## EE24BTECH11003 - Akshara Sarma Chennubhatla

**Question:** A train travels 360km at a uniform speed. If the speed had been 5km/h more, it would have taken 1 hour less for the same journey. Find the speed of the train.

#### Solution:

#### **Theoretical Solution:**

Let s be the speed of the train, then,

$$\frac{360}{s} - 1 = \frac{360}{s + 5} \tag{1}$$

$$\implies s^2 + 5s = 1800 \tag{2}$$

This is a quadratic equation whose roots are the possible values of the speed. Using the quadratic formula,

$$s = \frac{-5 \pm \sqrt{5^2 - 4(1)(-1800)}}{2(1)} \tag{3}$$

$$s = -45, 40 \tag{4}$$

### **Simulated Solution:**

### By Newton-Ralphson method,

Take initial guess  $s_0$ , then run the following loop,

$$s_{n+1} = s_n - \frac{f(s_n)}{f'(s_n)} \tag{5}$$

$$f(s) = s^2 + 5s - 1800 (6)$$

$$f'(s) = 2s + 5 \tag{7}$$

$$s_{n+1} = s_n - \frac{s_n^2 + 5s_n - 1800}{2s_n + 5} \tag{8}$$

This method converges for real roots but when roots are complex, it can go to infinity as well. To avoid that, if our roots are complex, take initial guess as a complex number. The values of s got through this method are,

$$s = -45 \tag{9}$$

$$s = 40 \tag{10}$$

Alternatively, we can solve the question by using the eigen values of the companion matrix.

1

For a polynomial equation of form  $x_n + c_{n-1}x^{n-1} + \cdots + c_2x^2 + c_1x + c_0 = 0$  the companion matrix if of the form

$$\begin{pmatrix}
0 & 0 & \cdots & 0 & -c_0 \\
1 & 0 & \cdots & 0 & -c_1 \\
0 & 1 & \cdots & 0 & -c_2 \\
\vdots & \vdots & \ddots & \vdots & \vdots \\
0 & 0 & \cdots & 1 & -c_{n-1}
\end{pmatrix}$$
(11)

The eigen values of this matrix are the roots of the polynomial equation. For this question,

$$n = 2 \tag{12}$$

$$c_0 = -1800 \tag{13}$$

$$c_1 = 5 \tag{14}$$

$$C = \begin{pmatrix} 0 & 1800 \\ 1 & -5 \end{pmatrix} \tag{15}$$

To find the eigen values of the matrix, we use the method of QR decomposition of the matrix.

The QR algorithm decomposes a matrix A into the product of an orthogonal matrix Q and an upper triangular matrix R, such that

$$A = QR \tag{16}$$

The matrix is then updated iteratively as:

$$A_{new} = RQ \tag{17}$$

This process is repeated until A converges to an upper triangular form. Steps to perform QR decomposition with accelerated convergence,

- 1) Convert to Upper Hessenberg form via Householder Reflections
- 2) Performing QR decomposition via Givens Rotations with shifts
- 3) Read off diagonal elements

#### **Householder Reflections:**

A square matrix A of order  $n \times n$  is said to be in upper Hessenberg form if all the entries below the first subdiagonal are zero. For example:

$$H = \begin{bmatrix} h_{11} & h_{12} & h_{13} & \cdots & h_{1n} \\ h_{21} & h_{22} & h_{23} & \cdots & h_{2n} \\ 0 & h_{32} & h_{33} & \cdots & h_{3n} \\ \vdots & \ddots & \ddots & \ddots & \vdots \\ 0 & \cdots & 0 & h_{n-1,n-1} & h_{n-1,n} \\ 0 & \cdots & \cdots & 0 & h_{nn} \end{bmatrix}$$

$$(18)$$

1) Select a Subvector x:

$$x = \begin{bmatrix} A_{2,1} \\ A_{3,1} \\ \vdots \\ A_{n,1} \end{bmatrix}. \tag{19}$$

2) Define the Target Vector: The goal is to transform  $\mathbf{x}$  into a new vector  $\mathbf{y}$  where only the first element is non-zero, and all the other elements are zero. First, compute  $||\mathbf{x}||$ :

$$\mathbf{y} = \pm \|\mathbf{x}\| \, e_1,\tag{20}$$

3) Construct the Householder Vector **v**: To generate a reflection that transforms *x* to *y*, the Householder vector *v* is defined as:

$$v = x - \text{sign}(x_1) ||x|| e_1$$
 (21)

$$sign(x_1) = \frac{x_1}{|x_1|},$$
(22)

After defining v, it is normalized to a unit vector:

4) Construct the Householder Matrix  $H_k$ : The Householder matrix  $H_k$  is constructed as:

$$H_k = I - 2\frac{vv^*}{v^*v},\tag{23}$$

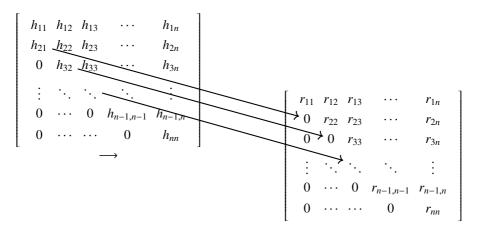
5) Apply the Householder Transformation: The matrix  $H_k$  is applied to A as:

$$A' = H_k A H_k^*, (24)$$

This will reduce the matrix to Hessenberg form by eliminating the sub-diagonal elements of the first column.

6) Repeat for Subsequent Columns: This Householder transformation approach ensures that the matrix is gradually transformed to a Hessenberg form, where all elements below the first sub-diagonal are zero.

### **Givens Rotations:**



Each Givens rotation zeros out a specific subdiagonal element, progressively transforming the Hessenberg matrix into an upper triangular matrix.

To choose the values of c and s for the Givens rotation in QR decomposition, let  $a_j$  be the element we wish to null out (i.e. make 0). Pick an arbitrary non-zero pivot element  $a_i$  (on a different row). Usually, if we wish to null a particular sub-diagonal element, we pick the principal diagonal element above it as a pivot.

$$c = \frac{\overline{a_i}}{\sqrt{a_i^2 + a_j^2}}, \quad s = \frac{-\overline{a_j}}{\sqrt{a_i^2 + a_j^2}}$$
 (25)

Givens rotation essentially rotates the two rows that  $a_i$  and  $a_j$  are on such that  $a_j = 0$  after rotation, other rows remain unaffected.

Visualizing the process,

$$\begin{bmatrix}
\times & \times & \times & \times \\
\times & \times & \times & \times \\
0 & \times & \times & \times \\
0 & 0 & \times & \times
\end{bmatrix}
\xrightarrow{G(3,2,\theta_1)}
\begin{bmatrix}
\times & \times & \times & \times \\
\times & \times & \times & \times \\
0 & 0 & \times & \times \\
0 & 0 & \times & \times
\end{bmatrix}
\xrightarrow{G(4,3,\theta_2)}
\begin{bmatrix}
\times & \times & \times & \times \\
\times & \times & \times & \times \\
0 & 0 & \times & \times \\
0 & 0 & 0 & \times
\end{bmatrix}.$$
(26)

After all Givens rotations, the resulting matrix is upper triangular:

$$R = \begin{bmatrix} \times & \times & \times & \times \\ 0 & \times & \times & \times \\ 0 & 0 & \times & \times \\ 0 & 0 & 0 & \times \end{bmatrix}. \tag{27}$$

The sequence of Givens rotations  $G_1, G_2, \ldots, G_m$  satisfies:

$$G_m \cdots G_2 G_1 A = R, \tag{28}$$

where R is upper triangular. The QR decomposition is obtained by combining the

transposes of the Givens rotations into Q:

$$A = QR, \quad Q = G_1^{\mathsf{T}} G_2^{\mathsf{T}} \cdots G_m^{\mathsf{T}}. \tag{29}$$

$$A_{k+1} = R_k Q_k \tag{30}$$

$$= (G_n \dots G_2 G_1) A_k (G_1^\top G_2^\top \dots G_n^\top)$$
(31)

$$= (G_n \dots G_2 G_1) A_k (G_n \dots G_2 G_1)^{\top}$$
 (32)

Iteratively repeating this process causes the matrix to converge to upper triangular.

# **Handling Jordan Blocks:**

Jordan blocks pose challenges in eigenvalue computation because the matrix cannot be diagonalized. A Jordan block for eigenvalue  $\lambda$  appears as:

$$\begin{bmatrix} a & b \\ c & d \end{bmatrix} \tag{33}$$

where a and b are the diagonal elements and c is a non zero sub-diagonal element. To handle Jordan blocks, the QR algorithm implemented here solves for the eigenvalues directly using the characteristic polynomial of the block. For a  $2 \times 2$  Jordan block, the eigenvalues are roots of:

$$\lambda^2 - (\text{trace})\lambda + \det = 0. \tag{34}$$

In this case, the eigen values of the matrix computed are,

$$\lambda_1 = -45 \tag{35}$$

$$\lambda_2 = 40 \tag{36}$$

(37)

Below is the plot for given quadratic equation, obtained by iterating through the values of x with step size of h

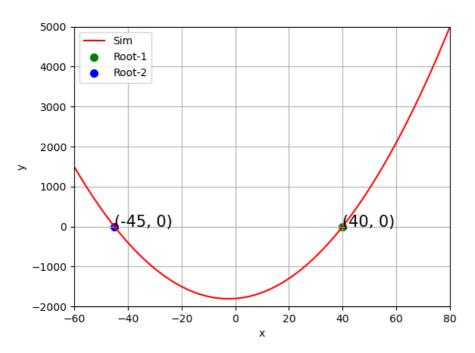


Fig. 1: Plot of  $s^2 + 5s = 1800$