

DNA Algorithm for Pancyclicity and Vertex Connectivity of Graph

Antony Xavier, Andrew Arokiaraj

Department of Mathematics, Loyola College, Chennai, India

Email: andrewarokiaraj@gmail.com

Abstract- Finding the Pancyclicity and Vertex connectivity for a general graph is the problem in NP complete class. In this paper, a bio-computation way of solving these two problems has been proposed. DNA computation is more powerful because of its massive parallelism and high density storage capacity. Thereby, producing the VERTEXCONNECTIVITY, PANCYCLIC and GIRTH algorithm solvable in polynomial time. The method in tracing the k-cycle and k-cut vertex was also used for the computation.

Keywords: DNA algorithm, pancyclic, Vertex connectivity, Girth.

I. INTRODUCTION

DNA computation is an emerging Bio-computing model in solving most of the NP complete problems, and making them solvable in polynomial time. The seminal work was done by Adleman in solving Hamiltonian path problem and Lipton solving the satisfiability problem. This new computational method has more parallelism and high storage capacity, which brings enormous break-through in theory of computation, graph theory, cryptography. Further both these computing ideas were extended in solving graph theoretical problems like TSP, coloring, maximum clique, isomorphism, vertex cover, etc. In this paper we use DNA to compute the pancyclicity, girth, vertex connectivity of the graph. For this we imbibe the style which Adleman used in solving computational problems with DNA. The various DNA operators used in the algorithm were also explained.

II. GRAPH PROBLEMS

The following are the graph theoretical problems, for which the DNA algorithm in polynomial time have been suggested.

Vertex Connectivity:

The minimum number of nodes whose deletion from a graph G disconnects it. Vertex connectivity is sometimes called "point connectivity" or simply "connectivity."

Pancyclic:

An n -vertex graph G is pancyclic if, for every k in the range $3 \leq k \leq n$, G contains a cycle of length k .

Girth:

The girth of a graph is the length of a shortest cycle contained in the graph

III. MOLECULAR BIOLOGY

DNA (Deoxyribonucleic acid) is double stranded helical in structure. Each strand is a chain of nucleotide. The four

nucleotides are adenine (A), guanine (G), Thymine (T), cytosine (C). DNA obeys the following complementarity: A always pairs with T and G always pairs with C known as Watson-Crick complementarities.

DNA IN CODING OF A PATH

The idea of encoding a graph path by DNA was introduced by Adleman. Consider a graph in fig 1.

Each of its vertex is represented by a string of length 'n' using combination of A,G,T,C with $m=2$, let $v_1=AA$, $v_2=AC$, $v_3=TT$, $v_4=CC$, $v_5=AT$. A path is given by double stranded helix

IV. STICKER BASED DNA COMPUTATION

The sticker model was introduced by S. Roweis et al. In this model, there is a memory strand with N bases in length subdivided into K non-overlapping regions each M bases long ($N > MK$). M can be for example 20. The subregions (bit regions) are significantly different from each other. One sticker is designed for each subregion; each sticker has M bases long and is complementary to one and only one of the K memory regions. If a sticker is annealed to its corresponding region on memory strand, then the particular region is said to be on. If no sticker is annealed to a region then the corresponding bit is off. Each memory strand along with its annealed stickers is called memory complex. In sticker model, a tube is a collection of memory complexes, composed of large number of identical memory strands each of which has stickers annealed only at the required bit positions. This method of representation of information differs from other methods in which the presence or absence of a particular subsequence in a strand corresponded to a particular bit being on or off. In sticker model, each possible bit string is represented by a unique association of memory strands and stickers. This model has a random access memory that requires no strand extension and uses no enzymes.

STICKER OPERATIONS:

- **separate (T, T^+, T^- , i):** Consider a test tube T and an integer i and create two new test tubes T^+ and T^- , where T^+ consists of all memory complexes in T in which i th substrand is on, while T^- is comprised of all memory complexes in T in which the i th substrand is off.
- **incident(i, j):** Checks the existence for i th vertex and j th vertex to be adjacent to each other.
- **merge(T_1, T_2, \dots, T_n):** Take tube T_1, \dots, T_n , produce their union $T_1 \cup T_2 \cup \dots \cup T_n$, put the result into the (possibly empty) tube T_n .

- discard (T_0): Take a test tube T and empty its contents.
- set ($T-, n+m+i_j$): Start with a test tube T and an integer i and generate a test tube in which the i th strand of each memory complex is turned on.
- clear ($T++, n+m+i_j$): Start with a test tube T and an integer i and generate a test tube in which the i th strand of each memory complex is turned on.

V. SUGGESTED ALGORITHM

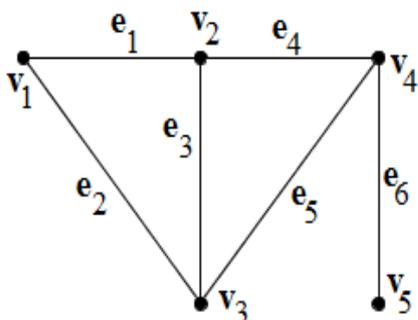
The initial test tube T_0 contains generated random paths of Graph G . This is achieved by annealing and ligation. The time complexity for this generation is taken as 1 unit because of immense parallelism.

The algorithm INCIDENTRELATION provides the incidence relation between vertices and edges of a graph. The input of the algorithm is an $[m+n ; \binom{n}{k}]$ library T , providing encoded DNA of all k -subsets of vertices, where $1 < k < n$. Two parameters lower (l) and upper (u) bound on the set of strands of size n , where $1 < l < u < n$. For those memory complexes whose i th strands is turned on, $1 < i < u$, the algorithm verifies in parallel if the vertex-edge pair (v_i, e_j) is incident and if so, turns on the $u+j$ th strand corresponding to the incident edge (statements 4-5). At the end of the loop, the strands composed of the last m substrands provide the incidence pairs (v_i, e_j) between vertices and edges. The algorithm requires $n(m+2)$ steps.

INCIDENTRELATION(T, l, u, m, n)

Input: $[m+n ; \binom{n}{k}]$ library T , $1 < l < u < n$

01. for $i \leftarrow 1$ to u do
 02. separate ($T, T+, T-, i$)
 03. for $j \leftarrow 1$ to m do
 04. if incident(i, j) then
 05. set ($T+, u+j$)
 06. end if
 07. end for
 08. merge($T+, T-, T$)
 09. end for
 10. return T
- $l=1$ and $u=5$



v_1	v_2	v_3	v_4	v_5	e_1	e_2	e_3	e_4	e_5	e_6
1	0	0	0	0	1	1	0	0	0	0
0	1	0	0	0	1	0	1	1	0	0
0	0	1	0	0	0	1	1	0	1	0
0	0	0	1	0	0	0	0	1	1	1
0	0	0	0	1	0	0	0	0	0	1

The algorithm WEIGHTENING extracts from an input test tube T_0 those memory complexes in which exactly k of the sunstrands $m+1, \dots, m+n$ are turned on, where $0 < k < n$. At the end of the loop (1-7), the test tube T_i , where $0 < i < n$, contains all memory complexes in which exactly i of the substrands $m+1, \dots, m+n$ are turned on. Thus the test tube T_k provides the output of the algorithm. The sticker algorithm requires $2n[(n+1)/2] = n^2 + n$ steps.

WEIGHTENING (T_0, m, n, k)

Input: input test tube T_0

01. for $i \leftarrow 0$ to $n-1$ do
02. for $j \leftarrow i$ down to 0 do
03. separate ($T_j, T+, T-, m+i+1$)
04. merge ($T+, T_{j+1}$)
05. merge ($T-, T_j$)
06. end for
07. end for
08. return T_k

Consider an input tube T_0 providing encoded DNA of the memory complexes 0001, 0110, 1011. The computation of weighting for $m=0$ and $n=4$ is as follows.

	T_0	T_1	T_2	T_3	T_4
Initial	0001 0110 1011				
$i=1$	0001 0110	1011			
$i=2$	0001	0110 1011			
$i=3$	0001		0110 1011		
$i=4$		0001	0110	1011	

K-CYCLE (T, m, n, k)

Input: $[m+n ; \binom{m}{k}]$ library T

01. $T \leftarrow \text{EDGEINDUCEDGRAPHS}(T, m, n, k)$
02. $T \leftarrow \text{WEIGHTENING}(T, m, n, k)$
03. for $i \leftarrow 1$ to m do
04. separate ($T, T+, T-, i$)
05. for $j \leftarrow 1$ to 2 do
06. separate ($T+, T++, T+, n+m+i_j$)
07. set ($T+, n+m+i_j$)
08. clear ($T++, n+m+i_j$)
09. end for
10. end for
11. $T \leftarrow \text{WEIGHTENING}(T, m+n, n, 0)$
12. if \neg empty (T) then
13. return (T)
14. else
15. report "No k -cycle"
16. end if

GIRTH (T, m, n)

Input: $[n+(m+n) ; \binom{n}{k}]$ library T

01. for $k \leftarrow 3$ to n do
02. $T_0 \leftarrow \text{K-CYCLE}(T, m, n, k)$
03. if \neg empty (T_0) then
04. return (T_0)

05. end if
06. end for

PANCYCLIC (T, m, n)Input: $[n+(m+n); \binom{n}{k}]$ library T

01. for $k \leftarrow 3$ to ndo
02. $T_0 \leftarrow$ K-CYCLE (T, m, n, k)
03. if empty (T_0) then
04. report "Not Pancyclic"
05. end if
06. discard (T_1)
07. end for
08. report "Graph is Pancyclic"

K-CUTVERTEX (T, m, n, k)Input: $[2n; \binom{n}{k}]$ library T

01. for $i \leftarrow 1$ to $n-1$ do
02. separate ($T, T+, T-, i$)
03. set ($T-, n+i$)
04. for $p \leftarrow n+i$ to $2n-1$ do
05. separate ($T-, T+, T-, p$)
06. for $q \leftarrow 2n$ down to $n+i+1$ do
07. if adjacency (p, q) then
08. set ($T+, q$)
09. end if
10. end for
11. for $j \leftarrow 1$ to ndo
12. separate ($T+, T++, T+-, j$)
13. clear ($T++, n+j$)
14. merge ($T++, T+-, T+$)
15. end for
16. merge ($T+, T-, T-$)
17. end for
18. merge ($T-, T_0$)
19. merge ($T+, T$)
20. end for
21. $T \leftarrow$ WEIGHTENING ($T, n+1, n, k+1$)
22. if \neg empty (T) then
23. report "k-cut vertex"
24. return (T)
25. else
26. report "No k-cut vertex"
27. end if

VERTEXCONNECTIVITY (T, m, n)Input: $[2n; \binom{n}{k}]$ library T

01. for $k \leftarrow 2$ to $n-2$ do
02. $T \leftarrow$ K-CUTVERTEX (T, m, n, k)
03. if \neg empty (T_1) then
04. report "k-vertex connectivity"
05. return (T_1)
06. end if
07. end for

VI. CONCLUSION

This paper demonstrates that Wiener Index can be solved using DNA computation. This method suggested can be applied for general graph in finding Girth, Pancyclic and Vertex Connectivity, which are NP-complete problems. The

complexity of these algorithm in worse case yields $O(n^2)$. Also, the method in tracing the shortest path was also discussed within the algorithm.

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