

A deep-dive into the QR algorithm

- Basic form of QR:

$$A_{m-1} = Q_m \cdot R_m$$

$$A_m = R_m \cdot Q_m$$

1. Why can we expect convergence at all?

Denote the eigenvectors by v_j and associated eigenvalues as λ_j .

If S is a 1-dim subspace then the sequence

$$S, AS, A^2S, \dots$$

converges to the eigenspace $T = \text{span}\{v_1\}$ if $|\lambda_1| > |\lambda_2| \geq |\lambda_3| \dots$

Correspondingly, if S consists of k vectors, then the sequence

$$A, AS, A^2S, \dots$$

converges to the space $T = \text{span}\{v_1, \dots, v_k\}$ if $|\lambda_k| > |\lambda_{k+1}|$.

Now let $Q^{(m)}$ be an orthogonal basis of A^m s.

We have

$$[Q^{(m)}]^T A Q^{(m)} = \begin{bmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{bmatrix}$$

$$\text{Let } Q^{(m)} = [q_0^{(m)} \mid q_1^{(m)} \mid \dots \mid q_k^{(m)}].$$

$$\text{Compute } Z^{(m+1)} = A \cdot Q^{(m)}$$

and let $Q^{(m+1)}, R = Z^{(m+1)}$ be its QR decomposition.

It follows that also the columns $q_0^{(m)}, \dots, q_j^{(m)}$ converge for each $j \leq k$ against an invariant subspace.

Hence, B_{11} converges to an upper triangular matrix.

For $k = n$ therefore

$$[Q^{(m)}]^T A Q^{(m)} \rightarrow R$$

upper triangular

The QR Iteration

Let $A^{(0)} = A$

That's it!!

$$\left\{ \begin{array}{l} A^{(m)} = Q^{(m)} R^{(m)} \\ A^{(m+1)} = R^{(m)} Q^{(m)} \end{array} \right.$$

Why does this work?

Have
$$A^{(m+1)} = \underbrace{\left[Q^{(m)} \right]^H}_{\text{}} A^{(m)} Q^{(m)}$$

Looks like the
testing from
simultaneous iteration.

With a bit of algebra
can show that with $A^{(0)} = \underline{I}$
simultaneous iteration
produces the same
sequence $A^{(m)}$.

Speeding up $\odot \mathbb{R}$.

1.) Transform A to Hessemberg form.

A matrix H is upper
Hessenberg if it is
upper triangular + first
subdiagonal is nonzero

A hand-drawn diagram on a grid background. It features a large rectangle with a vertical line on the left side and a horizontal line at the top. Inside the rectangle, there are several diagonal lines sloping downwards from left to right. To the left of the rectangle, there is a vertical line and a horizontal line, possibly representing a coordinate system or a specific point.

We can construct \rightarrow
H explicitly through
eigenvalue preserving
transformations

$$H = Q^T A Q$$

is $O(n^3)$ ops.

The QR iteration

$$\text{Step } H^{(m)} = Q^{(m)} R^{(m)}$$

$$H^{(m+1)} = R^{(m)} Q^{(m)}$$

preserves Hessenberg form and can be done in $O(n^2)$ ops.

2.) Use shifts

$$A^{(m)} - \sigma I = Q^{(m)} R^{(m)}$$

$$A^{(m+1)} = R^{(m)} Q^{(m)} + \sigma I.$$

How to choose σ ?

Can show that for $\sigma = A^{(m)}[n, n]$ [bottom right elem of $A^{(m)}$] the QR iteration acts like a Rayleigh-quotient iteration on the last columns of the Q factors [the exact result is a bit more involved]

\Rightarrow Expect fast convergence in the last column.

$$A^{(m)} = \begin{bmatrix} a_{11}^{(m)} & & & \\ a_{21}^{(m)} & \ddots & & \\ & \ddots & \ddots & \\ 0 & & a_{n-1,n-1}^{(m)} & a_{n-1,n}^{(m)} \end{bmatrix}$$

Upon convergence

$$a_{n,n-1}^{(m)} \rightarrow 0.$$

If sufficiently small,
reduce the matrix
and only continue
with the first $n-1$
rows/cols.