

On A-eccentric Graphs

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Abstract

Motivated by the concept called unique eccentric point (u.e.p) [14], Kishori et al. in [2] generalized the concept as k -eccentric point graph. A graph tt is called an *unique eccentric point (u.e.p) graph* if each point of tt has a unique eccentric point where as in *k -eccentric point graph* every vertex has exactly k -eccentric vertices. Here we are denoting k -eccentric graph as A -eccentric graph and studied its property with peripheral path matrix.

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1 Introduction

Let tt be a connected, nontrivial graph with vertex set $V(tt)$ and edge set $E(tt)$ and let $|V(tt)| = n$ and $|E(tt)| = m$. Let u and v be two vertices of a graph tt . The distance $d(u, v|tt)$ between the vertices u and v is the length of a shortest path connecting u and v . If $u = v$ then $d(u, v|tt) = 0$. The eccentricity $e(v)$ of a vertex v in a graph tt is the distance between v and a vertex farthest from v in tt . The diameter $diam(tt)$ of tt is the maximum eccentricity of tt , while the radius $rad(tt)$ is the smallest eccentricity of tt . A vertex v with $e(v) = diam(tt)$ is called a peripheral vertex of tt . A set of peripheral vertices of tt is called as periphery and is denoted as $P(tt)$. The peripheral path matrix $M_p(tt) = [p_{ij}]$, whose entries are 1 if there is a peripheral path between v_i and v_j in tt and 0 otherwise.

The peripheral path energy (p -energy (in short)) of a graph tt is defined as the sum of the absolute values of p -eigenvalues of the $M_p(tt)$. i.e,

$$E_p = E_p(tt) = \sum_{i=1}^n |\alpha_i| \quad (1)$$

The form of Eq. (1) is chosen so as to be fully analogous to the definition of graph energy [8, 5, 7].

$$E = E(tt) = \sum_{i=1}^n |\lambda_i|$$

where, $\lambda_1, \lambda_2, \dots, \lambda_n$ are the ordinary eigenvalues [9], i.e the eigenvalues of the adjacency matrix $A(tt)$. Observe that the graph energy $E(tt)$ in a past few years has been extensively studied and surveyed in Mathematics and Chemistry [12, 13, 15, 17, 18, 19, 21, 25, 26, 10, 11, 16, 20, 23]. Through out the paper $|P(tt)| = k$ with labellings v_1, v_2, \dots, v_k , where $2 \leq k \leq n$.

The characteristic polynomial of $M_p(tt)$ is the $\det(\alpha I - M_p(tt))$. It is referred to as a characteristic polynomial of tt and is denoted by $\psi(tt; \alpha) = c_0\alpha^n + c_1\alpha^{n-1} + c_2\alpha^{n-2} + \dots + c_n$. The roots $\alpha_1, \alpha_2, \dots, \alpha_n$ of the polynomial $\psi(tt; \alpha)$ is called the eigenvalues of $M_p(tt)$.

The eigenvalues of $M_p(tt)$ are said to be the peripheral path eigenvalues (or p -eigenvalues (in short)) of tt . Since $M_p(tt)$ is a real, symmetric matrix, the p -eigenvalues are real and can be ordered in non-increasing order, $\alpha_1 \geq \alpha_2 \geq \dots \geq \alpha_n$. Then the p -spectrum of a graph tt is the set of p -eigenvalues of $M_p(tt)$, together with the multiplicities as p -eigenvalues of $M_p(tt)$. If the p -eigenvalues of $M_p(tt)$ are $\alpha_1 \geq \alpha_2 \geq \dots \geq \alpha_n$ and their multiplicities are $m(\alpha_1), m(\alpha_2), \dots, m(\alpha_n)$, then the result will be

$$p\text{-Spec}(tt) = \begin{matrix} \alpha_1 & \alpha_2 & \dots & \alpha_n \\ m(\alpha_1) & m(\alpha_2) & \dots & m(\alpha_n) \end{matrix} \sum$$

For Example, let tt be a graph as shown below Figure 1.

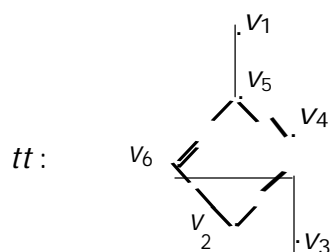


Figure 1: tt is a graph of order $n = 6$ with $k = 3$ peripheral vertices.

$$M_p(tt) = \begin{matrix} & \begin{matrix} v_1 & v_2 & v_3 & v_4 & v_5 & v_6 \end{matrix} \\ \begin{matrix} v_1 \\ v_2 \\ v_3 \\ v_4 \\ v_5 \\ v_6 \end{matrix} & \begin{bmatrix} 0 & 1 & 1 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix} \end{matrix} \quad \begin{matrix} \square \\ \square \\ \square \\ \square \\ \square \\ \square \end{matrix}$$

6x6

Let $\varepsilon_v = \{u : d(u, v) = \text{diam}(tt)\}$. Clearly, $|\varepsilon_{v_1}| = 2$ and $|\varepsilon_{v_2}| = |\varepsilon_{v_3}| = 1$ and characteristic polynomial of tt is $\psi(tt, \alpha) = -2\alpha^4 + \alpha^6$

whose, p - eigenvalues are $-2, 2, 0, 0, 0, 0$. And hence, p -energy of tt is 2.8284.

Motivated by the concept called unique eccentric point (u.e.p) [14], Kishori et al. in [2] generalized the concept as k -eccentric point graph. A graph tt is called an *unique eccentric point (u.e.p) graph* if each point of tt has a unique eccentric point where as in *k-eccentric point graph* every vertex has exactly k -eccentric vertices. Here we are denoting k -eccentric graph as A -eccentric graph and studied its property with peripheral path matrix. For more details about the Peripheral path matrix, Peripheral path energy, Peripheral distance energy, Peripheral path equi-energy and Peripheral Wiener index one can refer [22],

Definition 1.1. A vertex u is said to be an eccentric vertex of v if $e(v) = d(u, v)$ $u, v \in V(tt)$. A graph tt is A -eccentric graph if for every x of $V(tt)$, there are A -eccentric vertices.

For Example: Refer the Figure 2.

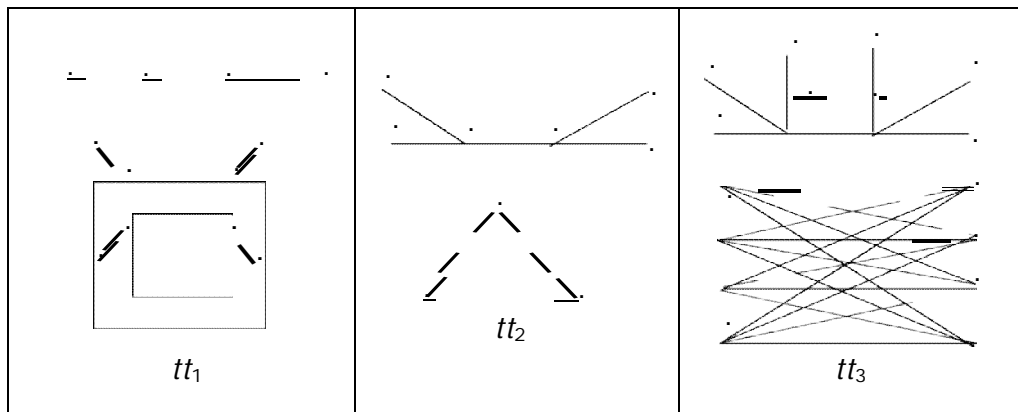


Figure 2: tt_1 , tt_2 , and tt_3 are the 1, 2 and 3 eccentric graphs respectively

2 Properties of A-eccentric graph

Theorem 2.1. Let tt be A-eccentric graph. Then,

1. A is an p -eigenvalues of tt .
2. If tt is connected, then the multiplicity of A is 1.
3. For any p -eigenvalue α of tt , we have $|\alpha| \leq A$

Proof.

1. Let $u = [1, 1, 1, \dots, 1]^t$, then if M_p is peripheral path matrix of tt , we have,

$$M_p = Au. \quad (2)$$

This is true because there are A 1's in each row. Thus A is a p -eigenvalue of tt .

2. Let $X = [x_1, x_2, \dots, x_n]^t$ denote any non zero vector for which

$$AX = AX \quad (3)$$

suppose that x_j is an entry of X with the largest absolute value, then the Eq (3) can be expressed as

$$(Ax)_j = Ax_j$$

and hence

$$\sum x_i = Ax_j$$

where summation runs over those A vertices v_i which are an eccentric vertex to v_j . If tt is connected, we may proceed successively in this way, eventually showing that all entries of X are equal. Thus X is a multiple of u and the space of eigenvector associated with the p -eigenvalue A has dimension 1.

3. Suppose

$$\text{that } Ay = \alpha y, \quad y \neq 0 \quad (4)$$

and y_j denote an entry of y which is least in absolute value, as in condition 2. we have,

$$\begin{aligned} \sum y_i &= \alpha y_j \text{ and} \\ |\alpha| |y_j| &= \left| \sum y_i \right| \\ |\alpha| |y_j| &\leq \sum |y_i| \\ |\alpha| |y_j| &= A |y_j| \\ \Rightarrow |\alpha| &\leq A \text{ as required.} \end{aligned}$$

□

Proposition 2.2. *If tt is A -eccentric graph and α is a p -eigenvalue of $M_p(tt)$, then no p -eigenvalue of $M_p(tt)$ has absolute value greater than A .*

Proof. Let α be a p -eigenvalue of $M_p(tt)$ and $x = [x_1, x_2, \dots, x_k]^t$ be corresponding eigenvector. Let x_i be the entry of x whose absolute value is greater. Hence,

$$\alpha x_i = \sum_{j=1}^k a_{ij} x_j; \quad i = 1, 2, \dots, k.$$

Now

$$|\alpha| |x_i| \leq |x_i| \sum_{j=1}^k a_{ij}; \quad i = 1, 2, \dots, k.$$

since tt is A -eccentric graph,

$$\sum_{j=1}^k a_{ij} = A; \quad i = 1, 2, \dots, k.$$

Hence,

$$|\alpha| |x_i| \leq |x_i| \cdot A \Rightarrow |\alpha| \leq A.$$

□

Corollary 2.3. Suppose tt is unique eccentric point graph then no p -eigenvalue of $M_p(tt)$ has absolute value greater than 1.

Proof. Let α be a p -eigenvalue of $M_p(tt)$ and $x = [x_1, x_2, \dots, x_k]^t$ corresponding eigenvector. Let x_i be the entry of x whose absolute value is greatest. Hence,

$$\alpha x_i = \sum_{j=1}^k a_{ij} x_j ; i = 1, 2, \dots, k.$$

Now,

$$|\alpha| |x_i| \leq |x_i| \sum_{j=1}^k a_{ij} ; i = 1, 2, \dots, k.$$

since tt is unique eccentric point graph,

$$\Rightarrow |\alpha| \leq 1.$$

□

Proposition 2.4. Let $M_p(tt)$ be $n \times n$ real matrix with v_1, v_2, \dots, v_k peripheral vertices. Let $M_{p_1}(tt)$ be $k \times k$ real sub matrix of $M_p(tt)$. If tt is unique eccentric point graph then there is a non-zero column vector x such that $Ax = x$

Proof. Let $M_{p_1}(tt) - I$ be a matrix which has the properties that the sum of the entries in each column is equal to zero. Since tt is unique eccentric point graph, $M_{p_1}(tt) - I$ results in to a matrix whose first row is zero. (If not a small operation can be applied, i.e. $R_1 \rightarrow R_1 + R_2 + \dots + R_k$ to $M_{p_1}(tt) - I$). Hence, $|M_{p_1}(tt) - I| = 0$. Thus 1 is a p -eigenvalue of A and there is an eigenvector $x \neq 0$ such that $Ax = x$. □

3 A-Eccentric Trees:

For A -eccentric tree, the peripheral path matrix $M_p(T)$ is as follows:

$$M(T) = \begin{matrix} & \begin{matrix} v_1 & v_2 & \dots & v_{\ell} & v_{\ell+1} & v_{\ell+2} & \dots & v_{2\ell} & v_{2\ell+1} & v_{2\ell+2} & \dots & v_n \end{matrix} \\ \begin{matrix} v_1 \\ v_2 \\ \vdots \\ v_{\ell} \\ v_{\ell+1} \\ v_{\ell+2} \\ \vdots \\ v_{2\ell} \\ v_{2\ell+1} \\ v_{2\ell+2} \\ \vdots \\ v_n \end{matrix} & \begin{bmatrix} 0 & 0 & \dots & 0 & 1 & 1 & \dots & 1 & 0 & 0 & \dots & 0 \\ 0 & 0 & \dots & 0 & 1 & 1 & \dots & 1 & 0 & 0 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots & \vdots & \vdots & \ddots & \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 0 & 1 & 1 & \dots & 1 & 0 & 0 & \dots & 0 \\ 1 & 1 & \dots & 1 & 0 & 0 & \dots & 0 & 0 & 0 & \dots & 0 \\ 1 & 1 & \dots & 1 & 0 & 0 & \dots & 0 & 0 & 0 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots & \vdots & \vdots & \ddots & \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & 1 & \dots & 1 & 0 & 0 & \dots & 0 & 0 & 0 & \dots & 0 \\ 0 & 0 & \dots & 0 & 0 & 0 & \dots & 0 & 0 & 0 & \dots & 0 \\ 0 & 0 & \dots & 0 & 0 & 0 & \dots & 0 & 0 & 0 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots & \vdots & \vdots & \ddots & \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 0 & 0 & 0 & \dots & 0 & 0 & 0 & \dots & 0 \end{bmatrix} \end{matrix}$$

This matrix representation takes the form

$$M_p(T) = \begin{matrix} \Sigma & \Sigma \\ \hline \begin{matrix} A_{2A \times 2A} \\ B_{(n-2A) \times 2A}^t \end{matrix} & \begin{matrix} B_{2A \times (n-2A)} \\ C_{(n-2A) \times (n-2A)} \end{matrix} \\ \hline & n \times n \end{matrix}$$

Note that sub matrix B , B^t and C are zero - matrix, where as A is non-zero matrix. A sub matrix A is $2A \times 2A$ sub matrix of $M_p(T)$, whose entries are as follows:

$$A = \begin{bmatrix} 0 & 0 & \dots & 0 & 1 & 1 & \dots & 1 \\ 0 & 0 & \dots & 0 & 1 & 1 & \dots & 1 \\ \vdots & \vdots & \ddots & \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 0 & 1 & 1 & \dots & 1 \\ 1 & 1 & \dots & 1 & 0 & 0 & \dots & 0 \\ 1 & 1 & \dots & 1 & 0 & 0 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & 1 & \dots & 1 & 0 & 0 & \dots & 0 \end{bmatrix}$$

Symbolically A can be represented as

$$A = \begin{matrix} \Sigma & \Sigma \\ \hline \begin{matrix} 0_{A \times A} \\ J_{A \times A}^t \end{matrix} & \begin{matrix} J_{A \times A} \\ 0_{A \times A} \end{matrix} \\ \hline & 2A \times 2A \end{matrix}$$

Where J is a $A \times A$ matrix having all entries 1.

Observation 3.1. A sub matrix A of $M_p(T)$ has just two linearly independent rows and so its rank is 2. Consequently 0 is an p -eigenvalue of $M_p(T)$ with multiplicity $2A - 2$.

Next, we give the Characteristic polynomial of a sub matrix A of $M_p(T)$. Suppose,

$$A = \begin{array}{c|cccc} \begin{array}{c} \square \\ \square \\ \square \\ \square \\ \square \\ \square \\ \square \\ \square \\ \square \\ \square \end{array} & \begin{array}{c} a_{11} \\ a_{21} \\ \vdots \\ a_{(A-1),1} \\ a_{A,1} \\ a_{(A+1),1} \\ a_{(A+2),1} \\ \vdots \\ a_{(2A-1),1} \\ a_{(2A),1} \end{array} & \cdots & \begin{array}{c} a_{1,(A-1)} \\ a_{2,(A-1)} \\ \vdots \\ a_{(A-1),(A-1)} \\ a_{A,(A-1)} \\ a_{(A+1),(A-1)} \\ a_{(A+2),(A-1)} \\ \vdots \\ a_{(2A-1),(A-1)} \\ a_{(2A),(A-1)} \end{array} & \begin{array}{c} a_{1,A} \\ a_{2,A} \\ \vdots \\ a_{(A-1),A} \\ a_{A,A} \\ a_{(A+1),A} \\ a_{(A+2),A} \\ \vdots \\ a_{(2A-1),A} \\ a_{(2A),A} \end{array} & \begin{array}{c} a_{1,(A+1)} \\ a_{2,(A+1)} \\ \vdots \\ a_{(A-1),(A+1)} \\ a_{A,(A+1)} \\ a_{(A+1),(A+1)} \\ a_{(A+2),(A+1)} \\ \vdots \\ a_{(2A-1),(A+1)} \\ a_{(2A),(A+1)} \end{array} & \cdots & \begin{array}{c} a_{1,2(A-1)} \\ a_{2,2(A-1)} \\ \vdots \\ a_{(A-1),2(A-1)} \\ a_{A,2(A-1)} \\ a_{(A+1),2(A-1)} \\ a_{(A+2),2(A-1)} \\ \vdots \\ a_{(2A-1),2(A-1)} \\ a_{(2A),2(A-1)} \end{array} & \begin{array}{c} a_{1,2A} \\ a_{2,2A} \\ \vdots \\ a_{(A-1),2A} \\ a_{A,2A} \\ a_{(A+1),2A} \\ a_{(A+2),2A} \\ \vdots \\ a_{(2A-1),2A} \\ a_{(2A),2A} \end{array} \\ \hline & & & & & & & \end{array} \quad \begin{array}{c} \square \\ \square \\ \square \\ \square \\ \square \\ \square \\ \square \\ \square \\ \square \\ \square \end{array}$$

$$\Rightarrow A = \begin{array}{c|cccc} \begin{array}{c} \square \\ \square \\ \square \\ \square \\ \square \\ \square \\ \square \\ \square \\ \square \\ \square \end{array} & \begin{array}{c} 0 \\ 0 \\ \vdots \\ 0 \\ 0 \\ 1 \\ 1 \\ \vdots \\ 1 \\ 1 \end{array} & \begin{array}{c} 0 \\ 0 \\ \vdots \\ 0 \\ 0 \\ 1 \\ 1 \\ \vdots \\ 0 \\ 0 \end{array} & \begin{array}{c} \cdots \\ \cdots \\ \vdots \\ \cdots \\ \cdots \\ \cdots \\ \cdots \\ \vdots \\ \cdots \\ \cdots \end{array} & \begin{array}{c} 0 \\ 0 \\ \vdots \\ 0 \\ 0 \\ 1 \\ 1 \\ \vdots \\ 0 \\ 0 \end{array} & \begin{array}{c} 1 \\ 1 \\ \vdots \\ 1 \\ 1 \\ 0 \\ 0 \\ \vdots \\ 0 \\ 0 \end{array} & \begin{array}{c} 1 \\ 1 \\ \vdots \\ 1 \\ 1 \\ 0 \\ 0 \\ \vdots \\ 0 \\ 0 \end{array} \\ \hline & & & & & & \end{array} \quad \begin{array}{c} \square \\ \square \\ \square \\ \square \\ \square \\ \square \\ \square \\ \square \\ \square \\ \square \end{array}$$

then,

$$|A - \alpha I| = \begin{array}{c|cccc} \begin{array}{c} \square \\ \square \\ \square \\ \square \\ \square \\ \square \\ \square \\ \square \\ \square \\ \square \end{array} & \begin{array}{c} -\alpha \\ 0 \\ \vdots \\ 0 \\ 0 \\ 1 \\ 1 \\ \vdots \\ 1 \\ 1 \end{array} & \begin{array}{c} 0 \\ 0 \\ \vdots \\ 0 \\ 0 \\ 1 \\ 1 \\ \vdots \\ 0 \\ 0 \end{array} & \begin{array}{c} \cdots \\ \cdots \\ \vdots \\ \cdots \\ \cdots \\ \cdots \\ \cdots \\ \vdots \\ \cdots \\ \cdots \end{array} & \begin{array}{c} 0 \\ 0 \\ \vdots \\ 0 \\ 0 \\ 1 \\ 1 \\ \vdots \\ 0 \\ 0 \end{array} & \begin{array}{c} 1 \\ 1 \\ \vdots \\ 1 \\ 1 \\ 0 \\ 0 \\ \vdots \\ 0 \\ 0 \end{array} & \begin{array}{c} 1 \\ 1 \\ \vdots \\ 1 \\ 1 \\ -\alpha \\ -\alpha \\ \vdots \\ -\alpha \\ -\alpha \end{array} \\ \hline & & & & & & \end{array} \quad \begin{array}{c} \square \\ \square \\ \square \\ \square \\ \square \\ \square \\ \square \\ \square \\ \square \\ \square \end{array}$$

now, subtract $(A + 1)^{th}$ row from $(A + 2)^{th}$, $(A + 3)^{th}$, ..., $(2A - 1)^{th}$, $(2A)^{th}$ rows, then

the result will be,

$$\begin{pmatrix}
 -\alpha & 0 & \dots & 0 & 0 & 1 & 1 & \dots & 1 & 1 \\
 0 & -\alpha & \dots & 0 & 0 & 1 & 1 & \dots & 1 & 1 \\
 \vdots & \vdots & \ddots & \vdots & \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\
 0 & 0 & \dots & -\alpha & 0 & 1 & 1 & \dots & 1 & 1 \\
 0 & 0 & \dots & 0 & -\alpha & 1 & 1 & \dots & 1 & 1 \\
 1 & 1 & \dots & 1 & 1 & -\alpha & 0 & \dots & 0 & 0 \\
 0 & 0 & \dots & 0 & 0 & \alpha & -\alpha & \dots & 0 & 0 \\
 \vdots & \vdots & \ddots & \vdots & \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\
 0 & 0 & \dots & 0 & 0 & \alpha & 0 & \dots & -\alpha & 0 \\
 0 & 0 & \dots & 0 & 0 & \alpha & 0 & \dots & 0 & -\alpha
 \end{pmatrix}_{2A \times 2A} = 0$$

Adding $(A + 2)^{th}$, $(A + 3)^{th}$, ..., $(2A - 1)^{th}$, $(2A)^{th}$ column to $(A + 1)^{th}$ column,

$$\begin{pmatrix}
 -\alpha & 0 & \dots & 0 & 0 & A & 1 & \dots & 1 & 1 \\
 0 & -\alpha & \dots & 0 & 0 & A & 1 & \dots & 1 & 1 \\
 \vdots & \vdots & \ddots & \vdots & \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\
 0 & 0 & \dots & -\alpha & 0 & A & 1 & \dots & 1 & 1 \\
 0 & 0 & \dots & 0 & -\alpha & A & 1 & \dots & 1 & 1 \\
 1 & 1 & \dots & 1 & 1 & -\alpha & 0 & \dots & 0 & 0 \\
 0 & 0 & \dots & 0 & 0 & 0 & -\alpha & \dots & 0 & 0 \\
 \vdots & \vdots & \ddots & \vdots & \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\
 0 & 0 & \dots & 0 & 0 & 0 & 0 & \dots & -\alpha & 0 \\
 0 & 0 & \dots & 0 & 0 & 0 & 0 & \dots & 0 & -\alpha
 \end{pmatrix}_{2A \times 2A} = 0$$

Again by subtracting $(1)^{st}$ column from $(2)^{nd}$, $(3)^{rd}$, ..., $(A)^{th}$ column, we have

$$\begin{pmatrix}
 -\alpha & \alpha & \dots & \alpha & \alpha & A & 1 & \dots & 1 & 1 \\
 0 & -\alpha & \dots & 0 & 0 & A & 1 & \dots & 1 & 1 \\
 \vdots & \vdots & \ddots & \vdots & \vdots & \vdots & \vdots & \ddots & \vdots & \vdots
 \end{pmatrix}$$

$2A \times 2A$

Expanding from 1st column.

(5)

$$\begin{aligned}
 \text{Let } A_1 = & \begin{pmatrix} -\alpha & \dots & 0 & 0 & A & 1 & \dots & 1 & 1 \\ \vdots & \ddots & \vdots & \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & \dots & -\alpha & -\alpha & A & 1 & \dots & 1 & 1 \\ 0 & \dots & 0 & 0 & -\alpha & 0 & \dots & 1 & 1 \\ 0 & \dots & 0 & 0 & 0 & - & \dots & 0 & 0 \\ \vdots & \ddots & & & & \alpha & \dots & 0 & 0 \\ \vdots & \ddots & \vdots & \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & \dots & 0 & 0 & 0 & 0 & \dots & -\alpha & 0 \\ 0 & \dots & 0 & 0 & 0 & 0 & \dots & 0 & -\alpha \end{pmatrix}_{(2A-1) \times (2A-1)} \\
 & \begin{pmatrix} \alpha & \dots & \alpha & \alpha & A & 1 & \dots & 1 & 1 \\ \vdots & \ddots & \vdots & \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ -\alpha & \dots & 0 & 0 & A & 1 & \dots & 1 & 1 \\ \vdots & \ddots & \vdots & \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & \dots & -\alpha & 0 & A & 1 & \dots & 1 & 1 \\ 0 & \dots & 0 & -\alpha & A & 1 & \dots & 1 & 1 \\ 0 & \dots & 0 & 0 & 0 & -\alpha & \dots & 0 & 0 \\ \vdots & \ddots & \vdots & \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & \dots & 0 & 0 & 0 & 0 & \dots & -\alpha & 0 \\ 0 & \dots & 0 & 0 & 0 & 0 & \dots & 0 & -\alpha \end{pmatrix}_{(2A-1) \times (2A-1)}
 \end{aligned}$$

now, the determinant of $A_1 = (-\alpha)^{2A-1}$, then to find the determinant of A_2 change $R_2 = R_1 + R_2, R_3 = R_3 + R_2, R_4 = R_4 + R_2, R_5 = R_5 + R_2, \dots, R_A = R_A + R_2$ then,

$$A_2 = \begin{pmatrix} \alpha & \dots & \alpha & \alpha & A & 1 & \dots & 1 & 1 \\ \vdots & \ddots & \vdots & \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & \dots & \alpha & \alpha & 2A & 2 & \dots & 2 & 2 \\ \vdots & \ddots & \vdots & \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & \dots & 0 & 0 & (A-1)A & 1 & \dots & 1 & 1 \\ 0 & \dots & 0 & 0 & A^2 & A & \dots & A & A \\ 0 & \dots & 0 & 0 & 0 & -\alpha & \dots & 0 & 0 \end{pmatrix}$$

$$\begin{pmatrix} \ddots & & & & & & & & \\ 0 & \dots & 0 & 0 & 0 & 0 & \dots & -\alpha & 0 \\ 0 & \dots & 0 & 0 & 0 & 0 & \dots & 0 & -\alpha \end{pmatrix}_{(2A-1) \times (2A-1)}$$

then $|A_2| = \alpha^{(A-1)} \cdot A^2 \cdot -\alpha^{(A-1)} \Rightarrow |A_2| = (-1)^{A-1} \cdot A^2 \cdot \alpha^{(2A-2)}$. By substituting A_1 and A_2 to the Eq (5) we get

$$\begin{aligned} |A - \alpha I| &= (-1)^{1+1} A_1 + (-1)^{A+1+1} A_2 = 0 \\ |A - \alpha I| &= (-1)^{1+1} (-\alpha)^{2A-1} + (-1)^{A+1+1} \{(-1)^{A-1} \cdot A^2 \cdot \alpha^{(2A-2)}\} = 0 \\ &\Rightarrow (-1)^{2A} \alpha^{2A} + (-1)^{A+2} A^2 \alpha^{2A-2} = 0. \end{aligned}$$

4 Basic Properties of A-Eccentric Tree:

Proposition 4.1. If T is A -eccentric tree and if A is one of a p -eigenvalue of a sub matrix A of $M_p(T)$, then $-A$ is another p -eigenvalue of $M_p(T)$. And multiplicity of A and $-A$ is 1 each.

Proof. Let T be A -eccentric tree and A be a sub matrix of $M_p(T)$.

$$A = \begin{pmatrix} 0 & J \\ J' & 0 \end{pmatrix}_{2A \times 2A}$$

Clearly J is $A \times A$ sub matrix of A with all its entries 1. Let x be an eigenvector of $M_p(T)$ corresponding to α . Then,

$$\begin{pmatrix} 0 & J \\ J' & 0 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \alpha \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$

And also one can verify that,

$$\begin{pmatrix} 0 & B \\ B' & 0 \end{pmatrix} \begin{pmatrix} x_1 \\ -x_2 \end{pmatrix} = \alpha \begin{pmatrix} x_1 \\ -x_2 \end{pmatrix}$$

Also, since T is A -eccentric tree $M_p(T)$, every 1^{st} column and $(A+1)^{th}$ column, every 2^{nd} column and $(A+2)^{th}$ column are linearly independent and so on, every i^{th} where $1 \leq i \leq A$ and $(i+A)^{th}$ column are linearly independent.

Clearly one linearly independent eigenvector for α produces one linearly independent eigenvector for $-\alpha$. Thus multiplicity of α and $-\alpha$ is one each. \square

Proposition 4.2. Suppose $A = 1$ then $M_p(T)$ has at least one positive p -eigenvalue α whose value is 1 and there is an eigenvector $y \neq 0$ such that $M_p(T)y = \alpha y$ for $\alpha > 0$.

Proof. Suppose T is A -eccentric tree and $A = 1$. Then clearly T has exactly two peripheral vertices. Hence, $M_p(T)$ is as follows:

$$M_p(T) = \begin{pmatrix} \boxed{0} & \boxed{1} & 0 & 0 & \dots & 0 \\ \boxed{1} & \boxed{0} & 0 & 0 & \dots & 0 \\ \hline 0 & 0 & 0 & 0 & \dots & 0 \\ 0 & 0 & 0 & 0 & \dots & 0 \\ \vdots & \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & 0 & \dots & 0 \end{pmatrix}_{n \times n}$$

Clearly a sub matrix A of $M_p(T)$ is 2×2 matrix

$$i.e., A = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}_{2 \times 2}$$

The characteristic polynomial of A is $\alpha^2 - 1 = 0$ whose p-eigenvalues are $\alpha_1 = +1$ and $\alpha_2 = -1$. Clearly α_1 and α_2 are two distinct real p-eigenvalues and one of the p-eigenvalue is 1. Thus $M_p(T)$ has at least one positive p-eigenvalue α and there is an eigenvector $y \neq 0$ such that $Ay = \alpha y$. \square

Observation 4.3. *If T is a tree with k peripheral vertices then T is A -eccentric if and only if $k = k_1 \cup k_2$ such that $|k_1| = |k_2|$.*

Observation 4.4. *Suppose a tree T with k peripheral vertices is A -eccentric then k is even.*

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