

Chapter 11

Energy of Graphs

Harishchandra S. Ramane

Karnatak University, India

ABSTRACT

The energy of a graph G is defined as the sum of the absolute values of the eigenvalues of its adjacency matrix. The graph energy has close correlation with the total π -electron energy of molecules calculated with Huckel molecular orbital method in chemistry. A graph whose energy is greater than the energy of complete graph of same order is called hyperenergetic graph. A non-complete graph having energy equal to the energy of complete graph is called borderenergetic graph. Two non-cospectral graphs are said to be equienergetic graphs if they have same energy. In this chapter, the results on graph energy are reported. Various bounds for graph energy and its characterization are summarized. Construction of hyperenergetic, borderenergetic, and equienergetic graphs are reported.

INTRODUCTION

The energy of a graph is the sum of the absolute values of the eigenvalues of its adjacency matrix. It has a correlation with the total π -electron energy of a molecule in the quantum chemistry as calculated with the Huckel molecular orbital method (Gutman & Polansky, 1986).

Let G be a finite, simple, undirected graph with vertex set $V(G)$ and edge set $F(G)$. The number of vertices of G is denoted by n and the number of edges of G is denoted by m . If $V(G) = \{v_1, v_2, \dots, v_n\}$ then the adjacency matrix of G is a square matrix $A(G) = [a_{ij}]$ of order n in which $a_{ij} = 1$, if the vertex v_i is adjacent to the vertex v_j and $a_{ij} = 0$, otherwise. The characteristic polynomial of $A(G)$ denoted by $\Phi(G; \lambda) = \det(\lambda I - A(G))$, where I is an identity matrix of order n . The roots of the equation $\Phi(G; \lambda) = 0$ are called the eigenvalues of G and they are labeled as $\lambda_1, \lambda_2, \dots, \lambda_n$. Their collection is called the spectrum of G denoted by $Spec(G)$ (Cvetkovic, Doob & Sachs, 1980).

If $\lambda_1, \lambda_2, \dots, \lambda_k$ are the distinct eigenvalues with respective multiplicities m_1, m_2, \dots, m_k then we write $Spec(G) = \left(\begin{array}{cccc} \lambda_1 & \lambda_2 & \cdots & \lambda_k \\ m_1 & m_2 & \cdots & m_k \end{array} \right)$. Since $A(G)$ is a real symmetric matrix, its eigenvalues are real and can be ordered as $\lambda_1 \geq \lambda_2 \geq \dots \geq \lambda_n$.

DOI: 10.4018/978-1-5225-9380-5.ch011

Two non-isomorphic graphs are said to be cospectral if they have same spectra. Details about the graph spectra can be found in the book (Cvetkovic, Doob & Sachs, 1980) and for graph theoretic terminology one can refer the book (Harary, 1999).

One of the chemical applications of spectral graph theory is based on the correspondence between the graph eigenvalues and the molecular orbital energy level of π -electron in conjugated hydrocarbons (Gutman & Polansky, 1986).

The molecular graph of a hydrocarbon is obtained as follows: the carbon atoms are represented by the vertices and two vertices are adjacent if and only if there is a carbon-carbon bond. Hydrogen atoms are ignored.

Within the Huckel molecular orbital (HMO) method (Huckel & Quantentheoretische Beitrage zum Benzolproblem, 1931), the energy level of π -electron in molecules of conjugated hydrocarbons are related to the eigenvalues of a molecular graph as $\varepsilon_i = \alpha + \beta_i$ where α and β are empirical constants of the HMO model. The total energy of π -electrons denoted by E_π is

$$E_\pi = \sum_{i=1}^n g_i \varepsilon_i,$$

where g_i is the occupation number with energy ε_i and $g_1 + g_2 + \dots + g_n = n$. This yields

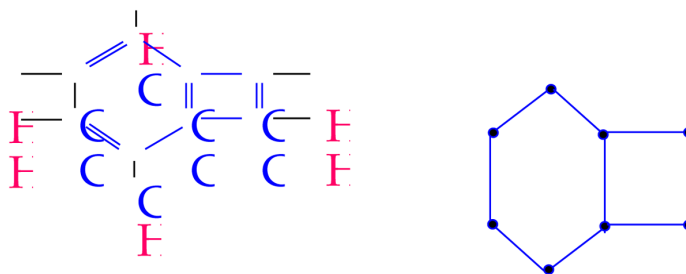
$$E_\pi = n\alpha + \beta \sum_{i=1}^n g_i \lambda_i \quad (1)$$

For majority of conjugated hydrocarbons $g_i = 2$ if $\lambda_i > 0$ and $g_i = 0$ if $\lambda_i < 0$. Therefore Eq. (1) can be written as

$$E_\pi = n\alpha + 2\beta \sum_{+} \lambda_i,$$

where \sum_{+} indicates the summation over positive eigenvalues of the molecular graph.

Figure 1. Molecule and its molecular graph



28 more pages are available in the full version of this document, which may be purchased using the "Add to Cart" button on the publisher's webpage:

www.igi-global.com/chapter/energy-of-graphs/235540

Related Content

Big Data and Analytics

Sheik Abdullah A. and Priyadharshini P. (2020). *Big Data Analytics for Sustainable Computing* (pp. 47-65).
www.irma-international.org/chapter/big-data-and-analytics/238604

Higher Order Neural Networks for Symbolic, Sub-symbolic and Chaotic Computations

João Pedro Neto (2010). *Artificial Higher Order Neural Networks for Computer Science and Engineering: Trends for Emerging Applications* (pp. 37-56).
www.irma-international.org/chapter/higher-order-neural-networks-symbolic/41661

Back-Stepping Control of Quadrotor: A Dynamically Tuned Higher Order Like Neural Network Approach

Abhijit Das, Frank L. Lewis and Kamesh Subbarao (2010). *Artificial Higher Order Neural Networks for Computer Science and Engineering: Trends for Emerging Applications* (pp. 484-513).
www.irma-international.org/chapter/back-stepping-control-quadrotor/41679

A New Data Hiding Scheme Combining Genetic Algorithm and Artificial Neural Network

Ayan Chatterjee and Nikhilesh Barik (2022). *Research Anthology on Artificial Neural Network Applications* (pp. 1522-1531).
www.irma-international.org/chapter/a-new-data-hiding-scheme-combining-genetic-algorithm-and-artificial-neural-network/289027

Novelty Detection in System Monitoring and Control with HONU

Cyril Oswald, Matous Cejnek, Jan Vrba and Ivo Bukovsky (2016). *Applied Artificial Higher Order Neural Networks for Control and Recognition* (pp. 61-78).
www.irma-international.org/chapter/novelty-detection-in-system-monitoring-and-control-with-honu/152097