

$$A^{(k+1)} = R^{(pq)}(\varphi_k)^T A^{(k)} R^{(pq)}(\varphi_k)$$

where p and q are such that

$$|A_{pq}^{(k)}| = \max \{ |A_{ij}^{(k)}| : i \neq j \}$$

and φ_k is chosen as in the $n = 2$ case so $A_{pq}^{(k+1)} = 0$ with the additional requirement to choose t to be smallest in absolute value from among the two possible choices given by the quadratic equation.

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function |jacobi(A)
    p,q=getpq(A)
    t1,t2=quadsolv(-A[p,q],A[p,p]-A[q,q],A[p,q])
    t=t1
    if abs(t2)<abs(t1)
        t=t2
    end
    c=1/sqrt(1+t^2)
    s=c*t
    B=copy(A)
    B[p,p]=A[p,p]*c^2-2*A[p,q]*s*c+A[q,q]*s^2
    B[q,q]=A[p,p]*s^2+2*A[p,q]*s*c+A[q,q]*c^2
    B[p,q]=0
    B[q,p]=0
    N,_=size(A)
    for i=1:N
        if i==p||i==q
            continue
        end
        B[i,p]=A[i,p]*c-A[i,q]*s
        B[p,i]=B[i,p]
        B[i,q]=A[i,p]*s+A[i,q]*c
        B[q,i]=B[i,q]
    end
    return B
end

```

✓ did the same thing

find the largest in magnitude

```
julia> Ak=copy(A)
```

5x5 Matrix{Float64}:

$$\begin{bmatrix} 8.52 & -1.8 & 1.08 & -3.27 & -3.25 \\ -1.8 & 1.76 & -6.67 & -2.67 & 4.05 \\ 1.08 & -6.67 & 1.16 & -8.34 & 2.78 \\ -3.27 & -2.67 & -8.34 & 7.18 & 6.27 \\ -3.25 & 4.05 & 2.78 & 6.27 & -3.28 \end{bmatrix}$$



Performs an orthogonal transformation using a plane rotation to put zeros where the largest was.

```
julia> Ak=jacobi(Ak)
```

5x5 Matrix{Float64}:

$$\begin{bmatrix} 8.52 & -1.8 & -0.995368 & -3.29675 & -3.25 \\ -1.8 & 1.76 & -6.99297 & 1.64807 & 4.05 \\ -0.995368 & -6.99297 & -4.69655 & 0.0 & 5.87835 \\ -3.29675 & 1.64807 & 0.0 & 13.0365 & 3.5336 \\ -3.25 & 4.05 & 5.87835 & 3.5336 & -3.28 \end{bmatrix}$$

$|A_{q,q}|$ was the largest

P

q

the rest
don't
change

q

these rows and
columns change

Today we show that $A^{(k)} \rightarrow D$ where D is a diagonal matrix with the eigenvalues of A.

The Trace of a matrix (The sum of the diagonal terms)

if $A \in \mathbb{R}^{n \times n}$

then $\text{trace}(A) = \sum_{i=1}^n A_{ii}$

The Spectral Norm of a Matrix

$\|A\|_2 = \max \left\{ \sqrt{\lambda_i} \right\}$ where λ_i are the eigenvalues of $B = A^T A$

Frobenius Norm

$$\|A\| = \sqrt{\sum_{i=1}^n \sum_{j=1}^n |A_{ij}|^2}$$

idea is treat the elements of A as if they were a vector and take the 2-vector norm

$$\text{Trace } B = \sum_{i=1}^n [A^T A]_{ii} = \sum_{i=1}^n \sum_{k=1}^n A_{ki} A_{ki}$$

$$[A^T A]_{ij} = \sum_{k=1}^n [A^T]_{ik} A_{kj} = \sum_{k=1}^n A_{ki} A_{kj}$$

$$\text{Trace } B = \sum_{i=1}^n \sum_{k=1}^n |A_{ki}|^2 \quad \text{since } A \in \mathbb{R}^{n \times n}$$

Therefore

$$\|A\|_F^2 = \text{trace}(A^T A) \quad \text{for } A \in \mathbb{R}^{n \times n}$$

Properties of trace $A, B \in \mathbb{R}^{n \times n}$

$$\text{trace } AB = \sum_{i=1}^n [AB]_{ii} = \sum_{i=1}^n \sum_{k=1}^n A_{ik} B_{ki}$$

$$[AB]_{ij} = \sum_{k=1}^n A_{ik} B_{kj}$$

the same

$$\text{trace } BA = \sum_{i=1}^n \sum_{k=1}^n B_{ik} A_{ki} = \sum_{k=1}^n \sum_{i=1}^n B_{ik} A_{ki}$$

$$= \sum_{i=1}^n \sum_{k=1}^n B_{ki} A_{ik} = \sum_{i=1}^n \sum_{k=1}^n A_{ik} B_{ki}$$

Therefore $\text{trace } AB = \text{trace } BA$.

$$A^{(k+1)} = R^{(pq)}(\varphi_k)^T A^{(k)} R^{(pq)}(\varphi_k)$$

$$A^{(k+1)} = R^{(pq)}(\varphi_k)^{-1} A^{(k)} R^{(pq)}(\varphi_k)$$

Consider $B = R^{-1} A R$ where R is an orthogonal matrix.

$$\text{trace } B = \text{trace } \underbrace{R^{-1} A R}_{B} = \text{trace } A R R^T = \text{trace } A$$

Therefore $\text{trace } A = \text{trace } A^{(1)} = \text{trace } A^{(2)} = \dots = \text{trace } A^{(k)}$



assume A is symmetric

$$A^T = A$$

$$\|A\|_F^2 = \text{trace}(A^T A) = \text{trace}(A^2)$$

$$\|B\|_F^2 = \text{trace}(B^T B) = \text{trace}(R^T A R)^T R^T A R$$

$$= \text{trace } R^T A^T R^{TT} R^T A R = \text{trace } R^T A^2 R$$

↑ Symmetric Cancel

$$\approx \text{trace } R^T A^2 R = \text{trace } A^2 R R^T = \text{trace } A^2 = \|A\|_F^2$$

Lemma 5.1 The sum of squares of the elements of a symmetric matrix is invariant under an orthogonal transformation: that is, if $A \in \mathbb{R}_{\text{sym}}^{n \times n}$

Theorem 5.3 Suppose that $A \in \mathbb{R}_{\text{sym}}^{n \times n}$, $n \geq 2$. In the classical Jacobi method the off-diagonal entries in the sequence of matrices $(A^{(k)})$, generated from $A^{(0)} = A$ according to Definition 5.3, converge to 0 in the sense that

$$\lim_{k \rightarrow \infty} \sum_{\substack{i,j=1 \\ i \neq j}}^n [(A^{(k)})_{ij}]^2 = 0. \quad (5.12)$$

Furthermore,

$$\lim_{k \rightarrow \infty} \sum_{i=1}^n [(A^{(k)})_{ii}]^2 = \text{Trace}(A^2). \quad (5.13)$$

Note once we know $A^{(k)} \rightarrow D$ and D is diagonal its automatic that the diagonal entries are the eigenvalues of A .

$$A^{(k+1)} = R^T A^{(k)} R = \underbrace{R^{-1} A^{(k)} R}_{\text{similarity}} \quad \text{doesn't change the eigenvalues.}$$

The eigenvalues are the roots of $\chi_A(\lambda) = \det(A - \lambda I)$

$$B = R^{-1} A R$$

$$\begin{aligned} \chi_B(\lambda) &= \det(B - \lambda I) = \det(R^{-1} A R - \lambda I) \\ &= \det(R^{-1} A R - \lambda R^{-1} R) \quad \text{now factor inside} \\ &= \det((R^{-1} A - \lambda R^{-1}) R) = \det \underbrace{R^{-1}}_A \underbrace{(A - \lambda I)}_B R \\ &= \det((A - \lambda I) R R^{-1}) = \det(A - \lambda I) = \chi_A(\lambda) \end{aligned}$$

It follows that the eigenvalues of $A^{(k+1)}$ are the same as $A^{(k)}$ and so forth.. so in the limit the eigenvalues of D are the same as A.

$$\text{We know } \|A^{(k+1)}\|_F^2 = \|A^{(k)}\|_F^2$$

Let $B = R^T A R$ where $A^T = A$ and $R^T = R^{-1}$
and $R = R^{(pq)}(\varphi_k)$ (plane rotation)

$$\|B\|_F^2 = \sum_{i,j=1}^n |B_{ij}|^2 = \sum_{i=1}^n |B_{ii}|^2 + \sum_{i \neq j} |B_{ij}|^2$$

||

$$\|A\|_F^2 = \sum_{i,j=1}^n |A_{ij}|^2 = \sum_{i=1}^n |A_{ii}|^2 + \sum_{i \neq j} |A_{ij}|^2$$

Thus

$$\underbrace{\sum_{i=1}^n |B_{ii}|^2}_{D(B)} + \underbrace{\sum_{i \neq j} |B_{ij}|^2}_{L(B)} = \underbrace{\sum_{i=1}^n |A_{ii}|^2}_{D(A)} + \underbrace{\sum_{i \neq j} |A_{ij}|^2}_{L(A)}$$

$$B[p,p] = A[p,p]*c^2 - 2*A[p,q]*s*c + A[q,q]*s^2$$

$$B[q,q] = A[p,p]*s^2 + 2*A[p,q]*s*c + A[q,q]*c^2$$

Thinking about the submatrices

$$\begin{bmatrix} b_{pp} & b_{pq} \\ b_{qp} & b_{qq} \end{bmatrix} = R^T \begin{bmatrix} a_{pp} & a_{pq} \\ a_{qp} & a_{qq} \end{bmatrix} R$$

where

$$R = \begin{bmatrix} \cos \varphi & \sin \varphi \\ -\sin \varphi & \cos \varphi \end{bmatrix}.$$

again

$$\left\| \begin{bmatrix} b_{pp} & b_{pq} \\ b_{qp} & b_{qq} \end{bmatrix} \right\|_F = \left\| \begin{bmatrix} a_{pp} & a_{pq} \\ a_{qp} & a_{qq} \end{bmatrix} \right\|_F$$

there are zero

Therefore

$$a_{pp}^2 + a_{qq}^2 + 2a_{qp} = b_{pp}^2 + b_{qq}^2 + 2b_{qp}$$

finish next time ...