Munkres' Assignment Algorithm
Modified for Rectangular Matrices

Assignment Problem - Let *C* be an *nxn* matrix representing the costs of each of *n* workers to perform any of *n* jobs. The assignment problem is to assign jobs to workers so as to minimize the total cost. Since each worker can perform only one job and each job can be assigned to only one worker the assignments constitute an *independent set* of the matrix *C*.

$$C(i,j) = \begin{array}{c c} & \begin{array}{c c c c} p & q & r & s \end{array} \\ \begin{array}{c} a \\ b \\ c \\ d \end{array} & \left[\begin{array}{c c c c} 1 & 2 & 3 & 4 \\ 2 & 4 & 6 & 8 \\ 3 & 6 & 9 & 12 \\ 4 & 8 & 12 & 16 \end{array} \right] \end{array}$$

```
Workers = {a,b,c,d}
   Jobs = {p,q,r,s}

An arbitrary assignment
  A = { (a,q), (b,s), (c,r), (d,p) }

Total cost = 23
```

An arbitrary assignment is shown above in which worker *a* is assigned job *q*, worker *b* is assigned job *s* and so on. The total cost of this assignment is *23*. Can you find a lower cost assignment? Can you find the minimal cost assignment? Remember that each assignment must be unique in its row and column.

A brute-force algorithm for solving the assignment problem involves generating all independent sets of the matrix *C*, computing the total costs of each assignment and a search of all assignment to find a minimal-sum independent set. The complexity of this method is driven by the number of independent assignments possible in an *nxn* matrix. There are *n* choices for the first assignment, *n-1* choices for the second assignment and so on, giving n! possible assignment sets. Therefore, this approach has, at least, an exponential runtime complexity.
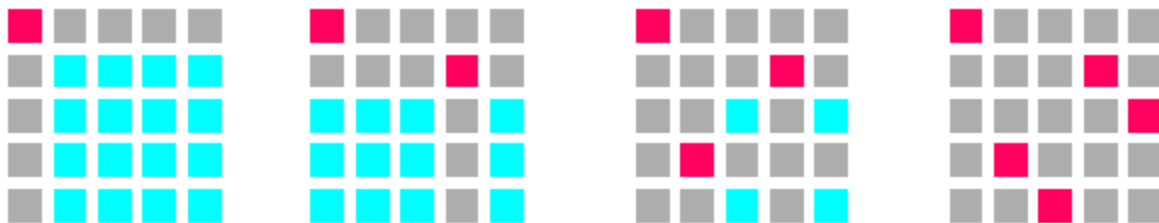
As each assignment is chosen that row and column are eliminated from consideration. The question is raised as to whether there is a better algorithm. In fact there exists a polynomial runtime complexity algorithm for solving the assignment problem developed by James Munkre's in the late 1950's despite the fact that some references still describe this as a problem of exponential complexity.

The following 6-step algorithm is a modified form of the original Munkres' Assignment Algorithm (sometimes referred to as the Hungarian Algorithm). This algorithm describes to the manual manipulation of a two-dimensional matrix by starring and priming zeros and by covering and uncovering rows and columns. This is because, at the time of publication (1957), few people had access to a computer and the algorithm was exercised by hand.

> **Step 0:** Create an nxm matrix called the cost matrix in which each element represents the cost of assigning one of n workers to one of m jobs. Rotate the matrix so that there are at least as many columns as rows and let k=min(n,m).
>
> **Step 1:** For each row of the matrix, find the smallest element and subtract it from every element in its row. Go to Step 2.
>
> **Step 2:** Find a zero (Z) in the resulting matrix. If there is no starred zero in its row or column, star Z. Repeat for each element in the matrix. Go to Step 3.
>
> **Step 3:** Cover each column containing a starred zero. If K columns are covered, the starred zeros describe a complete set of unique assignments. In this case, Go to DONE, otherwise, Go to Step 4.
>
> **Step 4:** Find a noncovered zero and prime it. If there is no starred zero in the row containing this primed zero, Go to Step 5. Otherwise, cover this row and uncover the column containing the starred zero. Continue in this manner until there are no uncovered zeros left. Save the smallest uncovered value and Go to Step 6.

Step 5: Construct a series of alternating primed and starred zeros as follows. Let Z_0 represent the uncovered primed zero found in Step 4. Let Z_1 denote the starred zero in the column of Z_0 (if any). Let Z_2 denote the primed zero in the row of Z_1 (there will always be one). Continue until the series terminates at a primed zero that has no starred zero in its column. Unstar each starred zero of the series, star each primed zero of the series, erase all primes and uncover every line in the matrix. Return to Step 3.

Step 6: Add the value found in Step 4 to every element of each covered row, and subtract it from every element of each uncovered column. Return to Step 4 without altering any stars, primes, or covered lines.

DONE: Assignment pairs are indicated by the positions of the starred zeros in the cost matrix. If C(i,j) is a starred zero, then the element associated with row i is assigned to the element associated with column j.

Some of these descriptions require careful interpretation. In Step 4, for example, the possible situations are, that there is a noncovered zero which get primed and if there is no starred zero in its row the program goes onto Step 5. The other possible way out of Step 4 is that there are no noncovered zeros at all, in which case the program goes to Step 6.

At first it may seem that the erratic nature of this algorithm would make its implementation difficult. However, we can apply a few general rules of programming style to simplify this problem. The same rules can be applied to any step-algorithm.

Good Programming Style and Design Practices

1. Strive to create readable source code through the use of blank lines, comments and spacing.

2. Use consistent naming conventions, for variable and constant identifiers and subprograms.

3. Use consistent indentation and always indent the bodies of conditionals and looping constructs.

4. Place logically distinct computations in their own execution blocks or in separate subprograms.

5. Don't use global variables inside subprograms except where such use is clear and improves readability and efficiency.

6. Use local variables where appropriate and try to limit the creation of unnecessary identifiers in the main program.

7. Open I/O files only when needed and close them as soon as they are no longer required.

8. Work to keep the level of nesting of conditionals and loops at a minimum.

9. Use constant identifiers instead of hardwiring for-loop and array ranges in the body of the code with literal values.

10. When you feel that things are getting out of control, start over. Re-coding is good coding.

By applying Rule 4 to the step-algorithm we decide to make each step its own procedure. Now we can apply Rule 8 by using a case statement in a loop to control the ordering of step execution.

A old implementation of Munkres - This implementation is

```
procedure munkres is
  n : constant integer := 20;              -- num rows/columns (this version
is hard-wired for square matrices)
  C : is array(1..n,1..n) of float;        -- cost matrix
  M : is array(1..n,1..n) of integer;      -- a mask matrix to indicate primed
(= 2) and starred (=1) zeros in C
  Row,Col : is array(1..n) of integer;     -- maintains record of which
```

```ada
      row/columns are covered.
    stepnum : integer;                              -- covered = 1, non-covered = 0
    done : boolean;

    function step1(stepnum: in out integer) is
      :
    function step2(stepnum: in out integer) is
      :
    function step3(stepnum: in out integer) is
      :

  begin
    done:=false;
    stepnum:=1;
    while not(done) loop
      case stepnum is
        when 1 => step1(stepnum);
        when 2 => step2(stepnum);
        when 3 => step3(stepnum);
        when 4 => step4(stepnum);
        when 5 => step5(stepnum);
        when 6 => step6(stepnum);
        when others => done:=true;
      end case;
    end loop;
  end munkres;
```

In each pass of the loop the step procedure called sets the value of stepnum for the next pass. When the algorithm is finished the value of stepnum is set to some value outside the range 1..6 so that done will be set to true and the program will end. In the completed program the tagged (starred) zeros flag the row/column pairs that have been assigned to each other. We will discuss the implementation of a procedure for each step of Munkres' Algorithm below. We will assume that the cost matrix $C(i,j)$ has already been loaded with the first index referring to the row number and the second index referring to the column number.

Step 1

For each row of the matrix, find the smallest element and subtract it from every element in its row. Go to Step 2. We can define a local variable called *minval* that is used to hold the smallest value in a row. Notice that there are two loops over the index *j* appearing inside an outer loop over the index *i*. The first inner loop over the index *j* searches a row for the *minval*. Once *minval* has been found this value is subtracted from each element of that row in the second inner loop over *j*. The value of step is set to *2* just before *stepone* ends.

```ada
 procedure stepone(step : in out integer) is
    minval : integer;
  begin
    for i in 1..n loop

      minval:=C(i,1);
      for j in 2..n loop
        if minval>C(i,j) then
          minval:=C(i,j);
        end if;
      end loop;

      for j in 1..n loop
        C(i,j):=C(i,j)-minval;
      end loop;

    end loop;
    step:=2;
  end stepone;
```

Step 2

Find a zero (Z) in the resulting matrix. If there is no starred zero in its row or column, star Z. Repeat for each element in the matrix. Go to Step 3. In this step, we introduce the mask matrix M, which in the same dimensions as the cost matrix and is used to star and prime zeros of the cost matrix. If $M(i,j)=1$ then $C(i,j)$ is a starred zero, If $M(i,j)=2$ then $C(i,j)$ is a primed zero. We also define two vectors R_cov and C_cov that are used to "cover" the rows and columns of the cost matrix C. In the nested loop (over indices i and j) we check to see if $C(i,j)$ is a zero value and if its column or row is not already covered. If not then we star this zero (i.e. set $M(i,j)=1$) and cover its row and column (i.e. set $R\_cov(i)=1$ and $C\_cov(j)=1$). Before we go on to Step 3, we uncover all rows and columns so that we can use the cover vectors to help us count the number of starred zeros.

```
procedure steptwo(step: in out integer) is
begin

   for i in 1..n loop
     for j in 1..n loop
       if C(i,j)=0 and C_cov(j)=0 and R_cov(i)=0 then
          M(i,j):=1;
          C_cov(j):=1;
          R_cov(i):=1;
       end if;
     end loop;
   end loop;

   for i in 1..n loop
     C_cov(i):=0;
     R_cov(i):=0;
   end loop;
   step:=3;

end steptwo;
```

Step 3

Cover each column containing a starred zero. If K columns are covered, the starred zeros describe a complete set of unique assignments. In this case, Go to DONE, otherwise, Go to Step 4. Once we have searched the entire cost matrix, we count the number of independent zeros found. If we have found (and starred) K independent zeros then we are done. If not we procede to Step 4.

```
procedure stepthree(step : in out integer) is
   count : integer;
begin
   for i in 1..n loop
     for j in 1..n loop
       if M(i,j)=1 then
          C_cov(j):=1;
       end if;
     end loop;
   end loop;
   count:=0;
   for j in 1..n loop
     count:=count + C_cov(j);
   end loop;
   if count>=n then
     step:=7;
   else
     step:=4;
   end if;
end stepthree;
```

Step 4

Find a noncovered zero and prime it. If there is no starred zero in the row containing this primed zero, Go to Step 5. Otherwise, cover this row and uncover the column containing the starred zero. Continue in this manner until there are no uncovered zeros left. Save the smallest uncovered value and Go to Step 6. In this step, statements such as "find a noncovered zero" are clearly distinct operations that deserve their own functional blocks. We have decomposed this step into a main procedure and three subprograms (2 procedures and a boolean function).

```
procedure stepfour(step : in out integer) is
   row,col  : integer;
   done     : boolean;

   procedure find_a_zero(row,col : out integer) is
      i,j : integer;
      done: boolean;
   begin
      row:=0;
      col:=0;
      i:=1;
      done:=false;
      loop
        j:=1;
        loop
         if C(i,j)=0 and R_cov(i)=0 and C_cov(j)=0 then
            row:=i;
            col:=j;
            done:=true;
         end if;
          j:=j+1;
          exit when j>n;
        end loop;
        i:=i+1;
        if i>n then done:=true; end if;
        exit when done;
      end loop;
   end find_a_zero;

   function star_in_row(row : integer) return boolean is
    tbool : boolean;
   begin
      tbool:=false;
      for j in 1..n loop
        if M(row,j)=1 then
           tbool:=true;
        end if;
      end loop;
      return tbool;
   end star_in_row;

   procedure find_star_in_row(row, col : in out integer) is
   begin
      col:=0;
      for j in 1..n loop
        if M(row,j)=1 then
           col:=j;
        end if;
      end loop;
   end find_star_in_row;

 begin
   done:=false;
   while not(done) loop
     find_a_zero(row,col);
     if row=0 then
       done:=true;
       step:=6;
     else
       M(row,col):=2;
       if star_in_row(row) then
         find_star_in_row(row,col);
         R_cov(row):=1;
         C_cov(col):=0;
```

```
      else
        done:=true;
        step:=5;
        Z0_r:=row;
        Z0_c:=col;
      end if;
    end if;
  end loop;
end stepfour;
```

Step 5

Construct a series of alternating primed and starred zeros as follows. Let Z_0 represent the uncovered primed zero found in Step 4. Let Z_1 denote the starred zero in the column of Z_0 (if any). Let Z_2 denote the primed zero in the row of Z_1 (there will always be one). Continue until the series terminates at a primed zero that has no starred zero in its column. Unstar each starred zero of the series, star each primed zero of the series, erase all primes and uncover every line in the matrix. Return to Step 3. You may notice that Step 5 seems vaguely familiar. It is a verbal description of the augmenting path algorithm (for solving the maximal matching problem) which we discussed in Lecture 3. We decompose the operations of this step into a main procedure and five relatively simple subprograms.

```
 procedure stepfive(step : in out integer) is
    count : integer;
    done  : boolean;
    r,c   : integer;

  procedure find_star_in_col(c : in integer; r : in out integer) is
  begin
    r:=0;
    for i in 1..n loop
      if M(i,c)=1 then
        r:=i;
      end if;
    end loop;
  end find_star_in_col;

  procedure find_prime_in_row(r : in integer; c : in out integer) is
  begin
    for j in 1..n loop
      if M(r,j)=2 then
        c:=j;
      end if;
    end loop;
  end find_prime_in_row;

  procedure convert_path is
  begin
    for i in 1..count loop
      if M(path(i,1),path(i,2))=1 then
        M(path(i,1),path(i,2)):=0;
      else
        M(path(i,1),path(i,2)):=1;
      end if;
    end loop;
  end convert_path;

  procedure clear_covers is
  begin
    for i in 1..n loop
      R_cov(i):=0;
      C_cov(i):=0;
    end loop;
  end clear_covers;
```

```
  procedure erase_primes is
  begin
    for i in 1..n loop
      for j in 1..n loop
        if M(i,j)=2 then
          M(i,j):=0;
        end if;
      end loop;
    end loop;
  end erase_primes;

  begin
    count:=1;
    path(count,1):=z0_r;
    path(count,2):=z0_c;
    done:=false;
    while not(done) loop
      find_star_in_col(path(count,2),r);
      if r>0 then
        count:=count+1;
        path(count,1):=r;
        path(count,2):=path(count-1,2);
      else
        done:=true;
      end if;
      if not(done) then
        find_prime_in_row(path(count,1),c);
        count:=count+1;
        path(count,1):=path(count-1,1);
        path(count,2):=c;
      end if;
    end loop;
    convert_path;
    clear_covers;
    erase_primes;
    step:=3;
  end stepfive;
```

Step 6

Add the value found in Step 4 to every element of each covered row, and subtract it from every element of each uncovered column. Return to Step 4 without altering any stars, primes, or covered lines. Notice that this step uses the smallest uncovered value in the cost matrix to modify the matrix. Even though this step refers to the value being found in Step 4 it is more convenient to wait until you reach Step 6 before searching for this value. It may seem that since the values in the cost matrix are being altered, we would lose sight of the original problem. However, we are only changing certain values that have already been tested and found not to be elements of the minimal assignment. Also we are only changing the values by an amount equal to the smallest value in the cost matrix, so we will not jump over the optimal (i.e. minimal assignment) with this change.

```
  procedure stepsix(step : in out integer) is
    minval : integer;

    procedure find_smallest(minval : out integer) is
    begin
      minval:=integer'last;
      for i in 1..n loop
        for j in 1..n loop
          if R_cov(i)=0 and C_cov(j)=0 then
            if minval>C(i,j) then
              minval:=C(i,j);
            end if;
          end if;
        end loop;
```

```
      end loop;
   end find_smallest;

begin
   find_smallest(minval);
   for i in 1..n loop
      for j in 1..n loop
         if R_cov(i)=1 then
            C(i,j):=C(i,j)+minval;
         end if;
         if C_cov(j)=0 then
            C(i,j):=C(i,j)-minval;
         end if;
      end loop;
   end loop;
   step:=4;
end stepsix;
```

An Example Execution of Munkres' Algorithm

Workers = { a, b, c}
Jobs = {p, q, r}

Cost of assigning job j to work i
is

$$C(i,j) = \begin{bmatrix} 1 & 2 & 3 \\ 2 & 4 & 6 \\ 3 & 6 & 9 \end{bmatrix}$$

	p	q	r
a	1	2	3
b	2	4	6
c	3	6	9

1. Step 0

	p	q	r
a	0	1	2
b	0	2	4
c	0	3	6

2. Step 1

	p	q	r
a	0*	1	2
b	0	2	4
c	0	3	6

3. Step 2

	p	q	r
a	0*	1	2
b	0	2	4
c	0	3	6

4. Step 3

	p	q	r
a	0*	1	2
b	0	2	4
c	0	3	6

5. Step 4

	p	q	r
a	0*	0	1
b	0	1	3
c	0	2	5

6. Step 6

	p	q	r
a	0*	0'	1
b	0'	1	3
c	0	2	5

7. Step 4

	p	q	r
a	0*	0'	1
b	0'	1	3
c	0	2	5

8. Step 5

	p	q	r
a	0	0*	1
b	0*	1	3
c	0	2	5

9. Step 3

	p	q	r
a	0	0*	1
b	0*	1	3
c	0	2	5

10. Step 4

	p	q	r
a	0	0*	0
b	0*	1	2
c	0	2	4

11. Step 6

	p	q	r
a	0	0*	0'
b	0*	1	2
c	0	2	4

12. Step 4

	p	q	r
a	1	0*	0'
b	0*	0	1
c	0	1	3

13. Step 6

	p	q	r
a	1	0*	0'
b	0*	0'	1
c	0'	1	3

14. Step 4

	p	q	r
a	1	0*	0'
b	0*	0'	1
c	0'	1	3

15. Step 5

	p	q	r
a	1	0	0*
b	0	0*	1
c	0*	1	3

16. Step 3

	p	q	r
a	1	0	0*
b	0	0*	1
c	0*	1	3

17. Done

This example illustrates the method of implementing a step-algorithm. It also serves to demonstrate why we do not attempt to implement every algorithm discussed in this course. :-)

Answers to Frequently Asked Questions

1. The algorthm will work even when the minimum values in two or more rows are in the same column.
2. The algorithm will work even when two or more of the rows contain the same values in the the same order.
3. The algorithm will work even when all the values are the same (although the result is not very interesting).
4. Munkres Assignment Algorithm is not exponential run time or intractable; it is of a low order polynomial run time.
5. Optimality is guaranteed in Munkres Assignment Algorithm.
6. Setting the cost matrix to $C(i,j) = i*j$ makes a good testing matrix for this problem.
7. In this algorithm the range of indices is[1..n] rather than [0..n-1]. Using the latter range makes the sample code less that helpful in your algorithm implementation.
8. Step 3 is an example of the greedy method. If the minimum values are all in different rows then their positions represent the minimal pairwise assignments.
9. Step 5 is an example of the Augmenting Path Algorithm (see Stable Marriage Problem).
10. Step 6 is an example of contraint relaxation. It is "giving up" on a particular cost and raising the constraint by the least amount possible.
11. If your implementation is jumping between Step 4 and Step 6 without entering Step 5, you probably have not properly dealt with recognizing that there are no uncovered zeros in Step 4.
12. In the matrix M 1=starred zero and 2=primed zero. So, if C[i,j] is a starred zero we would set M[i,j]=1. All other elements in M are set to zero
13. The Munkres assignment algorithm can be implemented as a sparse matrix, but you will need to ensure that the correct (optimal) assignment pairs are active in the sparse cost matrix C
14. Munkres Assignment can be applied to TSP, pattern matching, track initiation, data correlation, and (of course) any pairwise assignment application.
15. Munkres can be extended to rectangular arrays (i.e. more jobs than workers, or more workers than jobs) .
16. The best way to find a maximal assignment is to replace the values ci,j in the cost matrix with C[i,j] = bigval - ci,j.
17. Original Reference: Algorithms for Assignment and Transportation Problems, James Munkres, Journal of the Society for Industrial and Applied Mathematics Volume 5, Number 1, March, 1957
18. Extension to Rectangular Arrays Ref: F. Burgeois and J.-C. Lasalle. An extension of the Munkres algorithm for the assignment problem to rectangular matrices. Communications of the ACM, 142302-806, 1971.