Parallelization of QR decomposition

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Contents

- QR Decomposition?
 - What is a QR decomposition?
 - Basic facts
 - Givens rotation
 - Householder reflection
- 2 Eigenvalues
 - How to compute it
- Implementation
 - Example in Matlab
 - Solution

What is a QR decomposition?

It is

- \bullet $A = Q \cdot R$
- A is SPD (symmetric positive definite)
- Q is orthogonal
- R is upper triangular
- it is one way to compute eigen numbers

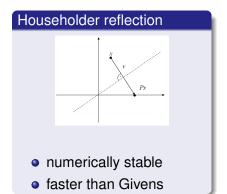
It is not

- the best way to compute solution of Ax = B
- the Gaussian elimination method is better

Basic facts

We make matrix A upper triangular by

Givens rotation $G(\alpha)v$ numerically unstable requires one more cycle



Givens rotation

$$\left(\begin{smallmatrix} \star & \star & \star \\ \star & \star & \star \\ \star & \star & \star \end{smallmatrix}\right) \cdot G_1 \to \left(\begin{smallmatrix} \star & \star & \star \\ \star & \star & \star \\ 0 & \star & \star \end{smallmatrix}\right) \cdot G_2 \to \left(\begin{smallmatrix} \star & \star & \star \\ 0 & \star & \star \\ 0 & \star & \star \end{smallmatrix}\right) \cdot G_3 \to \left(\begin{smallmatrix} \star & \star & \star \\ 0 & \star & \star \\ 0 & 0 & \star \end{smallmatrix}\right)$$

$$G(i,k,\alpha) = \begin{pmatrix} 1 & \cdots & 0 & \cdots & 0 & \cdots & 0 \\ \vdots & \ddots & \vdots & & \vdots & & \vdots & \vdots \\ 0 & \cdots & c & \cdots & s & \cdots & 0 \\ \vdots & & \vdots & \ddots & \vdots & & \vdots \\ 0 & \cdots & -s & \cdots & c & \cdots & 0 \\ \vdots & & \vdots & & \vdots & \ddots & \vdots \\ 0 & \cdots & 0 & \cdots & 0 & \cdots & 1 \end{pmatrix} k$$

$$A \cdot (G_1 G_2 G_3) = R \rightarrow Q = G_3^T G_2^T G_1^T$$

Householder reflection

$$\left(\begin{array}{ccc} \star & \star & \star \\ \star & \star & \star \\ \star & \star & \star \end{array}\right) \cdot P_1 \rightarrow \left(\begin{array}{ccc} \star & \star & \star \\ 0 & \star & \star \\ 0 & \star & \star \end{array}\right) \cdot P_2 \rightarrow \left(\begin{array}{ccc} \star & \star & \star \\ 0 & \star & \star \\ 0 & 0 & \star \end{array}\right)$$

$$P_{i} = \begin{pmatrix} 1 & 0 & \cdots & \cdots & 0 \\ 0 & \ddots & \cdots & \ddots & \vdots \\ \vdots & \cdots & \star & \cdots & \star \\ \vdots & \cdots & \vdots & \ddots & \vdots \\ 0 & \cdots & \star & \cdots & \star \end{pmatrix}$$

$$A \cdot (P_1 P_2) = Q \rightarrow R = P_2^T P_1^T$$

Which one should we choose?

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Householder

How to use the QR decomposition to find eigenvalues?

It's simple. We compute matrices Q, R and update matrix A

$$A_{new} = Q' \cdot A \cdot Q$$
.

We repeat QR decomposition with new matrix A. After 6 iterations for matrices less than 100 (approximately 10 iterations for matrices until 1000) the matrix A is like diagonal.

The items on diagonal of A are approximations of eigenvalues.

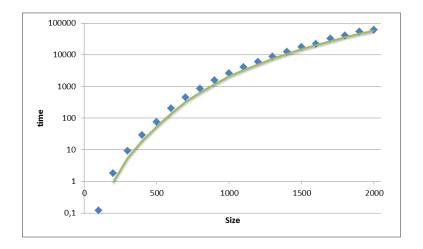
Example in matlab

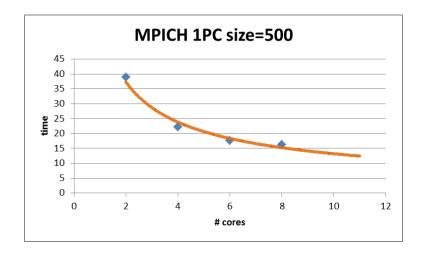
Givens rotation

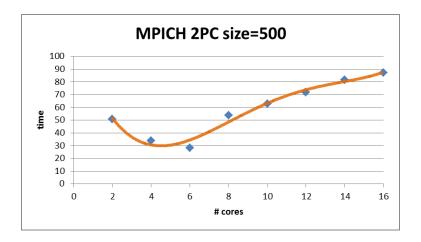
Hauseholder reflection

```
[m,n] = size(A);
Q = eye(m,m); R = A;
for j=1:min(m,n)
    x = R(j:m,j);
    v=sign(x(1))*norm(x)*eye(m-j+1,1)-x;
    if norm(v) > 0
        v=v/norm(v);
        P=eye(m);
        P(j:m,j:m)=P(j:m,j:m)-2*v*v';
        R=P*R;
        Q=Q*P;
end
end
```

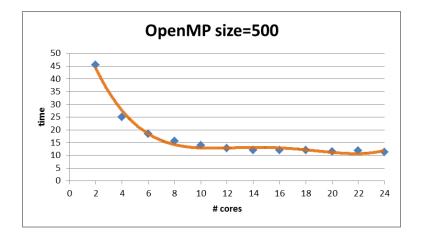
Sequential approach

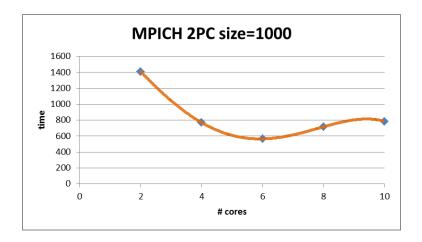




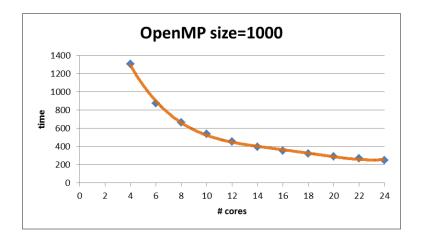


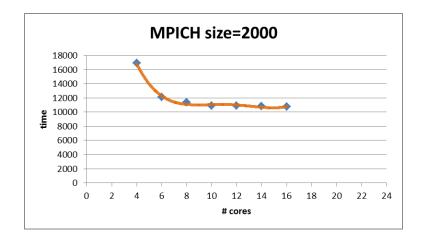
OpenMP



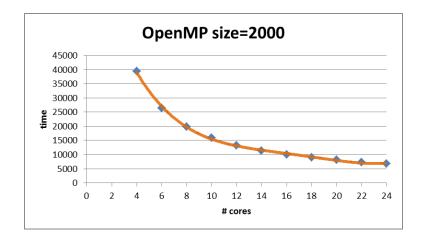


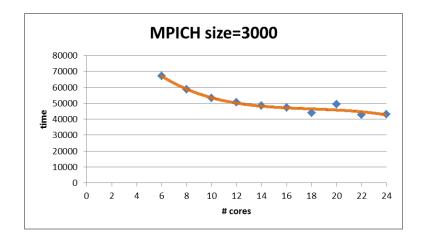
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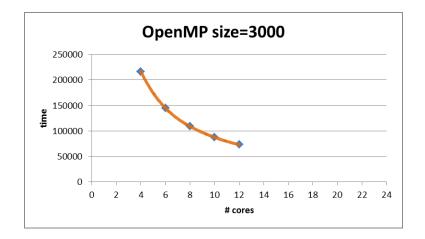


OpenMP





OpenMP



Thanks

Thank you!