Numerical Analysis in Econom(etr)ic Softwares: the Data-Memory Shortage Management

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Numerical Analysis in Econom(etr)ic* Softwares: the Data-Memory Shortage Management

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Summary

The econometricians and the economic modelers have to know and use numerical analysis not only to have a good understanding of their results, but sometimes to built their own tools. The increasing tendency to the use of large-scale models leads the softwares to reach the limit of the saturation of the Data-Memory. In this paper, we present an intuitive procedure DMT - i.e. Disk-Matrix Technique - which could help econom(etr)ic software developers to correct the problem of the data-memory shortage during the building of their own software.

Key-words : Algorithms ; Numerical Analysis ; Econom(etr)ic Softwares ; Operating System ; Data-Memory ; Data Storage

JEL Classification : C15, C82, C87

* - Please read "economic and/or econometric".
Econometricians and Economic Modelers usually use black boxes during their works\(^1\). Most of them use the metalanguage of softwares such as GAMS, MATLAB, MAPLE, Mathematica, SAS, GAUSS, G, and so on. However, econometricians and/or economic modelers have to take part to the building of the econometric softwares\(^2\). That is a condition of understanding between developers and users - see \([1, 3, 4, 10]\). Thus, some other prefer to use primary languages such as C, Fortran, Java, Pascal and so on to implement their own procedures \([11, 9]\). Our paper reports some algorithmic problems we had to solve in building a multi-dimensional\(^3\) economic modelling software (SIMUL i.e. Système Intégré de Modélisation multi-dimensional)**\(^4\). Since the 70th, the econometric modelers developed large-scale models \([19]\) and this use should never decrease in the future, especially if we consider the micro-simulation modelling, the agent-based computational economics, or the artificial societies simulations. That means the software often reach the limits of the Data-memory. Some languages don’t manage data-memory efficiently\(^5\). Even if the memory of computers increases, the size of the data-memory of such language is limited to 640 Ko - e.g. : Turbo-Pascal\(^6\). An algorithm has already developed \([12]\) which uses hard disk and a buffer to solve this trouble. However, in this paper we present another algorithm only using hard disk to storage the too large arrays - technique called "disk-matrix technique" (DMT). Such algorithms increase the size of array in programs without decreasing readability of the program, but the disk-access delay is longer than the RAM-access delay.

In the first part, we’ll present the solution of the two dimensional problem, then we’ll generalize it in presenting the solution of the multi-dimensional problem. In the second part, we’ll expose the implementation of our algorithm to the matrix

\(^1\) Not at the begin of the econometric period, during the econometricians have to built their programs. About this period, see [15]

\(^2\) Despite of this fact, "There is no ideal software" [20].

\(^3\) Multi-regional (r), multi-sectoral (s), multi-periodical (t) analysis.

\(^4\) The building of this software is integrated in the economic thesis we are finishing, "Modélisation multi-dimensionnelle et analyse multirégionale de l’économie française" under the supervision of the Professor Raymond Courbis. Our algorithm has been integrated in the Data bank management module GEBANK. This software usually can quickly estimate multi-dimensional econometric equations and solve the whole system: we only need an instruction to implement such equations : \(Y_t^{r,s} = X_t^{r,s} \cdot a^{r,s} + e_t^{r,s}\). So it’s divided into some modules (estimation, data bank management, resolving, etc.). For an overview of SIMUL software, see our papers [6, 7].

\(^5\) As computational economics’ referees said us, Windows 95 introduced virtual memory in Microsoft’s line of operating systems and it has been used in every subsequent version of Windows. It automatically uses disk space if RAM proves insufficient for a program. Further, the limits are quite large today. Windows XP, currently the most common version, supports 4GB of RAM with up to 2 GB per per process. We agree with this, however, we consider this technique useful in the perspective of a very large simulation.

\(^6\) We can implement some memory units [2, 22] but they depend on the hardware.
product. The third part will present the application of the DMT to the matrix inversion procedure. Then we’ll discuss the relevance of this technique.

1. The Two Dimensional Problem

The location of an element in a matrix is given by the row \( i \) and column \( j \) ranks. In a vector, there is nor row neither column, so we have to find a relationship between the couple \((i, j)\) and the rank of the same element in an output vector. The problem is the following : we try to find the rank \( r \) in the vector \( V \) of the element \( m_{i,j} \) of a matrix \( M_{N_1,N_2} \)-sized. We can observe three steps.

1st step - The first row is filled until the \((N_1)\)-th cell.

\[
\begin{array}{c|c|c|c}
\text{Matrix} & \text{Vector} \\
\hline
& \end{array}
\]

\[
\begin{array}{cccc}
1 & 2 & \ldots & N_1 \\
& & \ddots & \vdots \\
& & & \ddots \\
& & & & \vdots \\
& & & & N_1 \\
\end{array}
\]

2nd step - The matrix \( M \) is filled until the \((j - 1)\)-th following rows.

\[
\begin{array}{c|c|c|c}
\text{Matrix} & \text{Vector} \\
\hline
& \end{array}
\]

\[
\begin{array}{cccc}
1 & 2 & \ldots & N_1 \\
N_1 + 1 & \ldots & \ldots & \ldots \\
(j - 2)N_1 + 1 & (j - 2)N_1 + 2 & \ldots & (j - 1)N_1 \\
& & \ddots & \vdots \\
& & & \vdots \\
& & & \vdots \\
& & & N_1 \\
\end{array}
\]

3rd step - When it reaches the \( j \)-e row, the algorithm only fill the \( i \) first cells. Then algorithm has reached \( M_{i,j} \).

\[
\begin{array}{c|c|c|c}
\text{Matrix} & \text{Vector} \\
\hline
& \end{array}
\]

\[
\begin{array}{cccc}
1 & 2 & \ldots & N_1 \\
N_1 + 1 & \ldots & \ldots & \ldots \\
(j - 2)N_1 + 1 & (j - 2)N_1 + 2 & \ldots & (j - 1)N_1 \\
& & \ddots & \vdots \\
& & & \vdots \\
& & & \vdots \\
& & & (j - 1)N_1 \\
\end{array}
\]
Thus, the relationship \( r = f(i, j) \) we obtain is given by
\[
 r = i + (j - 1).N_1
\]

2. The Multi-Dimensional Problem

Let’s consider now an hyper-matrix with \( k \) dimensions \( N_1, N_2, \ldots, N_k \). We want to find the relationship between \( r \) and the rank \((i_1, i_2, \ldots, i_k)\) of all its dimensions. Firstly, we can observe that in the two dimensional problem, the formula doesn’t depend on the second dimension \( N_2 \)
\[
r = i + (j - 1).N_1
\]
in three dimensions, the formula doesn’t depend on \( N_3 \)
\[
r = i_1 + (i_2 - 1).N_1 + (i_3 - 1).N_1.N_2
\]
in four dimensions, the formula doesn’t depend on \( N_4 \)
\[
r = i_1 + (i_2 - 1).N_1 + (i_3 - 1).N_1.N_2 + (i_4 - 1).N_1.N_2.N_3
\]
and so on.

Then, to find the general formula, we have to observe the following equivalences
\[
r = i + (j - 1).N_1 \quad \Leftrightarrow \quad r = 1 + (i - 1) + (j - 1).N_1
\]
\[
r = i_1 + (i_2 - 1).N_1 + (i_3 - 1).N_1.N_2 \quad \Leftrightarrow \quad r = 1 + (i_1 - 1) + (i_2 - 1).N_1 + (i_3 - 1).N_1.N_2
\]
\[
r = i_1 + (i_2 - 1).N_1 + (i_3 - 1).N_1.N_2 + (i_4 - 1).N_1.N_2.N_3 \quad \Leftrightarrow \quad r = 1 + (i_1 - 1) + (i_2 - 1).N_1 + (i_3 - 1).N_1.N_2 + (i_4 - 1).N_1.N_2.N_3
\]
hence \( \forall M_K \) an hyper-matrix with \( K \) dimensions \( N_1^H, N_2^H, \ldots, N_K^H \), where \( N_i^H \) are the size of the \( i \)-dimensions and \( V \) vector \( N^V \)-sized
\[
 N^V = \prod_{j=1}^{K} N_j^H
\]
hence we obtain \( r \) the rank in \( V \) of the \((i_1,i_2,\ldots,i_K)\)-the element with the formula
\[
r = 1 + \sum_{p=1}^{K} (i_p - 1).\prod_{j=1}^{p-1} N_j^H
\]
3. Implementation in The Matrix Product Algorithm

The classical algorithm of the matrix product is given by

**Algorithm 1** - The Classical Matrices Product Algorithm

```plaintext
for i = 1 to dim1 do
    for j = 1 to dim2 do
        X(i, j) ← 0,
        for k = 1 to dim2 do
            X(i, j) ← X(i, j) + A(i, k) * B(k, j)
        end for
    end for
end for
```

To apply our algorithm, we have replaced the ARRAY OF REAL of the classical matrix product procedure by a REAL FUNCTION - see Fig.2. The element $i, j$ or of the array $X$ becomes the function $XX$ applied to the parameters $^7$ $dim_1, i, j$.

**Algorithm 2** - The Disk-Matrix Procedure

```plaintext
FUNCTION xx(VAR fil:FILE OF REAL;
            maxsiz1:INTEGER;
            x1,x2:INTEGER) : REAL;
VAR pos:INTEGER;
   data:REAL;
BEGIN
    pos ← (x1 - 1) * maxsiz1 + x2 - 1;
    SEEK(fil, pos);
    READ(fil, data);
    xx ← data;
END;
```

**Algorithm 3** - The Classical Matrices Product Including the Disk-Matrix Procedure

```plaintext
PROCEDURE new_promat(VAR x:MATRI
    for i=1 to dim1 do
        for j=1 to dim2 do
            x ← 0,
            for k=1 to dim2 do
                x ← x + xx(f1, dim1, i, k) * xx(f2, dim1, k, j)
            end for
        end for
end for
```

7. We find another parameter for the file. In fact we have to create two functions : one function to read the element $i, j$ in the file, and one function to write the element $i, j$ in the file.
4. Application: The Matrix Inversion Procedure

The following procedure of Matrix inversion is translated from the program of [14] - about the numerical analysis see [13] and in Turbo-Pascal [5] too.

```pascal
PROCEDURE INVMA2(VAR sma:STRING;
                  VAR D1:integer;
                  VAR smb:STRING);
Var ii,jj,lig,col:integer;
found_inv,found_piv,acheve:boolean;
xdata:real;
_frl:FRL; (** FRL=FILE OF REAL ***)
begin
  for ii:=1 to D1 do begin
    for jj:=1 to D1 do begin
      if (ii=jj) then xdata:=1.0
        else xdata:=0.0;
    WM(smb,xdata,D1,D1,ii,jj);
    end;
  end;

  jj:=1;
  acheve:=false;
  found_inv:=true;
  while((jj<=D1) and (found_inv=true)) do begin
    CHEPIV2(sma,D1,D1,jj,lig,col,found_piv);
    if ((found_piv=true) and (lig=jj)) then begin
      if (col<>jj) then begin
        ECHCO2(sma,jj,col,D1);
        ECHCO2(smb,jj,col,D1);
      end;
    ELIDE2(sma,smb,D1,lig,jj);
    end
    else begin
      acheve:=true;
      found_inv:=false;
    end;
    jj:=jj+1;
  end; { while }

  if (found_inv=true) then begin
    for jj:= D1 DownTo 1 do begin
      gotoxy(64,12); write(jj:5);
      ELIAS2(sma,smb,D1,jj,jj);
    end;
  end;

  for ii:=1 to D1 do begin
    for jj:=1 to D1 do begin
      xdata:=RM(smb,D1,D1,ii,jj)/RM(sma,D1,D1,jj,jj);
      WM(smb,xdata,D1,D1,ii,jj);
    end;
  end;
end;
```
PROCEDURE ELIAS2(VAR smata, smatb: STRING;
VAR D1, lig, col: integer);
Var ii, jj: integer;
facteur: EXTENDED;
mdat: real;
begin
for jj:=col-1 DownTo 1 do begin
  facteur:=RM(smata, D1, D1, lig, jj)/RM(smata, D1, D1, lig, col);
  for ii:=1 to D1 do begin
    mdat:=RM(smata, D1, D1, ii, jj)-facteur*RM(smata, D1, D1, ii, col);
    WM(smata, mdat, D1, D1, ii, jj);
    mdat:=RM(smatb, D1, D1, ii, jj)-facteur*RM(smatb, D1, D1, ii, col);
    WM(smatb, mdat, D1, D1, ii, jj);
  end;
end;
end;

PROCEDURE ELIDE2(VAR smata, smatb: STRING;
VAR D1, lig, col: integer);
Var ii, jj: integer;
facteur: EXTENDED;
mdat: real;
begin
for jj:=col+1 to D1 do begin
  facteur:=RM(smata, D1, D1, lig, jj)/RM(smata, D1, D1, lig, col);
  for ii:=1 to D1 do begin
    mdat:=RM(smata, D1, D1, ii, jj)-facteur*RM(smata, D1, D1, ii, col);
    WM(smata, mdat, D1, D1, ii, jj);
    mdat:=RM(smatb, D1, D1, ii, jj)-facteur*RM(smatb, D1, D1, ii, col);
    WM(smatb, mdat, D1, D1, ii, jj);
  end;
end;
end;

PROCEDURE CHEPIV2(VAR _smatr: STRING;
VAR D1, D2, ordre, lig, col: integer;
VAR found: boolean);
Var jj: integer;
max: EXTENDED;
xval: real;
begin
  max:=Epsilon;
  found:=false;
  lig:=ordre;
  repeat
    for jj:=lig to D2 do begin
      xval:=RM(_smatr, D2, D1, lig, jj);
      if (abs(xval)>max) then begin
        found:=true;
        col:=jj;
        max:=abs(xval);
      end;
    end;
    if (found=false) then lig:=lig+1;
  until ((found=true) or (lig>D1));
end;
PROCEDURE ECHCO2(VAR _sm:STRING;
VAR col1,col2,D1:integer);
Var ii:integer;
tmp1,tmp2:real;
begin
for ii:=1 to D1 do begin
  tmp1:=RM(_sm,d1,d1,ii,col1);
  tmp2:=RM(_sm,d1,d1,ii,col2);
  WM(_sm,tmp2,d1,d1,ii,col1);
  WM(_sm,tmp1,d1,d1,ii,col2);
end;
end;

FUNCTION RM(VAR _smat:STRING;
VAR _dd1,_dd2:INTEGER;
VAR _iii,_jjj:INTEGER): REAL;
VAR _pos:longint;
_fr:FRL;
_x:REAL;
BEGIN
  Assign(_fr,'\LARGMAT\'+_smat+'.FRL');
  Reset(_fr);
  _pos:=1+(_iii-1)+(_jjj-1)*_dd1;
  Seek(_fr,_pos);
  read(_fr,_x);
  Close(_fr);
  RM:=_x;
END;

PROCEDURE WM(VAR _SMAT:STRING;
VAR _x:REAL;
VAR _dd1,_dd2:INTEGER;
VAR _iii,_jjj:INTEGER);
VAR _pos:longint;
IOError:INTEGER;
_fr:FRL;
BEGIN
  Assign(_fr,'\LARGMAT\'+_smat+'.FRL');
  Reset(_fr);
  _pos:=1+(_iii-1)+(_jjj-1)*_dd1;
  Seek(_fr,_pos);
  write(_fr,_x);
  Close(_fr);
END;

The matrix inversion procedure INVMA2 includes the procedures ELIAS2 - "élimination ascendante" -, ELIDE2 - "élimination descendante" -, CHEPIV2 - "changement de pivot" -, ECHCO2 - "échange de colonne". The function RM - "read matrix" - gets a read access to the data on the hard disk, and the procedure WM - "write matrix" - gets a write access to save the data on the hard disk. The \LARGMAT\ directory is used to store the disk-matrix in temporary files. In our library we implemented two matrix inversion procedures. One without DMT (INVMA1) if the size of the matrix is weak, and another (INVMA2) with DMT procedures. The main program chooses the relevant procedure after a matrix-size test.
5. Conclusion

The DMT function and procedure are easy to implement and can be integrated in a library. The classical numerical analysis procedures can be easily changed to receive the DMT function and procedure, and the new procedures are always easily readable. The inconvenient of this technique is the time of calculation - see Tables 1 and 2. However, except the progress of the hardware, there is no really alternative, in considering the economic modelers should built more and more large-scale models, whatever the field: econometrics, micro-simulation, artificial life, ACE. We have assume we used the "classical" calculation algorithms, however, we have to quote the recent advances of matrix multiplication techniques [8, 17, 18] which could help to economize a lot of resources.

Table 1: Disk-Matrix Technique Applied to Matrix Product

<table>
<thead>
<tr>
<th>Matrix size</th>
<th>Duration</th>
<th>Memory</th>
</tr>
</thead>
<tbody>
<tr>
<td>75</td>
<td>4&quot;</td>
<td>33 750</td>
</tr>
<tr>
<td>100</td>
<td>8&quot;</td>
<td>60 000</td>
</tr>
<tr>
<td>150</td>
<td>30&quot;</td>
<td>135 000</td>
</tr>
<tr>
<td>175</td>
<td>46&quot;</td>
<td>183 750</td>
</tr>
<tr>
<td>200</td>
<td>1’02&quot;</td>
<td>240 000</td>
</tr>
<tr>
<td>500</td>
<td>16’16&quot;</td>
<td>1 500 000</td>
</tr>
</tbody>
</table>

With 3 GHz CPU.

Table 2: Disk-Matrix Technique Applied to Matrix Inversion

<table>
<thead>
<tr>
<th>Size</th>
<th>Durations According Methods</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Pointers</td>
</tr>
<tr>
<td>10</td>
<td>01&quot;</td>
</tr>
<tr>
<td>20</td>
<td>01&quot;</td>
</tr>
<tr>
<td>50</td>
<td>02&quot;</td>
</tr>
<tr>
<td>80</td>
<td>03&quot;</td>
</tr>
<tr>
<td>100</td>
<td>05&quot;</td>
</tr>
<tr>
<td>200</td>
<td>-</td>
</tr>
<tr>
<td>250</td>
<td>-</td>
</tr>
<tr>
<td>500</td>
<td>-</td>
</tr>
</tbody>
</table>

With 3 GHz CPU

8. See [23]
9. See the paper of L. Tesfatsion [21].
10. About some sophisticated algorithms see http://www.nag.co.uk or http://www.vni.com/products/imsl/ among others.
References


