

2 Use the min-max theorem to show that $\|A\|_2 = \sigma_1(A)$ - the largest singular value of A .

Solution: This comes from the fact that:

$$\begin{aligned}\|A\|_2^2 &= \max_{x \neq 0} \frac{\|Ax\|_2^2}{\|x\|_2^2} \\ &= \max_{x \neq 0} \frac{(Ax, Ax)}{(x, x)} \\ &= \max_{x \neq 0} \frac{(A^T Ax, A)}{(x, x)} \\ &= \lambda_{max}(A^T A) \\ &= \sigma_1^2\end{aligned}$$




3 Suppose that $A = LDL^T$ where L is unit lower triangular, and D diagonal. How many negative eigenvalues does A have?

Solution: It has as many negative eigenvalues as there are negative entries in D

4 Assume that A is tridiagonal. How many operations are required to determine the number of negative eigenvalues of A ?

Solution: The rough answer is $O(n)$ – because an LU (and therefore LDLT) factorization costs $O(n)$. Based on doing the LU factorization of a triangular matrix, a more accurate answer is $3n$ operations.

5 Devise an algorithm based on the inertia theorem to compute the i -th eigenvalue of a tridiagonal matrix.

Solution: Here is a matlab script:

```
function [sigma] = bisect(d, b, i, tol)
%% function [sigma] = bisect(d, b, i, tol)
%% d      = diagonal of T
```

```

%% b    = co-diagonal
%% i    = compute i-th eigenvalue
%% tol  = tolerance used for stopping
    b(1) = 0;
    n = length(d);
%%----- guershgorin
tmin = d(n) - abs(b(n));
tmax = d(n) + abs(b(n));
for j=1:n-1
    rho = abs(b(j)) + abs(b(j+1));
    tmin = min(tmin, d(j)-rho);
    tmax = max(tmax, d(j)+rho);
end
tol = tol*(tmax-tmin);
for iter=1:100
    sigma = 0.5*(tmin+tmax);
    count = sturm(d, b, sigma);
    if (count >= i)
        tmin = sigma;
    else
        tmax = sigma;
    end
    if (tmax - tmin) < tol
        break
    end
end
end

```



 6 What is the inertia of the matrix

$$\begin{pmatrix} I & F \\ F^T & 0 \end{pmatrix}$$

where F is $m \times n$, with $n < m$, and of full rank?

[Hint: use a block LU factorization]

Solution: We start with

$$\begin{aligned} \begin{pmatrix} I & F \\ F^T & 0 \end{pmatrix} &= \begin{pmatrix} I & 0 \\ F^T & I \end{pmatrix} \begin{pmatrix} I & F \\ 0 & -F^T F \end{pmatrix} \\ &= \begin{pmatrix} I & 0 \\ F^T & I \end{pmatrix} \begin{pmatrix} I & 0 \\ 0 & -F^T F \end{pmatrix} \begin{pmatrix} I & F \\ 0 & I \end{pmatrix} \\ &= \begin{pmatrix} I & 0 \\ F^T & I \end{pmatrix} \begin{pmatrix} I & 0 \\ 0 & -F^T F \end{pmatrix} \begin{pmatrix} I & 0 \\ F^T & I \end{pmatrix}^T \end{aligned}$$

This is of the form \mathbf{XDX}^T where \mathbf{X} is invertible.. Therefore the inertia is the same as that of the block diagonal matrix which is: m positive eigenvalues (block \mathbf{I}) and n negative eigenvalues since $-\mathbf{F}^T\mathbf{F}$ is $n \times n$ and negative definite.

 7 Let $\|\mathbf{A}_O\|_I = \max_{i \neq j} |a_{ij}|$. Show that

$$\|\mathbf{A}_O\|_F \leq \sqrt{n(n-1)} \|\mathbf{A}_O\|_I$$

Solution: This is straightforward:

$$\|\mathbf{A}_O\|_F^2 = \sum_{i \neq j} |a_{ij}|^2 \leq n(n-1) \max_{i \neq j} |a_{ij}|^2 = n(n-1) \|\mathbf{A}_O\|_I^2.$$



 8 Use this to show convergence in the case when largest entry is zeroed at each step.

Solution: If we call \mathbf{B}_k the matrix \mathbf{A}_O after each rotation then we have according to result in the previous

page and using the previous exercise:

$$\begin{aligned}\|B_{k+1}\|_F^2 &= \|B_k\|_F^2 - 2a_{pq}^2 \\ &= \|B_k\|_F^2 - 2\|B_k\|_I^2 \\ &\leq \|B_k\|_F^2 - \frac{2}{n(n+1)}\|B_k\|_F^2 \\ &= \left[1 - \frac{2}{n(n+1)}\right] \|B_k\|_F^2\end{aligned}$$

which shows that the norm will be decreasing by factor less than a constant that is less than one - therefore it converges to zero. \square