

Heuristic algorithm for solving the graph isomorphism problem

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In the paper we consider heuristic algorithm for solving graph isomorphism problem. The algorithm based on a successive splitting of the eigenvalues of the matrices which are modifications (to positive defined) of graphs' adjacency matrices. Modification of the algorithm allows to find a solution for Frobenius problem. Formulation of the Frobenius problem is following one. Given a pair of two matrices with the same number of rows and columns. We must find out whether one of the matrix can be acquired from another by permutation of it's rows and strings or not. For example, solution of Frobenius problem can give to us efficient way for decrypting of double permutation cyphers problem for high dimension matrices.

The graph isomorphism problem

Graph isomorphism problem is one of the problems for which we can't say definitely whether this problem is polynomial or not [1]. But it's known that this problem is polynomial for some classes of graphs such as plane graphs, regular graphs and some others [2], [3], [4]. Our algorithm is heuristic algorithm for solving of the problem.

We consider heuristic algorithm for solving the graph isomorphism problem. The algorithm based on successive splitting of eigenvalues of the matrices that are modifications (to positive defined) of graphs' adjacency matrices. During the work of the algorithm we solving the linear equations that defines inverse matrices of these matrices. Solutions of these systems gives to us a permutation which is sought bijection.

There are two nonoriented graphs in the graph isomorphism problem: graph $G_A = \langle V_A, E_A \rangle$ and graph $G_B = \langle V_B, E_B \rangle$. V_A , V_B are sets of graphs' vertices. E_A , E_B are sets of it's edges. They are supposed to be the sets of the same power: $|V_A| = |V_B|$, $|E_A| = |E_B|$. Graph isomorphism problem formulation is following one: whether exists such bijection $\varphi : V_A \rightarrow V_B$ that if $(i, j) \in E_A$ then $(\varphi(i), \varphi(j)) \in E_B$ or not?

Algorithm works with modifications of adjacency matrices. Let A_0 be an adjacency matrix of G_A , that is to say that $A_0 = (a_{ij}^0)$ and

$$a_{ij}^0 = \begin{cases} 1, & \text{if } (i, j) \in E_A, \\ 0, & \text{else.} \end{cases}$$

Let B_0 be an adjacency matrix of graph G_B .

Construct matrix D_{A_0} according to A_0 :

$$\begin{pmatrix} d_1 & 0 & \dots & 0 \\ 0 & d_2 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & d_n \end{pmatrix}$$

D_{A_0} is a scalar matrix with following elements:

$$d_i = \sum_{j=1}^n a_{ij}^0 + 1.$$

Construct matrix D_{B_0} similarly according to B_0 . Let

$$A = A_0 + D_{A_0}, \quad B = B_0 + D_{B_0}. \quad (1)$$

Matrices A and B are further considered matrices. They are matrices which algorithm works with. They are symmetric and positive defined.

If graphs $G_A = \langle V_A, E_A \rangle$ and $G_B = \langle V_B, E_B \rangle$ are isomorphic then we can acquire matrix A from matrix B by successive permutations of it's rows with simultaneous permutations of it's columns with same numbers. So we can formulate the graph isomorphism problem as a particular case of the Frobenius problem which formulation is following one. Can we acquire matrix B from matrix A by successive permutations of it's rows and columns?

Permutation of rows numbers i and number j for arbitrary matrix is equal to right multiplying matrix by permutation matrix P_{ij} . Permutation of columns is equal to left multiplying matrix by the same permutation matrix. Permutation of vector's components with numbers i and j take place for both left and right multiplying vector by permutation matrix P_{ij} .

Let us consider the following systems of linear equations:

$$Ax = e_j, \quad By = e_k. \quad (2)$$

The vectors $e_i = (0, \dots, 0, 1, 0, \dots, 0)$ are basis vectors in the space R^n , matrices A and B such as described above. Both systems are solvable and each solution is unique solution because of A and B are matrices with diagonal predominance and their determinants aren't equal to zero. Let x^j be a solution of the system $Ax = e_j$ and y^k be a solution of the system $By = e_k$.

Note that solution of the systems of linear equations (1) gives to us inverse matrices of matrix A and matrix B . So for i -th component of vector x^j the following is true: $x_i^j = (-1)^{i+j} A_{ij} / \det(A)$, where A_{ij} are algebraic adjuncts for element a_{ij} of the matrix A . That is to say x^j , $j = 1, \dots, n$ are columns of inverse matrix for A .

If $B = P_{jk}AP_{jk}$ then the following is true for solutions of the systems of linear equations (2):

$$\begin{aligned} x_i &= y_i, \quad i \neq j, \quad i \neq k; \\ x_j &= y_k, \quad x_k = y_j. \end{aligned}$$

Indeed: $(Ax = e_j) \sim (P_{jk}Ax = P_{jk}e_j) \sim (P_{jk}AxP_{jk} = P_{jk}e_jP_{jk}) \sim (P_{jk}AP_{jk}x = e_j) \sim (P_{jk}AP_{jk}xP_{jk} = e_jP_{jk}) \sim (BxP_{jk} = e_k)$. That is to say $xP_{jk} = y$.

If matrix B can be acquired from matrix A by successive permutations of it's rows with simultaneous permutations of it's columns with same numbers then:

$$B = P_{j_1k_1} \dots P_{j_1k_1}AP_{j_1k_1} \dots P_{j_1k_1},$$

and consequently $xP_{j_1k_1} \dots P_{j_1k_1} = y$.

So if we'll change vector e_k in the system of linear equations (2) with fixed j by changing index k from 1 to n then vectors x^j and y^k , which are corresponding solutions of the systems (2), will be the same vectors correct to permutation of their components only if row with number j of matrix A corresponds to row with

number k of matrix B . That is to say elements of row number k of matrix B are permuted elements of row number i of matrix A . The same is true for columns of the matrices.

We'll accomplish the successive perturbations of its diagonal elements for finding unique row and unique column of the matrix B that corresponds to row and column with number i of the matrix A . The algorithm works with symmetric matrices which can be transformed into scalar matrices by orthogonal transformations. So

$$\begin{aligned}\tilde{A} &= U_A A U_A^T, \\ \tilde{B} &= U_B B U_B^T,\end{aligned}$$

where \tilde{A} , \tilde{B} are scalar matrices with eigenvalues on their diagonals and U_A , U_B are matrices of the orthogonal transformations. Diagonal elements of \tilde{A} and \tilde{B} are eigenvalues of the matrices A and B .

The spectrums of the isomorphic graphs' adjacency matrices are the same [5]. The same true for the matrices which algorithm works with because of described above procedure of changing of the diagonal elements of adjacency matrices leads only to shifting spectrums of the matrices. So if the matrices correspond to the isomorphic graphs then spectrums of the matrices will be congruent after shifting.

It's obvious that if the spectrums of the both matrices is simple or multiplicity of eigenvalues is not big then the problem of graph isomorphism can be solved by a comparison of rows of the matrices U_A , U_B , \tilde{A} and \tilde{B} [7].

The main difficulties arise with consideration of graphs which spectrums contains multiple eigenvalues. Perturbation of the matrices will allow us to perturb their spectrums while algorithm works. Splitting of the eigenvalues will take place by this perturbation. So we can maintain a unique correspondence between rows and columns of the matrices.

If there is multiple eigenvalues in spectrums of the matrices and then they will be split and we'll be able to find the permutation which is sought bijection φ . This bijection establishes isomorphism of graphs $G_A = \langle V_A, E_A \rangle$ and $G_B = \langle V_B, E_B \rangle$. Computing experiments give us evidence that split needed for determination of the corresponding rows and columns take place much early than at the last iteration of the algorithm. So at iteration with number \sqrt{n} there is no need for further perturbations of diagonal elements of the matrices A and B for maintaining a unique correspondence between vertices of graphs that represents lattice on torus which numbers of vertices is equal to the n where n changes from 9 up to 400.

At the algorithm's implementation on every iteration algorithm works with already perturb matrices not with initial matrices. So if we've got a correspondence between row number j of matrix A^j and row number k of matrix B^j at the iteration number j and for columns with the same numbers then we'll considerate further perturb matrices A^{j+1} and B^{j+1} :

$$A^{j+1} = A^j + \varepsilon C^j, \quad B^{j+1} = B^j + \varepsilon C^k.$$

We make perturbation by scalar matrices C^k with elements c_i on it's diagonal:

$$c_i = \begin{cases} 1, & \text{if } i = j = k, \\ 0, & \text{else.} \end{cases}$$

As a result if matrix A can be acquired from matrix B by successive permutations of it's rows with simultaneous permutations of it's columns with same numbers then

we get the sought permutation P while j changes from 1 to n . That is to say

$$P = \begin{pmatrix} 1 & 2 & \dots & n \\ k_1 & k_2 & \dots & k_n \end{pmatrix},$$

k_j is a number of row of matrix B that obtained at j -th iteration of the algorithm.

As a matter of fact permutation P is one of the possible permutations. So P is a bijection $\varphi : V_A \rightarrow V_B$ that sets isomorphism of graphs G_A and G_B .

The spectral splitting algorithm

(First schema)

Step 0. $A^0 := A$, $j := 1$.

Step 1. If $j < n$ then $A^j := A^{j-1} + \varepsilon C^j$, else stop the algorithm's implementation.

Step 2. Solving of the system of linear equations $Ax = e_j$. x^j is the solution.

Step 3. $k := 1$. If $k < n$ then go to Step 3.1, else go to Step 4.

Step 3.1. $B^k := B^{k-1} + \varepsilon C^k$.

Step 3.2. Solving of the system of linear equations $B^k y = e_k$. y^k is the solution

Step 3.3. $k := k + 1$. Go to Step 3.

Step 4. Comparing norms of x^j and y^k , $k = 1, \dots, n$.

If $\forall k \|x^j\| \neq \|y^k\|$ then graphs G_A and G_B are not isomorphic. Stop the algorithm's implementation.

If exists $k : \|x^j\| = \|y^k\|$ $x_j^j = y_k^k$ then $P(j) := k$ (Maintaining a correspondence between vertex j of graph G_A and vertex k of graph G_B), $B^j := B^{j-1} + \varepsilon C^k$. Else graphs G_A and G_B are not isomorphic. Stop the algorithm implementation.

Step 5. $j := j + 1$. Go to Step 1.

The hardness of this scheme is equal to $O(n^4)$, n is a number of rows at the square matrices A and B . We have to notice that this scheme can be modified to another scheme which laboriousness is equal to $O(n^{3.5})$. We can make it by adding the split checking procedure (Step 6 and step 7 of the scheme stated below).

The spectral splitting algorithm

(Second schema)

Step 0. $A^0 := A$, $j := 1$.

Step 1. If $j < n$ then $A^j := A^{j-1} + \varepsilon C^j$, else stop the algorithm's implementation.

Step 2. Solving of the system of the linear equations $Ax = e_j$. x^j is the solution.

Step 3. $k := 1$. If $k < n$, then go to Step 3.1, else then go to Step 4.

Step 3.1. $B^k := B^{j-1} + \varepsilon C^k$.

Step 3.2. Solving of the system of the linear equations $B^k y = e_k$. y^k is the solution.

Step 3.3. $k := k + 1$. Go to Step 3.

Step 4. Comparing norms of x^j and y^k , $k = 1, \dots, n$.

If $\forall k \|x^j\| \neq \|y^k\|$ then graphs G_A and G_B are not isomorphic. Stop the algorithm's implementation.

If exists $k : \|x^j\| = \|y^k\|$ $x_j^j = y_k^k$ then $P(j) := k$ (Maintaining a correspondence between vertex j of graph G_A and vertex k of graph G_B), $B^j := B^{j-1} + \varepsilon C^k$. Else graphs G_A and G_B are not isomorphic. Stop the algorithm's implementation.

Step 5. $j := j + 1$.

Step 6. Solving of the system of linear equations $A^j x = e_l$. x^l , $l = 1, \dots, n$ is the solution.

Step 7. If $\forall l \forall p : \|x^l\| \neq \|x^p\|$ Go to Step 8, else go to Step 1.

Step 8. If $j < n$ then implement solving of the system of the linear equations

$B^k y = e_k$, $k = 1, \dots, n$; where k such that not exists $i : P(i) = j$.

Step 9. Comparing norms of $\|x^l\|$ and $\|y^k\|$, $l = j, \dots, n$, $k = 1, \dots, n$; and k such that not exists $i : P(i) = j$. If $\|x^l\| = \|y^k\|$ then $P(l) := k$.

Step 10. Stop the algorithm's implementation.

Though the spectral splitting algorithm show oneself good by computing experiments we have to notice that we can't guarantee finding solution of the graph isomorphism problem for arbitrary graphs because of we can't say that the following situation is impossible. There are two vectors x_j and y_k with equal or such close norms for we can't say whether one vector can be acquired from another by permutation of it's components or not because of we implements check procedure only by one components of each vectors. They are components x_j^j and y_k^k .

In particular we realized computing experiments for regular graphs with number of vertices up to 2500 for the graph isomorphism problem.

Solving of the Frobenius problem for arbitrary square matrices of complete rank

The spectral splitting algorithm for the graph isomorphism problem can be applied for weighted graphs without any modifications. Scheme of the algorithm is the same; the only difference in matrices algorithm works with.

There are two nonoriented graphs in the graph isomorphism problem: graph $G_A = \langle V_A, E_A \rangle$ and graph $G_B = \langle V_B, E_B \rangle$. V_A, V_B are sets of graphs' vertices, E_A, E_B are sets of it's edges. They are supposed to be the sets of the same power: $|V_A| = |V_B|$, $|E_A| = |E_B|$. Given a function $H_A : E_A \rightarrow R$ $H_B : E_B \rightarrow R$ which defines weights of graphs' edges. Graph isomorphism problem for weighted graphs formulation is following one: whether exists such bijection $\varphi : V_A \rightarrow V_B$, that if $(i, j) \in E_A$, then $(\varphi(i), \varphi(j)) \in E_B$ and if $(i, j) \in E_A$ then $(\varphi(i), \varphi(j)) \in E_B$ $H_A(i, j) = H_B(\varphi(i), \varphi(j))$ or not?

The adjacency matrices of weighted nonoriented graphs transforms into positive defined matrices with diagonal predominance, but we should select its diagonal elements in such way that conditionality number should be confined.

Let matrix $A_0 = (a_{ij}^0)$ be an adjacency matrix of weighted nonoriented graph G_A and matrix D_{A_0} be a diagonal matrix with such elements d_i that:

$$d_i = d + \sum_{j=1}^n a_{ij}^0,$$

where $d = \max_{1 \leq i \leq n} \sum_{j=1}^n a_{ij}^0$.

Construct matrix D_{B_0} for weighted nonoriented graph G_B by same way according to adjacency matrix of graph B_0 .

For conditionality number $\mu(A)$ of the symmetric matrix A the following is true [6]:

$$\mu(A) \leq \frac{\eta(A)}{\chi(A)},$$

where $\eta(A) = \max_{1 \leq i \leq n} (a_{ii} + \sum_{j \neq i} |a_{ij}|)$, $\chi(A) = \min_{1 \leq i \leq n} (a_{ii} - \sum_{j \neq i} |a_{ij}|)$.

By our choice of d :

$$\eta(A) = \max_{1 \leq i \leq n} (a_{ii} + \sum_{j=1}^n |a_{ij}^0|) = a_{i_1 i_1} + \sum_{j=1}^n |a_{i_1 j}^0| = d_{i_1} + \sum_{j=1}^n |a_{i_1 j}^0| = d + 2 \sum_{j=1}^n |a_{i_1 j}^0| =$$

$$\begin{aligned}
&= 3 \sum_{j=1}^n |a_{i_1 j}^0|, \\
\chi(A) &= \min_{1 \leq i \leq n} (a_{ii} - \sum_{j=1}^n |a_{ij}^0|) = a_{i_2 i_2} - \sum_{j=1}^n |a_{i_2 j}^0| = d_{i_2} - \sum_{j=1}^n |a_{i_2 j}^0| = \sum_{j=1}^n |a_{i_2 j}^0| + d - \\
&- \sum_{j=1}^n |a_{i_2 j}^0| = \sum_{j=1}^n |a_{i_2 j}^0| + \sum_{j=1}^n |a_{i_1 j}^0| - \sum_{j=1}^n |a_{i_2 j}^0| = \sum_{j=1}^n |a_{i_1 j}^0|.
\end{aligned}$$

Consequently

$$\mu(A) = \frac{\eta(A)}{\chi(A)} \leq \frac{3 \sum_{j=1}^n |a_{i_1 j}^0|}{\sum_{j=1}^n |a_{i_1 j}^0|} = 3.$$

There are two matrices F_A and F_B with the same number of rows and columns in the Frobenius problem. We must find out whether one matrix can be acquired from another by some permutation of its rows and columns.

Suppose that matrices F_A and F_B are square matrices of the complete rank with number of rows equal to n . Matrix A_0 of the following structure:

$$\begin{pmatrix} 0 & F_A \\ F_A^T & 0 \end{pmatrix}.$$

This matrix can be considered as an adjacency matrix of some weighted nonoriented graph. Construct matrix D_{A_0} with structure described above for matrix A_0 . Then construct matrix A :

$$A = A_0 + D_{A_0}.$$

That is to say

$$A = \begin{pmatrix} D_{A_0}^1 & F_A \\ F_A^T & D_{A_0}^2 \end{pmatrix},$$

and

$$D_{A_0} = \begin{pmatrix} D_{A_0}^1 & 0 \\ 0 & D_{A_0}^2 \end{pmatrix},$$

where

$$D_{A_0}^1 = \begin{pmatrix} d_1 & 0 & \dots & 0 \\ 0 & d_2 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & d_n \end{pmatrix}, \quad D_{A_0}^2 = \begin{pmatrix} d_{n+1} & 0 & \dots & 0 \\ 0 & d_{n+2} & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & d_{2n} \end{pmatrix}.$$

So matrix A is matrix for initial matrix F_A for algorithm to work with.

Matrix A is positive defined symmetric matrix with diagonal predominance. It corresponds to some weighted nonoriented dicotyledonous graph. Matrix F_B is a square matrix which number of rows the same as the number of rows of matrix F_A .

Construct matrix B by the same way for matrix F_B . Form the solving permutation of rows and columns of matrices A and B by applying the spectral splitting algorithm to matrices A and B .

Form the solving permutation for initial matrices F_A and F_B by the permutations of rows and columns of the matrices A and B in the following way. Let it given by the algorithm that the row and column with number i of matrix A corresponds to the row and column with number j of matrix B . It is obvious that if $1 \leq i \leq n$, $1 \leq j \leq n$ then the permutation of pair of rows $\{i, j\}$ with simultaneous permutation of its columns $\{i, j\}$ of matrix A corresponds to permutation of rows number i and

number j of initial matrix F_A . If $n + 1 \leq i \leq 2n$, $n + 1 \leq j \leq 2n$ then permutation of pair of rows $\{i, j\}$ with simultaneous permutation of columns $\{i, j\}$ of matrix A corresponds to permutation of columns with number i and number j of initial matrix F_A . As soon as matrices A and B which algorithm works with have the structure as following one:

$$\begin{pmatrix} * & 0 & \dots & 0 & * & * & \dots & * \\ 0 & * & \dots & 0 & * & * & \dots & * \\ \vdots & \vdots & \ddots & \vdots & \vdots & \vdots & & \vdots \\ 0 & 0 & \dots & * & * & * & \dots & * \\ * & * & \dots & * & * & 0 & \dots & 0 \\ * & * & \dots & * & 0 & * & \dots & 0 \\ \vdots & \vdots & & * & \vdots & \vdots & \ddots & \vdots \\ * & * & \dots & * & 0 & 0 & \dots & * \end{pmatrix},$$

(* are positions admissible for nonzero elements) so the situation of maintaining of the correspondence for pair of rows and pair of columns of matrices A and B with such numbers i and j that $\{i, j\}$ and $1 \leq i \leq n$, $n + 1 \leq j \leq 2n$ is impossible. It is impossible because of it means appearance of nonzero nondiagonal elements at the left and upper part of the matrix A (of the matrix $D_{A_0}^1$) and right and lower part (of the matrix $D_{A_0}^2$) if only the initial matrices F_A and F_B are not scalar. This is impossible by constructing of the matrices A and B . Consequently appearance of such correspondence is impossible while algorithm works.

Deciphering of the double permutation cipher by the spectral splitting algorithm

Suppose we have a text presented by some square matrix of symbols. Let us form matrix of the cipher with some digital code for every symbol of text. The second matrix presented some coding by this way text too. It's needed to find out whether the text presented by the second matrix is the double permutation cipher for the first matrix or not. That is to say whether this text can be obtained from the first text presented by matrix by permutation of its rows and columns. Such problem may be interpreted as a case of the Frobenius problem for digital matrices. So we can apply the spectral splitting algorithm for solving of this problem.

There are computational experiments was implemented for deciphering of the double permutation ciphers for texts with number of symbols up to 10000.

References

1. M. Garey and D. Johnson *Computers and Intractability, A Guide to Theory of NP-completeness*. Freeman and Co., 1979
2. Hopcroft, Wong *A linear time algorithm for isomorphism of planar graphs*. Proceedings of the Sixth Annual ACM Symposium on Theory of Computing, p. 172-184, 1974.
3. Luks *Isomorphism of graphs of bounded valence can be tested in polynomial time*. Proc. 21st IEEE FOCS Symp., 42, 49, 1980.

4. Hoffmann *Group-Theoretic Algorithms and Graph Isomorphism*. Lecture Notes in Computer Science (Chapter V). P.127-138, 1982

5. Tzvetkovich D. and others *Spectrums of graphs. Theory and application*. Kiev: The scientific thought, 1984.

6. Godunov S. and others *Guaranteed exactness of solutions of systems of linear equations in euclidian spaces*. Novosibirsk: Science, 1988.

7. Kikina A., Faizullin R. *The algorithm for testing graph isomorphism*. VINITI 21.06.95 1789-95.