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# Algorithm 455

# Analysis of Skew Representations of the Symmetric Group [Z]

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Key Words and Phrases: symmetric group, skew representation, partition, Young diagram, lattice permutation, binary model, outer product

CR Categories: 5.30 Language: Algol

## Description

This algorithm analyzes the skew representation  $[\lambda]-[\mu]$  of the symmetric group  $\sigma_n$  corresponding to a pair of partitions

 $(\lambda) = (\lambda_1, \lambda_2, \dots, \lambda_r)$  and  $(\mu) = (\mu_1, \mu_2, \dots, \mu_s)$  where

$$\begin{array}{c} r \geq s \\ \lambda_1 \geq \lambda_2 \geq \cdots \geq \lambda_r \\ \mu_1 \geq \mu_2 \geq \cdots \geq \mu_s \\ \lambda_i \geq \mu_i \quad (1 \leq i \leq s) \\ n = \sum_{i=1}^r \lambda_i - \sum_{i=1}^s \mu_i \end{array}$$
(1)

(see Robinson [4, sec. 2.5]). The analysis takes the form

$$[\lambda] - [\mu] = \sum_{(\nu)} c_{(\nu)} [\nu], \qquad (2)$$

where the summation is over all partitions ( $\nu$ ) of *n*, the coefficients  $c_{(\nu)}$  being nonnegative integers.

The method used may be described as follows: construct all possible diagrams which can be built up in accordance with the following two rules.

(a) Replace  $\mu_s$  of the nodes in the Young diagram corresponding to ( $\lambda$ ) by identical symbols  $\alpha_s$  in such a way that: (i) the unchanged

nodes form a regular Young diagram; and (ii) no two identical symbols  $\alpha_s$  lie in the same column. Then replace  $\mu_{s-1}$  further nodes by identical symbols  $\alpha_{s-1}$  in accordance with the same rules, and so on, finally replacing  $\mu_1$  nodes by identical symbols  $\alpha_1$ .

(b) In the final diagram the altered nodes should form a lattice permutation of  $\alpha_1^{\mu_1}\alpha_2^{\mu_2}\cdots\alpha_s^{\mu_s}$  (Robinson [4, sec. 2.4]) when read from right-to-left through successive rows.

Then the pattern of unchanged nodes in each diagram so constructed defines a term  $[\nu]$  in the analysis.

This method appears not to have been explicitly stated in the above form before, but is an immediate consequence of Littlewood's method for analyzing the outer product  $[\lambda]$ ,  $[\mu]$  (see Littlewood [3, sec. 6.3, th. V], Robinson [4, sec. 3.3]), noting that  $c_{(\nu)}$ is also the coefficient of  $[\lambda]$  in the analysis of  $[\mu]$ ,  $[\nu]$  (Littlewood [3, sec. 6.4, th. VIII]).

In the procedure, binary models of those partitions  $(\nu)$  in (2) for which  $c_{(\nu)} \neq 0$  are stored, in lexicographic order, in nu[1],  $nu[2], \ldots, nu[p]$ , the corresponding values  $c_{(\nu)}$  being stored in  $c[1], c[2], \ldots, c[p]$ . The binary model used is due to Comét [1], a partition  $(\nu) = (\nu_1, \nu_2, \ldots, \nu_t)$  being represented by the number  $2^{n-\nu_1} + 2^{n-\nu_1-\nu_2} + \cdots + 2^{\nu_t} + 1.$  (3)

The techniques used are similar to those employed in [2]. In particular, two two-dimensional arrays *lam* and *sigma* are required. Corresponding to any particular diagram, *lam* [i, j] specifies the number of nodes in row *j* which are still unchanged when all the symbols  $\alpha_s$ ,  $\alpha_{s-1}$ , ...,  $\alpha_i$  have been inserted (j = i, i + 1, ..., r), and *sigma* [i, j] specifies the total number of symbols  $\alpha_i$  inserted in rows *i*, i + 1, ..., j. Thus the quantities *lam*[i, j] are generated by the equation

$$lam[i,j] = lam[i+1,j] - sigma[i,j] + sigma[i,j-1].$$

$$\tag{4}$$

The rules for constructing the diagrams impose the restrictions

$$sigma[i-1,j-1] \ge sigma[i-1,j] - lam[i,j] + lam[i,j+1]$$
 (5)  
and

$$sigma[i-1,j-1] \ge sigma[i,j].$$
(6)

Each time array *lam* is completed, a term

$$(\nu) = (lam[1,1], lam[1,2], \dots, lam[1,r])$$
(7)

is added to the analysis.

Note 1. In view of the identity

 $[\lambda]. [\mu] = [\lambda_1 + \mu_1, \lambda_1 + \mu_2, \ldots, \lambda_1 + \mu_s, \lambda_1, \lambda_2, \ldots, \lambda_r] - [\lambda_1^s],$ 

procedure *skew* may also be used to analyse the outer product  $[\lambda]$ .  $[\mu]$ . It is, however, less convenient for this purpose than procedure *outer product* of Hunter [2].

Note 2. Value of p. It is difficult to predict the value of p in any example. Clearly,  $p \le p(n)$ , where p(n) denotes the number of partitions of n. On the other hand, for any value of n, there are partitions ( $\lambda$ ) and ( $\mu$ ) for which p = p(n), namely, ( $\lambda$ ) =  $(n, n-1, \ldots, 1)$ , ( $\mu$ ) =  $(n-1, \ldots, 1)$ .

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#### Algorithm

procedure skew (r, s, lambda, mu, p, c, nu);

value r, s; integer r, s, p; integer array lambda, mu, c, nu; begin

comment Input parameters.

- r: the number of parts in partition  $(\lambda)$ .
- s: the number of parts in partition  $(\mu)$ .

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*lambda*: the part  $\lambda_i$  is stored in *lambda*[i], i = 1, 2, ..., r. mu: the part  $\mu_i$  is stored in mu[i],  $i = 1, 2, \ldots, s$ . Output parameters. D: the number of terms on the right in (2) for which  $c_{(\nu)} \neq 0.$ Binary models (3) of the partitions ( $\nu$ ) in (2) for which nu:  $c_{(\nu)} \neq 0$  are placed in lexicographic order in nu[1],  $nu[2], \ldots, nu[p].$ *c*: c[i] contains the coefficient  $c_{(v)}$  of the partition whose binary model is in nu[i]; integer i, j, k, x, y; integer array lam[1:s+1,1:r], sigma[1:s+1,0:r]; p := 0; for i := 1 step 1 until s do lam[i+1,i] := lambda [i]; for j := s+1 step 1 until r do begin lam[s+1,j] := lambda[j]; sigma[s+1,j-1] := 0end; for i := 1 step 1 until s do sigma[i, r] := mu[i];k := mu[s] - lambda[r]; sigma[s, s-1] := 0;for j := r - 1 step -1 until s do begin  $sigma[s, j] := if k \ge 0$  then k else 0; k := sigma[s, j] - lambda[j] + lambda[j+1]end: i := s;build: for i := i step -1 until 1 do begin for j := i step 1 until r do lam[i, j] := lam[i+1, j] - sigma[i, j] + sigma[i, j-1];if  $i \neq 1$  then begin k := mu[i-1] - lam[i, r]; sigma[i-1, i-2] := 0;for j := r step -1 until i do begin  $sigma[i-1, j-1] := if k \ge sigma[i, j]$  then k else sigma[i, j]; k := sigma[i-1, j-1] - lam[i, j-1] + lam[i, j]end end end; x := j := 1;for j := j + 1 while (if j > r then false else lam[i,j] > 0) do  $x := x \times 2 \uparrow lam[1,j] + 1;$ if (if p = 0 then true else x > nu[p]) then begin p := p + 1; nu[p] := x; c[p] := 1end else if x = nu[p] then c[p] := c[p] + 1else begin j := 1; k := p;search:  $y := (j+k) \div 2$ ; if x = nu[y] then c[y] := c[y] + 1else if  $nu[y] < x \land x < nu[y+1]$  then begin for k := p step -1 until y + 1 do begin. c[k+1] := c[k]; nu[k+1] := nu[k]end: c[y+1] := 1; nu[y+1] := x; p := p + 1end else begin if x < nu[y] then k := y else j := y; go to search end end: for i := 1 step 1 until s do

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for y := i step 1 until r - 1 do
  if sigma[i,y] < sigma[i,y+1] then
  begin
    sigma[i,y] := sigma[i,y] + 1;
    for j := y step -1 until i do
    begin
      k := sigma[i,j] - lam[i+1,j] + lam[i+1,j+1];
      sigma[i,j-1] := if k > sigma[i+1,j] then k
      else sigma[i+1,j];
      if sigma[i, j-1] = 0 then
      begin
        for x := j - 1 step -1 until i do sigma[i,x-1] := 0;
        go to build
      end
    end
  end
end skew
```

# Algorithm 456 Routing Problem [H]

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The algorithm was originally developed as a part of vector ordering procedures at the Design Automation Center, RCA, Marlborough, Massachusetts, and was extended to general use in the traveling salesman and nonsymmetric routing problem.

Key Words and Phrases: routing problem, shortest path, traveling salesman problem, Hamiltonian circuit

CR Categories: 5.40 Language: Fortran

# Description

The algorithm finds the shortest serial (branchless) connection between n nodes of a net beginning in the start node sn and terminating in the end node en or terminating in any node. Also given is the  $m \times m$  matrix d of distances (with zero diagonal and not necessarily symmetric) between all pairs of nodes, and the vector p containing n node numbers to be connected referring to appropriate entries in the matrix d. The algorithm is constructed so that for one net (given by the matrix d) various connections, not necessarily exhausting all of m nodes, may be created; hence  $n \leq m$ . The case sn = en is also permitted, which actually yields a Hamiltonian circuit—traveling-salesman problem. If, in input,  $e_{1} = 0$ , the start-to-any connection is assumed. Also as an input is the number of runs r, which is discussed below. In the output, the original vector p is replaced by conjectured optimal sequence of n nodes, and l contains the connection length. The matrix d does not need to represent a Euclidean net nor be symmetric. Thus the algorithm may serve as a more general tool to solutions of related problems.

Since the method is heuristic, which implies it is approximate, guaranty of an optimal solution is based on empiric probability. The algorithm uses a tour-building method combined with tour-totour improvements.

In the first phase, the tour, or sequence of nodes, is built up by successively inserting not-yet-involved nodes into the tour.

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If, in the middle of tour building, the tour, for instance, consists of the nodes  $p_1, p_2, \ldots, p_k$ , the next node among the nodes  $p_{k+1}$ ,  $p_{k+2}$ , ...,  $p_n$ , and the arc (to be split by the chosen inserted node) among the arcs  $p_1p_2$ ,  $p_2p_3$ ,...,  $p_kp_1$ , are chosen so that the tour increment will be minimum; i.e.  $i (1 \le i \le k)$  and  $j(k < j \le n)$  are chosen in such a manner that  $d_{(p_i, p_j)} + d_{(p_j, p_{i+1})}$  $-d_{(p_i, p_{i+1})} = \min$ . Tour building starts with the arc  $p_1 p_1$  and terminates when all n nodes have been included. The tour-building approach of this kind for the traveling-salesman problem was originated by Karg and Thompson [1] and further developed by Raymond [2]. This algorithm, however, handles an open connection-start-to-end or start-to-any node. The maintenance of this property is ensured in the algorithm by assigning to the endto-start or each-to-start distance sufficiently large negative values  $(-n \times \max_{ij}[d_{ij}])$  which, in some way, firmly attach the end or any of *n* nodes to the start node permitting a circuit to form. In fact, the algorithm works on a net as if it were a closed circuit and keeps the node configuration by modifying the distance matrix. In output, the distance matrix is returned to its original form.

A tour thus built is hardly optimal and for larger nets it is probably far from optimum. The second phase improves the tour (for  $n \ge 3$ ) by the so-called 3-opt method proposed by Lin [3]. Improvements consist in exchanging three arcs, or links of the given connection by three other links. If there are no more 3 links to exchange for tour improvement, the tour is said to be 3-optimal. In general,  $\lambda$ -optimality can be considered. The implication of the 3-link exchange is essentially in reinsertions. Consecutive node chains of length k ( $1 \le k < n$ ) are successively tried to be reinserted (both as are and inverted) into remaining links for tour improvements, which actually represent 3-link exchanges (and also 2-link at the same time). A 3-opt tour shows a certain probability to be an optimal one in relation to n. Different 3-opt tours can be achieved if different initial nodes are chosen, which allows us to increase the probability of obtaining an optimal solution.

The algorithm can run r trials (as specified in input) with different initial nodes  $(p_1, set automatically)$ , thus obtaining different solutions while the best is saved and replaced in the vector p in output. For runs r > n ( $r \le 2n$ ) there is little chance for further improvement, because initial nodes repeat and the tour development can be affected only by previous contents of the vector p on which the tour is built. Probability that the 3-opt tour is optimal is somewhat higher in this algorithm, than in the one Lin suggests. In contrast to finding a 3-opt solution from a given random sequence of nodes, the fast building of an appropriate tour in the first phase considerably reduces the number of reinsertions in the second phase. The algorithm generalization to the noncyclic and nonsymmetric problems, in comparison to the traveling-salesman problem, increases computational time.

A considerable number of test examples have been run by the algorithm including the three problem types mentioned and the non-Euclidean and nonsymmetric problems. To outline the capability and how the "cost-approximation" factor r should be set for various n's, a survey of tested problems is presented, most of which problems have been solved and published before. The algorithm in Fortran was run on the RCA's SPECTRA 70/45 (fixed-point add time equals 8.88  $\mu$ sec), and is recommended for a high probability (over 95 percent) of obtaining an optimum if r = 2 to 5 for  $n \le 10$  and r = 5 to 15 for  $n \le 30$ . For higher n's, unless cost is out of consideration and r can be set up to 2n, the checking of successive results is advisable to see how improvements are developing (p and  $l_1$  should be checked after the tour-length calculation). These checks can also serve for getting suboptimal solutions.

In the program, the distance matrix d is in fixed-point mode, which makes computation faster and does not seem to be a serious restriction. Decimal order range of distances is expected to be small enough to be represented in fixed point, and calculations (additions and subtractions) will, most likely, not face overflow problem.

The arrays ID and Q should have the maximum subscript set at least to n.

Survey of tested problems

				Conject	ured		tı –
Ref.	n	sn	en	optimur	n	ropt	[sec]
Karg and Thomp-		1	2		118	1	
son [1]	5	1	0	en = 5	108	1	<1
		1	1		148	1	
Raymond [2]		1	5		165	1	
	7	1	0	en = 4	140	1	<1
		1	1		179	1	
Barachet [4]		1	2		350	1	
	10	1	0	en = 7	298	1	
		1	1		378	2	1.4
		1	2		308	1	
	10*	1	0	en = 7	257	2	
		1	1		336	2	
Author		1	2		102	1	
	12	1	0	en = 12	95	1	3.0
		1	1		114	1	
Author		1	6		117	1	
	13	1	0	en = 12	102	1	3.0
		1	1		130	1	
Held and Karp [5]		1	25	**	1517	10	21.8
	25	1	0	** en =25	1517	2	22.3
		1	1	**	1711	1	29.7
Karg and Thomp-		1	33	**	10655	2	53.6
son [1]	33	1	0	$\frac{**}{en} = 14$	10585	10	53 4
son [1]	55	î	1	**	10000		55.4

\* Nonsymmetric problem (two distances changed: (6, 5) = 1, and (8, 3) = 1). \*\* Results obtained from 10 runs.

The algorithm is believed to be applicable also to problems in which all connections do not necessarily exist. In terms of graph theory a graph representing the net to be routed need not be complete; i.e. every pair of vertices may be connected only in one of the two possible directions. The graph, however, must be strongly connected; i.e. there must be a path joining any pair of arbitrary distinct vertices. Nonexisting arcs might be expressed by assigning to the appropriate distances  $d_{kl}$  sufficiently large positive values, for instance  $n \times \max_{ij} [d_{ij}]$ .

### Symbol summary

- *n* number of nodes to be connected  $(2 \le n \le m)$ .
- *p* vector containing *n* node numbers (in output, it contains node number sequence of conjectured shortest path).
- sn start node number  $(1 \le sn \le m;$  no check is provided whether sn is contained in p).
- en end node number  $(1 \le en \le m; \text{ if } en = 0, \text{ start-to-any con$  $nection is assumed; } en = sn is allowed, which is traveling$ salesman problem; no check is provided whether sn is contained in <math>p)
- m order of distance matrix  $d \ (m \ge n \ge 2)$ .
- d  $m \times m$  matrix of distances of all node pairs (zero diagonal, not necessarily symmetric).
- *l* length of conjectured shortest path (output).
- r number of runs (trials;  $r \leq 2n$ ).
- ropt serial run number during which optimum has been achieved.
- $t_1$  average computational time of one run in seconds.

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40 IF (SN\_EGLEN. + 0R. EN. + 00.0 G0 T0 50
ID(1) = D(EN\_{\*}SN)
D(EN\_{\*}SN) = -LARGE
C RUN R TRIALS
50 L = LARGE
D0 280 IRS=1.R
C BUILD TOUR BY SUCCESSIVE INSERTING
C N0T-YET-INV0LVED N0DES
C INITIATE TOUR IS CONSIDERED AS
C ARC P(1) T0 P(1)
D0 90 JS=2.N
MININC = LARGE
C TRACE ALL N0T-YET-INV0LVED N0DES
C T0 CH00SE THE 0NE WITH MINIMUM INCREMENT
D0 70 J=JS.N
JP = P(J)
JE = JS - 1
C F0R EACH N0T-YET-INV0LVED N0DE TRACE ALREADY
C BUILT-UP TOUR T0 CH00SE THE MINIMUM INCREMENT ARC
D0 60 I=1.JE
IP = P(1)
IF (I.EQ.JE) IP = P(1) IF (I=E0.E) IP1 = P(1) INC = D(IP.JP) + D(JP.IP1) - D(IP.IP1) IF (INC.GE.MININC) GØ TØ 60 J1 = J JI = J II = I MININC = INC CONTINUE 60 CØNTINUE 70 CØNTINUE C STRETCH TOUR BY INSERTING THE CHØSEN NØDE P(J1) C BETWEEN THE NØDES P(I1) AND P(I1+1) 80 J1 = J1 - 1 IF (J1.EQ.I1) GØ TØ 90 IP = P(J1) P(J1) = P(J1+1) P(J1) = P(J1+1) GØ TØ 80 90 CØNTINUE C CØRRECT TOUR BY 3-ØPT METHØD C VARY CONSECUTIVE CHAIN LENGTH K 60 C VARY CONSECUTIVE CHAIN LENGTH K NI = N - 1 IF (N.LT.3) GØ TØ 210 IF (N.LT.3) G0 T0 210 D0 200 K=1.N1 IC0UNT = 0 C SHIFT C0NSECUTIVE CHAIN C THR0YGH0UT SEQUENCE 0F N N0DES 100 IC0R = 0 D0 190 J=1.N C CALCULATE CHAIN LENGTH IN F0RWARD C AND BACKWARD DIRECTION  $\begin{array}{l} \text{ARD DIRECTION} \\ \text{L1} = 0 \\ \text{LR} = 0 \\ \text{IF} (K \cdot EQ \cdot 1) & \text{G0 T0 120} \\ \text{I} = J \\ \text{K1} = 1 \\ \end{array}$  $IF (I \cdot GT \cdot N) I = I - N$  IP = P(I) IP1 = I + 1  $IF (IP1 \cdot GT \cdot N) IP1 = 1$ 110 IP1 = P(IP1)

L1 = L1 + D(IP, IP1) LR = LR + D(IP1, IP) I = I + 1 K1 = K1 + 1 IF (K1.LT.K) G0 T0 110 C FØR EACH POSITIONED CHAIN (AS IS AND INVERTED) C CHECK ALL ARCS IF INSERTION IMPROVES TOUR 120 MININC = LARGE J1 = J + K - 1 IF (J1.GT.N) J1 = J1 - N D0 150 I=1.N LF (J2.LE.J1 AND, (12.GE.J AND, IJ) IF (J.LE.JI .AND. (I.GE.J .AND. I.LE.JI)) G0 T0 IF (J.GT.JI .AND. (I.LE.JI .0R. I.GE.J)) G0 T0 150 \* IF (J.GT.J] .AND. (1.LE.J] .0R. I.GE.J)) G0 T0 15
IP = P(J)
JP = P(J)
JP1 = P(J)
IP1 = I + 1
IF (IP1.GT.N) IP1 = 1
JE = IP1
IF (IP1.E0.J) IP1 = J1 + 1
IF (IP1.GT.N) IP1 = 1
IP1 = P(IP1)
LN = L1
IR = 0
INC = D(IP.JP) + LN + D(JP1.IP1) - D(IP.IP1)
IF (INC.GT.MININC .0R. (INC.E0.MININC .AND.
(JE.NE.J .0R. JE.E0.J .AND. IR.E0.1))) G0 T0 140
I1 = I 130 \* (JE.NE.J.0R. JE.EQ.J.II = IIR1 = IRMININC = INCIF (IR.EQ.1) 60 T0 150IR = ILN = LRJS = JPJP = JPIPI = IS140 JP = JPI JP1 = JS G0 T0 130 I50 CONTINUE I = II + 1 IF (I.GT.N) I = 1 IF (I.EG.J .AND. IR1.E0.0) G0 T0 190 C REINSERT CHAIN OF LENGTH K STARTING IN J C BETWEEN NØDES P(I) AND P(II+1) ICCR = 1 JS = J JE = 0 IF (IE1.F0.0) G0 T0 160IF (IR1.EQ.0) GØ TØ 160 JS = J1 JE = -1 KI = 0 KI = KI + 1IF (K1.GT.K) G0 T0 190 I = JS160 170 JS = JS + JEIF (JS.LT.1) JS = N IP = I + 1 180 IP = I + 1 IF (IP.GT.N) IP = 1  $\begin{array}{l} IF & (IF+6I+N) & IF = I \\ JF = F(I) \\ F(IP) = JF \\ I = I + J \\ IF & (I+6T+N) & I = I \\ IF & (I+6T+N) & I = I \\ IF & (I-F-II) & I80, & I70, & 180 \\ CGNTINUE \\ IF & (IF0E) & FO = 0. & GA & IC & 200 \\ \end{array}$ 190 CONTINUE IF (ICOR-E0-0) G0 TC 200 ICOUNT = ICOUNT + 1 IF (ICOUNT-LT-N) GC TO 100 CONTINUE 200 CONTINUE C ORIENT TOUR WITH SN IN P(1) 210 DO 230 I=1.N IF (P(1).E0.SN) GO TO 240 JS = P(1) DO 220 J=1.N1 P(J) = P(J+1) 220 CONTINUE P(N) = JS 200 P(N) = JS230 CONTINUE 230 CONTINUE C CALCULATE TOUR LENGTH 240 L1 = 0 DC 250 I=1,N1 IP = P(I) IP1 = P(I+1) L1 = L1 + D(IP,IP1) 250 CONTINUE 250 CONTINUE 250 CONTINUE IP = P(1) IF (SN.E0.EN) L1 = L1 + D(IP1,IP) C SAVE SOLUTION, IF BETTER, AND SET NEW INITIATE NODE IF (L1.60.L) GO TC 270 L = L1 DO 260 I=1,N 0(1) = P(1) CONTINUE J = IKS + 1IF (J.GT.N) J = J - N260 270  $\begin{array}{c} \text{IF } (J, \text{GT,N}) \; J = J - N \\ JS = r(1) \\ P(1) = P(J) \\ P(1) = JS \\ 280 \; \text{COVELLUS} \\ 280 \; \text{COVELLUS} \\ 06 \; 290 \; \text{IELS} \\ 06 \; 290 \; \text{IELS} \\ P(I) = v(I) \\ 290 \; \text{COMTINUS} \\ IEL \; (EA+42.0) \; \text{GOLO 310} \\ D0 \; 300 \; \text{IELS} \\ 06 \; \text{ISSN} = ID(I) \\ 300 \; \text{COVELUS} \\ \end{array}$ 300 CONTINUE 310 IF (SN.ED.EN .GR. EN.EU.D) GO TO 320 D(EN, SN) = ID(1) 320 AETUAN END

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# Algorithm 457

# Finding All Cliques of an Undirected Graph [H]

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CR Categories: 3.71, 5.32 Language: Algol

### Description

*Introduction.* A maximal complete subgraph (clique) is a complete subgraph that is not contained in any other complete subgraph.

A recent paper [1] describes a number of techniques to find maximal complete subgraphs of a given undirected graph. In this paper, we present two backtracking algorithms, using a branchand-bound technique [4] to cut off branches that cannot lead to a clique.

The first version is a straightforward implementation of the basic algorithm. It is mainly presented to illustrate the method used. This version generates cliques in alphabetic (lexicographic) order.

The second version is derived from the first and generates cliques in a rather unpredictable order in an attempt to minimize the number of branches to be traversed. This version tends to produce the larger cliques first and to generate sequentially cliques having a large common intersection. The detailed algorithm for version 2 is presented here.

Description of the algorithm—Version I. Three sets play an important role in the algorithm. (1) The set compsub is the set to be extended by a new point or shrunk by one point on traveling along a branch of the backtracking tree. The points that are eligible to extend compsub, i.e. that are connected to all points in compsub, are collected recursively in the remaining two sets. (2) The set candidates is the set of all points that will in due time serve as an extension to the present configuration of compsub. (3) The set not is the set of all points that have at an earlier stage already served as an extension of the present configuration of compsub and are now explicitly excluded. The reason for maintaining this set not will soon be made clear.

The core of the algorithm consists of a recursively defined extension operator that will be applied to the three sets just described. It has the duty to generate all extensions of the given configuration of *compsub* that it can make with the given set of candidates and that do not contain any of the points in *not*. To put it differently: all extensions of *compsub* containing any point in *not* have already been generated. The basic mechanism now consists of the following five steps:

Step 1. Selection of a candidate.

Step 2. Adding the selected candidate to compsub.

Step 3. Creating new sets candidates and not from the old sets by

removing all points not connected to the selected candidate (to remain consistent with the definition), keeping the old sets in tact.

Step 4. Calling the extension operator to operate on the sets just formed.

Step 5. Upon return, removal of the selected candidate from *compsub* and its addition to the old set *not*.

We will now motivate the extra labor involved in maintaining the sets *not*. A necessary condition for having created a clique is that the set *candidates* be empty; otherwise *compsub* could still be extended. This condition, however, is not sufficient, because if now *not* is nonempty, we know from the definition of *not* that the present configuration of *compsub* has already been contained in another configuration and is therefore not maximal. We may now state that *compsub* is a clique as soon as both *not* and *candidates* are empty.

If at some stage *not* contains a point connected to all points in *candidates*, we can predict that further extensions (further selection of candidates) will never lead to the removal (in Step 3) of that particular point from subsequent configurations of *not* and, therefore, not to a clique. This is the branch and bound method which enables us to detect in an early stage branches of the backtracking tree that do not lead to successful endpoints.

A few more remarks about the implementation of the algorithm seem in place. The set *compsub* behaves like a stack and can be maintained and updated in the form of a global array. The sets *candidates* and *not* are handed to the extensions operator as a parameter. The operator then declares a local array, in which the new sets are built up, that will be handed to the inner call. Both sets are stored in a single one-dimensional array with the following layout:

### not candidates

index values: 1.....ne.......ce....

The following properties obviously hold:

2. ne = ce:empty (candidates)

3. ne = 0 :empty (not)

4. ce = 0 :empty (not) and empty (candidates) = clique found

If the selected candidate is in array position ne + 1, then the second part of Step 5 is implemented as ne := ne + 1.

In version 1 we use element ne + 1 as selected candidate. This strategy never gives rise to internal shuffling, and thus all cliques are generated in a lexicographic ordering according to the initial ordering of the candidates (all points) in the outer call.

For an implementation of version 1 we refer to [3].

Description of the algorithm—Version 2. This version does not select the candidate in position ne + 1, but a well-chosen candidate from position, say s. In order to be able to complete Step 5 as simply as described above, elements s and ne + 1 will be interchanged as soon as selection has taken place. This interchange does not affect the set *candidates* since there is not implicit ordering. The selection does affect, however, the order in which the cliques are eventually generated.

Now what do we mean by "well chosen"? The object we have in mind is to minimize the number of repetitions of Steps 1-5 inside the extension operator. The repetitions terminate as soon as the bound condition is reached. We recall that this condition is formulated as: there exists a point in *not* connected to all points in *candidates*. We would like the existence of such a point to come about at the earliest possible stage.

Let us assume that with every point in *not* is associated a counter, counting the number of candidates that this point is not connected to (*number of disconnections*). Moving a selected candidate into *not* (this occurs after extension) decreases by one all counters of the points in *not* to which it is disconnected and introduces a new counter of its own. Note that no counter is ever

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Fig. 1. Random graphs show the computing time per clique (in ms) versus dimension of the graph (in brackets: total number of cliques in the test sample).



decreased by more than one at any one instant. Whenever a counter goes to zero the bound condition has been reached.

Now let us fix one particular point in *not*. If we keep selecting candidates disconnected to this fixed point, the counter of the fixed point will be decreased by one at every repetition. No other counter can go down more rapidly. If, to begin with, the fixed point has the lowest counter, no other counter can reach zero sooner, as long as the counters for points newly added to *not* cannot be smaller. We see to this requirement upon entry into the extension operator, where the fixed point is taken either from *not* or from the original *candidates*, whichever point yields the lowest counter value after the first addition to *not*. From that moment on we only keep track of this one counter, decreasing it for every next selection, since we will only select disconnected points.

The Algol 60 implementation of this version is given below. Discussion of comparative tests. Augustson and Minker [1] have evaluated a number of clique finding techniques and report an algorithm by Bierstone [2] as being the most efficient one.

In order to evaluate the performance of the new algorithms, we implemented the Bierstone algorithm<sup>1</sup> and ran the three algorithms on two rather different testcases under the Algol system for the EL-X8.

For our first testcase we considered random graphs ranging in dimension from 10 to 50 nodes. For each dimension we generated a collection of graphs where the percentage of edges took on the following values: 10, 30, 50, 70, 90, 95. The cpu time per clique for each dimension was averaged over such a collection. The results are graphically represented in Figure 1.

The detailed figures [3] showed the Bierstone algorithm to be of slight advantage in the case of small graphs containing a small number of relatively large cliques. The most striking feature, however, appears to be that the time/clique for version 2 is hardly dependent on the size of the graph. Fig. 2. Moon-Moser graphs show the computing time (in ms) versus k. Dimension of the graph = 3k. Plotted on logarithmic scale.



The difference between version 1 and "Bierstone" is not so striking and may be due to the particular Algol implementation. It should be borne in mind that the sets of nodes as they appear in the Bierstone algorithm were coded as one-word binary vectors, and that a sudden increase in processing time will take place when the input graph is too large for "one-word representation" of its subgraphs.

The second testcase was suggested by the referee and consisted of regular graphs of dimensions  $3 \times k$ . These graphs are constructed as the complement of k disjoint 3-cliques. Such graphs contain  $3^k$  cliques and are proved by Moon and Moser [5] to contain the largest number of cliques per node.

In Figure 2 a logarithmic plot of computing time versus k is presented. We see that both version 1 and version 2 perform significantly better than Bierstone's algorithm. The processing time for version 1 is proportional to  $4^k$ , and for version 2 it is proportional to  $(3.14)^k$  where  $3^k$  is the theoretical limit.

Another aspect to be taken into account when comparing algorithms is their storage requirements. The new algorithms presented in this paper will need at most  $\frac{1}{2}M(M+3)$  storage locations to contain arrays of (small) integers where M is the size of largest connected component in the input graph. In practice this limit will only be approached if the input graph is an almost complete graph. The Bierstone algorithm requires a rather unpredictable amount of store, dependent on the number of cliques that will be generated. This number may be quite large, even for moderate dimensions, as the Moon-Moser graphs show.

Finally it should be pointed out that Bierstone's algorithm does not report isolated points as cliques, whereas the new algorithm does. Either algorithm can, however, be modified to produce results equivalent to the other. Suppression of 1-cliques in the new algorithm is the simplest adaption.

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<sup>&</sup>lt;sup>1</sup> Bierstone's algorithm as reported in [1] contained an error. In our implementation the error was corrected. The error was independently found by Mulligan and Corneil at the University of Toronto, and reported in [6].

## References 1. Augustson, J.G., and Minker, J. An analysis of some graph

theoretical cluster techniques, J. ACM 17 (1970), 571-588. 2. Bierstone, E. Unpublished report. U of Toronto. 3. Bron, C., Kerbosch, J.A.G.M., and Schell, H.J. Finding cliques in an undirected graph. Tech. Rep. Technological U. of Eindhoven, The Netherlands. 4. Little, John D.C., et al. An algorithm for the traveling salesman problem. Oper. Res. 11 (1963), 972-989. 5. Moon, J.W., and Moser, L. On cliques in graphs. Israel J. Math. 3 (1965), 23-28. 6. Mulligan, G.D., and Corneil, D.G. Corrections to Bierstone's algorithm for generating cliques. J. ACM 19 (Apr. 1972), 244-247. Algorithm procedure output maximal complete subgraphs 2(connected, N);value N; integer N; Boolean array connected; comment The input graph is expected in the form of a symmetrical Boolean matrix connected. N is the number of nodes in the graph. The values of the diagonal elements should be true; begin integer array ALL, compsub[1:N]; integer c; procedure extend version 2(old, ne, ce); value ne, ce; integer ne, ce; integer array old; begin integer array new[1 : ce]; integer nod, fixp; integer newne, newce, i, j, count, pos, p, s, sel, minnod; comment The latter set of integers is local in scope but need not be declared recursively;  $minnod := ce; \quad i := nod := 0;$ DETERMINE EACH COUNTER VALUE AND LOOK FOR MINIMUM: for i := i + 1 while  $i \leq ce \land minnod \neq 0$  do begin p := old[i]; count := 0;j := ne;**COUNT DISCONNECTIONS:** for j := j + 1 while  $j \le ce \land count < minnod$  do if  $\neg$  connected[p, old[j]] then begin count := count + 1;SAVE POSITION OF POTENTIAL CANDIDATE: pos := jend: **TEST NEW MINIMUM:** if count < minnod then begin fixp := p; minnod := count;if  $i \leq ne$  then s := poselse begin s := i; *PREINCR*: nod := 1 end end NEW MINIMUM; end i: comment If fixed point initially chosen from candidates then number of disconnections will be preincreased by one; BACKTRACKCYCLE: for nod := minnod + nod step -1 until 1 do begin INTERCHANGE: p := old[s]; old[s] := old[ne + 1];sel := old[ne + 1] := p;FILL NEW SET not: *newne* := i := 0; for i := i + 1 while  $i \le ne$  do if connected[sel, old[i]] then

**begin** newne := newne + 1; new[newne] := old[i] end; FILL NEW SET cand: *newce* := *newne*; i := ne + 1; for i := i + 1 while  $i \leq ce$  do if connected[sel, old[i]] then **begin** newce := newce + 1; new[newce] := old[i] end; ADD TO compsub: c := c + 1; compsub[c] := sel;if newce = 0 then begin integer loc; outstring(1, `clique = ');for loc := 1 step 1 until c do outinteger(1, compsub[loc]) end output of clique else if newne < newce then extend version 2(new, newne, newce); **REMOVE FROM compsub:** c := c - 1;ADD TO not: ne := ne + 1;if nod > 1 then begin SELECT A CANDIDATE DISCONNECTED TO THE FIXED POINT: s := ne: LOOK: FOR CANDIDATE: s := s + 1;if connected[fixp, old[s]] then go to LOOK end selection end BACKTRACKCYCLE end extend version 2; for c := 1 step 1 until N do ALL[c] := c; c := 0; extend version 2(ALL, 0, N)

end output maximal complete subgraphs 2;

Remark on Algorithm 323 [G6] Generation of Permutations in Lexicographic Order [R.J. Ord-Smith, Comm. ACM 11 (Feb. 1968), 117]

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In presenting Algorithm 323, *BESTLEX*, for generating permutations in lexicographic order, the author has mentioned the number of transpositions. It may be remarked here that equal numbers of transpositions are required by both *BESTLEX* and the previously fastest algorithm, Algorithm 202 [1]. The exact number of transpositions  $(T_n)$  necessary to generate the complete set of *n*! permutations is given by

$$T_n = n! (\psi_{n-1}) - (n+1)/2, \text{ if } n \text{ is odd, and} T_n = n! (\psi_{n-2}) - n/2, \text{ if } n \text{ is even,} where  $\psi_{2n} = 1 + \frac{1}{2!} + \frac{1}{4!} + \dots + \frac{1}{(2n)!} \doteq 1.543 \text{ for } n \ge 3.$$$

The above expressions do not include the few extra transpositions (equal to the integral part of n/2) required by *BESTLEX* to generate the initial arrangement from the final one, as this portion has

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not been included in Algorithm 202. Therefore, the number of transpositions has no importance in the context of the claim that *BESTLEX* is more than twice as fast as Algorithm 202.

The main factor contributing to the speed of *BESTLEX* is the substantial reduction in the number of comparisons required, by the introduction of the **own integer array** q. Taking into account only those comparisons which involve array elements, the number of comparisons ( $C_n$ ) required to generate all the n! permutations can be shown to be equal to

$$C_n \text{ (Algorithm 202)} = \frac{n!}{2} [1 + 3\varphi_{n-2}] + n,$$
  

$$C_n (BESTLEX) = n! \left[ \frac{1}{2} + \varphi_{n-1} \right],$$
  
where  $\varphi_n = 1 + \frac{1}{2!} + \frac{1}{3!} + \dots + \frac{1}{n!} \doteq 1.718 \text{ for } n \ge 6.$ 

... 1

This shows that the number of comparisons required by *BESTLEX* is lower by .859(n!) (approximately) in the case of the generation of all the n! arrangements.

Finally, a modification of the *BESTLEX* algorithm is suggested which will reduce the number of comparisons again by (n!)/2. The modification involves replacement of lines 2–14 of Algorithm 323 by the following.

begin own integer array q[3:n]; integer k, m; real t; own Boolean flag;

**comment** Own dynamic arrays are not often implemented. The upper bound will have to be given explicitly;

if first then

begin first := false; flag := true for m := 3 step 1 until n do q[m] := 1

end of initialization process;

if *flag* then

begin flag := false; t := x[1]; x[1] := x[2]; x[2] := t;

$$x_1 = x_1(1), x_1(1) = x_1(2), x_1(2),$$
  
go to finish  
end;  
flag := true;

for k := 3 step 1 until n do

### References

1. Shen, Mok-Kong. Algorithm 202, generation of permutations in lexicographical order. *Comm. ACM* 6 (Sept. 1963), 517.

Added in proof: An improved version of *BESTLEX*, viz. Algorithm 323A, Generation of Permutation Sequences: Part 2, by R.J. Ord-Smith [*Comp. J. 14*, 2 (May 1971), 136–139], which also incorporates the modification suggested here, has come to the author's attention.

# Remark on Algorithm 408 [F4]

A Sparse Matrix Package (Part I)

[John Michael McNamee, Comm. ACM 14 (Apr. 1971), 265-273]

E.E. Lawrence [Recd. 1 February 1972, 12 March 1973] Central Application Laboratory, Mullard Limited, New Road, Mitcham, Surrey CR4 4XY, England

The subroutines constituting Algorithm 408 were, with the exception of MVSPMX and WRSPMX, tested on an IBM 360/65 using CALL/360-0S. The author's alteration (iii) was introduced, i.e. declaration of the *M*-array to be half length. Other changes were introduced in order: (a) to make the algorithm more conversational in a time shared environment; and (b) to improve the speed of the sorting procedure in *PERCOL*.

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The following deficiencies in the algorithm were noted 1. The dimensional parameters of ACSPMX, ADSPMX, and MUSPMX are incomplete. As an illustration of this consider the two matrices

1	1	0	0	07	
	0	2	0	0	
A =	0	0	3	0	
	_0	0	0	4	
	Γ0	0	0	47	
D	Г0 0	0 0	0 3	4 0	
<i>B</i> =	0 0 0	0 0 2	0 3 0	4 0 0	

each of which has four nonzero elements.

Then the sum matrix has eight such elements, and in general, for two matrices with  $n_1$  and  $n_2$  nonzero elements, the number of nonzero elements,  $n_3$ , in the sum matrix is in the range  $0 \le n_3 \le n_1 + n_2$ .

However in ADSPMX the condition used is  $n_1 = n_2 = n_3$ .

Similar arguments apply to ACSPMX and MUSPMX.

To correct this requires extensions to the parameter lists and dimension statements, and also it changes the conditional statements within the subroutines concerned.

This shows up with the CALL/360-0S system since the compiler performs subscript checking. It would not be evident on most compilers including the IBM Fortran IV G compiler. It is, however, bad practice to rely on default effects of compilers.

2. There are three, probably copying, errors in *MUSPMX* (page 270).

- (i) Line 33 should be: IF(NCA.EQ.NCB) GO TO 3
- (ii) Line 55 should be: DO 14 J = 1, NRB
- (iii) Line 102 should be: CALL IPK(NRB,MC,2,NM)

# Remark on Algorithm 420 [J6]

Hidden-Line Plotting Program [Hugh Williamson, Comm. ACM 15 (Feb. 1972), 100-103]

Hugh Williamson [Recd. 9 Oct. 1972] National Con-Serv, Incorporated, Austin, Texas

The input quantities to subroutine HIDE referred to in the following paragraphs (e.g. N1, NFNS, "input curve to be plotted") are described in the block of comment statements at the beginning of HIDE as originally published.

If N1 < 0, DO loop 71 is not executed properly, since the upper limit, N1, is less than the lower limit, 2. This affects only checking for monotonicity in the input abscissa array; otherwise, if the inputs are correct, the performance of the program is not affected.

The error is corrected if the first 11 executable statements are replaced by the following (the first executable statement of the original program, which is not changed, is listed for clarity):

IF(MAXDIM.LE.0) RETURN IFPLOT = 1 IF(N1.GT.0) GO TO 76 N1 = -N1IFPLOT = 0

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76 DO 71 I = 2,N1  
IF(
$$X(I-1)$$
.LT. $X(I)$ ) GO TO 71  
MAXDIM = 0  
GO TO 75  
71 CONTINUE  
IF(NG.GT.0) GO TO 5000

On computers in which all variables are not automatically set to zero before execution, FNSM1 is not properly initialized if  $NFNS \leq 0$ . To correct this, simply insert the statement

### FNSM1 = 0.

before the statement

#### IF(NFNS.LE.0) GO TO 46

The latter is the sixth statement after Fortran statement number 74.

*FNSM*1 will still be improperly defined if NFNS = 1. If only one curve is to be plotted, however, translating to simulate stepping in the depth dimension will not be done, so set NFNS = -1 for only one curve to be plotted.

In some cases, the three-dimensional surface is easier to visualize if (nearly) vertical lines are drawn at the left edge of each curve; this effect is illustrated by Figures 1 and 2. The verticals are added by inserting (XMIN- $\epsilon$ , YMIN) as the first point in each input curve to be plotted, where  $\epsilon$  is a small positive number (10<sup>-4</sup>×DELTAX would be appropriate).

The author appreciates very much the comments received form readers of Communications regarding implementation of *HIDE* on different computers.

### Remark on Algorithm 429 [C2]

Localization of the Roots of a Polynomial [C2] [W. Squire, *Comm. ACM 15* (Aug. 1972), 776–777]

H.B. Driessen and E.W. LeM. Hunt [Recd. 13 Oct. 1972, 29 Jan. 1973]

Supreme Headquarters Allied Powers of Europe, Technical Center, P.O. Box 174, The Hague, The Netherlands

There seems to be an error in this algorithm. If we take the polynomial:

 $z^4 + a_2 z^2 + a_3 z^3 + a_4 z + a_5 = 0,$ 

then after the second pass through the K-loop of the logical function HRWTZR(C, N), the term  $(a_2a_3-a_4)a_4 - a_5a_2$  is tested for a minus sign. However, the term which should be tested according to the Routh-Hurwitz criterion is  $(a_2a_3-a_4)a_4 - a_5a_2^2$ . If this term is negative then there are no roots with positive real parts.

As an example, if the polynomial

 $z^4 + 5.6562 \, z^3 + 5.8854 \, z^2 + 7.3646 \, z + 6.1354 = 0$ 

is studied with the help of Algorithm 429 one will find as output:

Roots are in an annulus of inner radius .454 E + 00 and outer radius .836 E + 01;

There are no real positive roots;

The negative roots (if any) are between -.454 E + 00 and -.836 E + 01;

There are no roots with positive real parts.

However, if one calculates the roots of this equation, one will find approximately:

- $z_1 = -1.0001$
- $z_2 = -4.7741$
- $z_{3.4} = +0.0089 \pm 1.1457 i$

Statement 20 + 1 in the logical function HRWTZR(C,N), which was originally "C1 = C(1)", should be amended to read "C1 = C(1)/C1".

As a by-product of our investigation, it turns out that the structure of the logical function HRWTZR can be simplified by abandoning the logically redundant steps C(K) = C(K+1).

The following listing incorporates both the correction and the simplifications. The function has been parameter tested on a CDC-6400.

LOGICAL FUNCTION HRWTZR (C,N) DIMENSION C(N) HRWTZR = .FALSE. IF (C(1) .LE.0..OR.C(N).LE.0.) RETURN C1 = C(1) M = N - 1DO 30 I = 2,M DO 20 K = I,M,2 20 C(K) = C(K) - C(K+1)/C1 C1 = C(I)/C1 IF (C1.LE.0.) RETURN 30 CONTINUE HRWTZR = .TRUE.

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RETURN

END