

# 1 Jacobi Method for Solving Eigenvalues

## 1.1 Intuition

Imagine we have a simple diagonal matrix C

$$\begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix}$$

Finding the eigenvalue and eigenvector is trivial. Its eigenvalue is  $\lambda_1 = 1$ ,  $\lambda_2 = -1$ . The eigenvectors are

$$v_1 = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$$

and

$$v_2 = \begin{pmatrix} 0 \\ 1 \end{pmatrix}$$

We now consider a rotation matrix that rotates the eigenvectors by 45 degree angle

$$\begin{aligned} u_1 &= Rv_1 \\ &= \begin{pmatrix} \frac{\sqrt{2}}{2} & -\frac{\sqrt{2}}{2} \\ \frac{\sqrt{2}}{2} & \frac{\sqrt{2}}{2} \end{pmatrix} \begin{pmatrix} 1 \\ 0 \end{pmatrix} = \begin{pmatrix} \frac{\sqrt{2}}{2} \\ \frac{\sqrt{2}}{2} \end{pmatrix} \end{aligned}$$

The eigenequation still holds for  $u_1$

$$\begin{aligned} CR^{-1}Rv_1 &= \lambda v_1 \\ RCR^{-1}Rv_1 &= \lambda Rv_1 \\ RCR^{-1}u_1 &= \lambda u_1 \end{aligned}$$

Let  $RCR^{-1} = A$ , where A is

$$\begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix}$$

Up to now we see we can transform a diagonal matrix to non-diagonal, still symmetric matrix by doing rotation.

Jacobi Method

The Jacobi method reverse the idea above by rotating a non-diagonal matrix back to a diagonal matrix.

$$\begin{aligned} A &= RCR^{-1} \\ C &= R^{-1}AR \end{aligned}$$

where matrix C is diagonal.

## 1.2 Eligibility

Since we apply similar transformation by rotation matrix and eventually we can the diagonal matrix which is symmetric, the original matrix has to be symmetrical.

## 1.3 Algorithm

The Jacobi iteration for a matrix A is

$$A^{(k)} = R_{pkqk}^T(\theta_k) A^{(k-1)} R_{pkqk}(\theta_k)$$

Where

$$G_{pq}(\theta) = \begin{pmatrix} I & 0 & 0 & 0 & 0 \\ 0 & \cos(\theta) & 0 & \sin(\theta) & 0 \\ 0 & 0 & I & 0 & 0 \\ 0 & -\sin(\theta) & 0 & \cos(\theta) & 0 \\ 0 & 0 & 0 & 0 & I \end{pmatrix}$$

It is an Identity matrix replaced by an rotation matrix on pth and qth columns and rows. The iteration is chosen to reduce the sum of the squares of the off-diagonal elements, which for any square matrix A is

$$\|A\|_F^2 - \sum_i a_{ii}^2$$

The orthogonal similarity transforms preserve the Frobenius norm

$$\|A^{(k)}\|_F = \|A^{(k-1)}\|_F$$

Because the rotation matrix change only (p,p), (q,q), (p,q), (q,p) positions. We have

$$(a_{pp}^{(k)})^2 + (a_{qq}^{(k)})^2 + 2(a_{pq}^{(k)})^2 = (a_{pp}^{(k-1)})^2 + (a_{qq}^{(k-1)})^2 + 2(a_{pq}^{(k-1)})^2$$

The off-diagonal sum of squares at the kth stage in terms of that at k-1 th stage is

$$\begin{aligned} & \|A^{(k)}\|_F^2 - \sum_i (a_{ii}^{(k)})^2 \\ &= \|A^{(k)}\|_F^2 - \sum_{i \neq p,q} (a_{ii}^{(k)})^2 - ((a_{pp}^{(k)})^2 + (a_{qq}^{(k)})^2) \\ &= \|A^{(k)}\|_F^2 - \sum_i (a_{ii}^{(k-1)})^2 - 2(a_{pq}^{(k-1)})^2 + 2(a_{pq}^{(k)})^2 \end{aligned}$$

In order to minimize this, we need

$$\begin{aligned} a_{pq}^{(k)} &= 0 \\ a_{pq}^{(k-1)} &= \max_{i < j} |a_{ij}^{(k-1)}| \end{aligned}$$

This implies

$$a_{pq}^{(k-1)}(\cos^2\theta - \sin^2\theta) + (a_{pp}^{k-1} - a_{qq}^{k-1})\cos\theta\sin\theta = 0$$

Solve for  $\theta$

$$\begin{aligned} \tan(2\theta) &= \frac{2a_{pq}^{(k-1)}}{a_{pp}^{k-1} - a_{qq}^{k-1}} \\ \tan(\theta) &= \frac{\tan(2\theta)}{1 + \sqrt{1 + \tan^2(2\theta)}} \\ \cos\theta &= \frac{1}{\sqrt{1 + \tan^2\theta}} \\ \sin\theta &= \cos\theta\tan\theta \end{aligned}$$

We use the above formula to update the matrix(by rotation).

## 1.4 Eigenvectors

Since

$$A = RCR^{-1}C = R^{-1}AR$$

We know the eigenvectors for diagonal matrix is unit vector  $e$ , so

$$Ce = R^{-1}ARe = \lambda e$$

Multiply by  $R$  from the left,

$$RPe = \lambda Re$$

so  $Re$  is the eigenvector for  $A$ .