1$.

Then recall from part a that $\psi(L)$ is an orthogonal projector. This implies that it has

eigenvalues either 0 or 1.

Combining these two facts, we have that

$$\mathrm{tr}(\psi(L)) = \sum_{i=1}^{n-1} 1 = n - 1.$$

We then have

$$\begin{aligned} \sum_{(i,j) \in E} p_{ij} &= \sum_{(i,j) \in E} \frac{w_{ij}}{n-1} \|\psi(E_{ii} + E_{jj} - E_{ij} - E_{ji})\| = \sum_{(i,j) \in E} \frac{w_{ij} \mathrm{tr}(\psi(E_{ii} + E_{jj} - E_{ij} - E_{ji}))}{n-1} \\ &= \frac{\mathrm{tr}(\psi(\sum_{(i,j) \in E} w_{ij}(E_{ii} + E_{jj} - E_{ij} - E_{ji})))}{n-1} = \frac{\mathrm{tr}(\psi(L))}{n-1} = 1 \end{aligned}$$

□

c)

Exercise 3:

a)

b)

c)

d)

e)