

# This document is discoverable and free to researchers across the globe due to the work of AgEcon Search. 

## Help ensure our sustainability. Give to AgEcon Search

AgEcon Search
http://ageconsearch.umn.edu
aesearch@umn.edu

Papers downloaded from AgEcon Search may be used for non-commercial purposes and personal study only. No other use, including posting to another Internet site, is permitted without permission from the copyright owner (not AgEcon Search), or as allowed under the provisions of Fair Use, U.S. Copyright Act, Title 17 U.S.C.

## UCD

 of Agricultural EconomicsWorking papers are circulated by the author without formal review. All inquiries should be addressed to the author, Department of Agricultural Economics, University of California, Davis, CA 95616.

# CONSISTENT AGGREGATION OF LINEAR COMPLEMENTARITY PROBLEMS 

by<br>Quirino Paris

Working Paper No. 78-1

## ABSTRACT <br> Consistent Aggregation of Linear Complementarity Problems

## by

Quirino Paris

Aggregating linear complementarity problems under a general definition of constrained consistency leads to the possibility of consistent aggregation of linear and quadratic programming models. Under this formulation, consistent aggregation of dual variables is also achieved. Furthermore, the existence of multiple sets of aggregation operators is illustrated with a numerical example. Such multiple operators allow considerable flexibility of the microstructures admitting consistent aggregation.

## Consistent Aggregation of Linear Complementarity Problems

Constrained consistency-rather than total consistency-was suggested in this journal by Guccione and Oguchi [2] as the proper framework to analyze the aggregation of linear programs. They dealt, however, with a very special case of constrained consistency and, furthermore, they did not consider the aggregation of dual variables.

In this note we discuss the aggregation of linear complementarity (LC) problems under a definition of constrained consistency which is more general than that suggested in [2]. Yet, it is empirically applicable, as demonstrated by a numerical example. We achieve the following results: (a) aggregation conditions are extended to cover any LP problem, symmetric and asymmetric quadratic programs of any structure, and two-person-non-zero-sum games; (b) perfect aggregation under constrained consistency includes that of dual variables in all the models admitted by the LC problem; (c) further restrictions on the structure of the LC problem allow the interpretation of the aggregation conditions within a variety of empirical contexts; (d) a numerical example illustrates the feasibility (in principle) of the approach.

## Specification of the Aggregation Problem

The micro linear complementarity problem to be aggregated is defined as follows: find vectors $z$ and $w$ such that

$$
\begin{equation*}
-q=-w+M z, \quad w \geq 0, \quad z \geq 0 \tag{1}
\end{equation*}
$$

and

$$
\begin{equation*}
z^{\prime} w=0 \tag{2}
\end{equation*}
$$

where $q$ is a $[(m+n) x l]$ vector of known coefficients and $M$ is a given $[(m+n) x(m+n)]$ positive semidefinite matrix.

Instead of solving problem (1) and (2), it is desired to consider a problem of the same form but of smaller dimensions such as: find vectors $z_{a}$ and $w_{a}$ which solve

$$
\begin{equation*}
-q_{a}=-w_{a}+M_{a} z_{a}, \quad w_{a} \geq 0, \quad z_{a} \geq 0 \tag{3}
\end{equation*}
$$

and

$$
\begin{equation*}
z_{a}^{\prime} w_{a}=0, \tag{4}
\end{equation*}
$$

where $q_{a}$ is a $[(h+k) x l]$ arbitrary vector of known coefficients, $M_{a}$ is a given $[(h+k) x(h+k)]$ positive semidefinite matrix and, of course $(h+k)<(m+n)$.

The aggregation problem relating (1) and (2) to (3) and (4)--as indicated by Guccione and Oguchi [2]--must be considered under some condition of constrained consistency. This notion was originally introduced by Ijiri [4] in his masterful survey of aggregation theory. It should be clear that an appropriate defintion of constrained consistency will allow to achieve the solution of the aggregation problem under empirically flexible structures. Thus, our choice is the following: the domain of the microvariables $q, w$ and $z$ is restricted by linear rules such that

$$
\begin{align*}
& q=U q_{a}  \tag{5}\\
& w^{*}=U w_{a}^{*}  \tag{6}\\
& z^{*}=T z_{a}^{*} \tag{7}
\end{align*}
$$

where $U$ and $T$ are nonnegative linear operators of full rank and of dimensions $[(m+n) x(h+k)]$; the starred elements are solutions of the respective LC problems. Under this specification, any arbitrary choice of the vector $q_{a}$ is admitted. This, in turn, will generate a wide spectrum of solutions ( $z_{a}^{*}, w_{a}^{*}$ ) of the aggregate LC problem (3) and (4). The stringency of the restrictions on the domains of themicrovariables $q$, wand $z$, imposed by
the linear rules specified above depend upon the structure of operators $T$ and $U$. Guccione and Oguchi chose to work with very simplified operators requiring a proportionality relation between the micro and the macroelements of their LP problems. In this paper, this proportionality relation is unnecessary.

It is now possible to state the perfect aggregation problem under constrained consistency: for any set of microvariables ( $q$, w*, $z^{*}$ ) satisfying (5), (6) and (7), the following relations must hold

$$
\begin{align*}
& q_{a}=T^{\prime} q  \tag{8}\\
& w_{a}^{*}=T^{\prime} w^{*}  \tag{9}\\
& z_{a}^{*}=U^{\prime} z^{*} . \tag{10}
\end{align*}
$$

Therefore, the conditions imposed upon the structure of problems [(1), (2)] and [(3), (4)] for achieving exact aggregation as defined above are given in the following

Theorem. Perfect aggregation of LC problems under constrained consistency is obtained if and only if

$$
\begin{align*}
& M^{\prime} T=U M_{a}^{\prime}  \tag{11}\\
& T^{\prime} U=I \tag{12}
\end{align*}
$$

where $I$ is an identity matrix.
Proof. (Necessity.) Premultiplying (1) by $T^{\prime}$ and substituting (10) into
(3) we get

$$
\begin{align*}
& T^{\prime}(-q)+T^{\prime} w=T^{\prime} M z  \tag{13}\\
& -q_{a}+w_{a}=M_{a} U^{\prime} z . \tag{14}
\end{align*}
$$

Hence, in view of (8) and (9),

$$
\begin{equation*}
z^{\prime} M^{\prime} T=z^{\prime} U M_{a}^{\prime} \tag{15}
\end{equation*}
$$

and (11) follows because (15) must hold for any $z$ satisfying (1) and (2) under parametric variation of $q$ as defined by (5) (recall that $q_{a}$ is arbitrary). Condition (12) is easily obtained by premultiplying (5) and (7) by $\mathrm{T}^{\prime}$ and comparing the result to (8) and (10), respectively.
(Sufficiency.) If (ll) holds, any solution ( $z, w$ ) of the expanded LC problem (1) and (2) which also satisfies (5), (6) and (7), produces a solution ( $z_{a}, w_{a}$ ) for the aggregate problem (3) and (4). In fact, assume that ( $z^{*}, w^{*}$ ) represents a solution to (1) and (2). Then, $z_{a}^{*} \geq 0$ and $w_{a}^{*} \geq 0$, since $T$ and $U$ are nonnegative operators, and

$$
\begin{equation*}
z_{a}^{*^{\prime}} M_{a}^{\prime}=z^{\prime} U M_{a}^{\prime}=z^{\prime \prime} M^{\prime} T=\left(-q^{\prime}+w^{\prime}\right) T=-q_{a}^{\prime}+w_{a}^{*^{\prime}} \tag{16}
\end{equation*}
$$

which establishes the feasibility of $z_{a}^{*}$ and $w_{a}^{*}$. To show that they constitute also a complementary solution satisfying (4) it is sufficient to recall that ( $z^{*}, w^{*}$ ) is a complementary solution (by assumption) and

$$
\begin{equation*}
z_{a}^{\star \prime} w_{a}^{\star}=z^{\star} U T^{\prime} w^{*}=z^{\star \prime} w^{\star}=0 \tag{17}
\end{equation*}
$$

because $U T^{\prime} w^{*}=U w_{a}^{*}=w^{*}$, according to (6) and (9). Q.E.D.

## Implications

It is interesting to note that the essential role of constrained consistency is required only for proving the complementarity of the aggregate solution in the sufficiency part of the theorem. For problems such as Leontief input-output analysis, where complementarity is trivially satisfied, it is not necessary (Hatanaka [3]) to invoke any form of constrained consistency to achieve the solution of the aggregation problem.

Another implication, especially important for empirical applications, concerns the multiplicity of the aggregation operators $T$ and $U$. Hence, any other pair of matrices ( $T *$, $\mathrm{U}^{*}$ ) such that (6), (7), (11) and (12) are satisfied, constitutes an alternative pair of aggregation operators.

The LC problem encompasses several mathematical programing structures. It is, thus, of interest to analyze the aggregation conditions more explicitly. To begin, let us consider the dual pair of symmetric quadratic programming problems formulated by Cottle [1]:

$$
\begin{align*}
& \max \left\{c^{\prime} x-x^{\prime} Q x-y^{\prime} E y\right\}  \tag{18}\\
& \text { subject to } A x-2 E y \leq b \\
& \text { and } \quad x \geq 0, y \geq 0 \\
& \text { min }\left\{b^{\prime} y+x^{\prime} Q x+y^{\prime} E y\right\} \\
& \text { subject to } A^{\prime} y+2 Q x \geq c
\end{align*}
$$

and $\quad \mathrm{x} \geq 0, \mathrm{y} \geq 0$
where $Q$ and $E$ are known positive semidefinite symmetric matrices of order $n$ and $m$, respectively; $c$ and $b$ are given vectors of coefficients which admit parametric variations. It is easy to show that this specification may be stated as a LC problem when the following correspondence is established

$$
q=\left[\begin{array}{c}
b \\
-c
\end{array}\right], z=\left[\begin{array}{l}
y \\
x
\end{array}\right], M=\left[\begin{array}{cc}
2 E & -A \\
A^{\prime} & 2 Q
\end{array}\right], w=\left[\begin{array}{l}
v \\
u
\end{array}\right]
$$

where $v$ and $u$ are vectors of slack variables associated with the primal and dual constraints, respectively. For this symmetric quadratic programing structure the aggregation operators $T$ and $U$ are specified as block-diagonal matrices

$$
T=\left[\begin{array}{ll}
W & 0  \tag{20}\\
0 & D
\end{array}\right], \quad U=\left[\begin{array}{ll}
P & 0 \\
0 & G
\end{array}\right]
$$

The submatrices $W$ and $P$ are of dimensions (mxh), while $D$ and $G$ are of dimensions ( nxk ), $\mathrm{h}<\mathrm{m}, \mathrm{k}<\mathrm{n}$. The aggregation condition (11) now becomes

$$
\left[\begin{array}{cc}
2 E & A  \tag{21}\\
-A^{\prime} & 2 Q
\end{array}\right]\left[\begin{array}{ll}
W & 0 \\
0 & D
\end{array}\right]=\left[\begin{array}{ll}
P & 0 \\
0 & G
\end{array}\right]\left[\begin{array}{cc}
2 E_{a} & A_{a} \\
-A^{\prime} & 2 Q_{a}
\end{array}\right]
$$

or, equivalently

$$
\begin{align*}
E W & =P E_{a}  \tag{22}\\
A^{\prime} W & =G A_{a}^{\prime}  \tag{23}\\
A D & =P A_{a}  \tag{24}\\
Q D & =G Q_{a} . \tag{25}
\end{align*}
$$

The matrices $E_{a}$ and $Q_{a}$ are symmetric and positive semidefinite of order $h$ and $k$, respectively, and belong to the aggregate symmetric quadratic programming problem. The matrix $A_{a}$ is of dimensions (hxk). The aggregation condition (12) corresponds to

$$
\begin{align*}
& W^{\prime} P=I_{h}  \tag{26}\\
& D^{\prime} G=I_{k} \tag{27}
\end{align*}
$$

where $I_{h}$ and $I_{k}$ are identity matrices of order $h$ and $k$, respectively. Notice that if $E=0$ and $E_{a}=0$, condition (22) vanishes and the remaining relations constitute the conditions for aggregating the traditional asymmetric quadratic programming problem. Furthermore, if also $Q=0$ and $Q_{a}=0$, relation (25) vanishes and the residual relations express the requirements for aggregating linear programs.

Further insight into the structure of conditions (22) through (25) is achieved, if by using (26) and (27) as needed, we restate the aggregation conditions as

$$
\begin{align*}
W^{\prime} E N & =E_{a}  \tag{28}\\
D^{\prime} A^{\prime} W & =A_{a}^{\prime}  \tag{29}\\
W^{\prime} A D & =A_{a}  \tag{30}\\
D^{\prime} Q D & =Q_{a} \tag{31}
\end{align*}
$$

This set of relations provides the fundamental guidelines for constructing consistent aggregate structures. In order to compare the generality of these conditions with those suggested by Guccione and Oguchi notice that, for them, $E=0, E_{a}=0, Q=0, Q_{a}=0$; the operators $D$ and $W$ are composed of identity submatrices stacked atop one another, and the matrix A is block diagonal with identical submatrices along the diagonal. None of these restrictions are implied by the above development. Finally, notice that exact aggregation of dual variables in quadratic programs is also achieved.

## An Example of Exact Aggregation Under Constrained Consistency

Tables 1 and 2 present a numerical illustration of the consistent aggregation outlined in this paper. The example hypothesizes a block diagonal matrix of the microtechnology A with unequal submatrices along the diagonal. We assume that the two submatrices $A_{1}$ and $A_{2}$ represent the technologies of two firms (regions, sectors) of different dimensions. The first activity of each firm is assumed to correspond to a homogeneous commodity whose quantities are, therefore, aggregated in terms of their original units (the weights of the G matrix are unitary). The first resources (land) of each firm are imperfectly homogeneous but it is desirable to aggregate them into a single measure. Hence, it is necessary to transform the original measures of the resources into efficiency units. For this reason, the nonzero weights of the $W$ matrix are not unitary. The programing problems considered in this example are asymmetric quadratic programs with different matrices $Q_{1}$ and $Q_{2}$ corresponding to the quadratic forms in the objective functions of the two firms.

It is also of interest to show the existence of multiple sets of aggregation operators, as stated above. The structure of such alternative matrices can be judiciously chosen for the purpose of allowing greater

TABLE 1: Aggregation of Quadratic Programs Under Constrained Consistency

|  | $A_{k}^{*}, k=a, 1,2$ | Aggregate |  | Firm 1 <br> 49/36 | Firm 2 |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | $\begin{aligned} & 2 \\ & 3 \\ & \hline \end{aligned}$ | $\begin{gathered} 3 / 2 \\ 4 \\ \hline \end{gathered}$ |  | $\begin{aligned} & 49 / 10 \\ & 21 / 5 \\ & \hline \end{aligned}$ | $\begin{aligned} & 12 / 5 \\ & 16 / 5 \\ & \hline \end{aligned}$ |
| Set Number | $\mathrm{Q}_{\mathrm{k}}, \mathrm{k}=\mathrm{a}, 1,2$ | $\frac{337}{2800}$ | $\frac{3}{40}$ | 359/3600 | 5/16 | 21/160 |
|  |  | $\frac{3}{40}$ | $\frac{1}{10}$ | --- | 21/160 | 1/10 |
| 1 | $c_{k}, \mathrm{k}=\mathrm{a}, 1,2$ | 2 | 31/12 | 1/24 | 111/32 | 31/12 |
|  | $b_{k}, k=a, 1,2$ | 5 | 9 | 13/12 | 254/35 | 36/5 |
|  | $\mathrm{x}_{\mathrm{k}}^{*}, \mathrm{k}=\mathrm{a}, 1,2$ | . 4875 | 1.8844 | . 2089 | . 2787 | 1.8842 |
|  | $\mathrm{y}_{\mathrm{k}}^{*}, \mathrm{k}=\mathrm{a}, 1,2$ | 0.0000 | . 5333 | . 0000 | . 0000 | . 6667 |
|  | $z_{k}^{\star}=r_{k}^{\star}, k=a, 1,2$ | 5.3215 |  | . 0044 | 5.3171 |  |
| 2 | $c_{k}, \mathrm{k}=\mathrm{a}, 1,2$ | 31/9 | 31/12 | 217/216 | 1519/288 | 31/12 |
|  | $b_{k}, \mathrm{k}=\mathrm{a}, 1,2$ | 7 | 12 | 5/3 | 352/35 | 48/5 |
|  | $\mathrm{x}_{\mathrm{k}}^{*}, \mathrm{k}=\mathrm{a}, 1,2$ | 2.8571 | . 8571 | 1.2245 | 1.6326 | . 8571 |
|  | $\mathrm{y}_{\mathrm{k}}^{*}, \mathrm{k}=\mathrm{a}, 1,2$ | 1.3036 | . 0070 | . 5587 | . 8147 | . 0087 |
|  | $z_{k}^{*}=r_{k}^{*}, k=a, 1,2$ | 10.6322 |  | 1.0806 | 9.5516 |  |
| 3 | $c_{k}, \mathrm{k}=\mathrm{a}, 1,2$ | 2 | 31/12 | 1/24 | 111/32 | 31/12 |
|  | $b_{k}, k=a, 1,2$ | 607/28 | 55 | 59/84 | 1676/49 | 44 |
|  | $\mathrm{x}_{\mathrm{k}}^{*}, \mathrm{k}=\mathrm{a}, 1,2$ | . 4866 | 12.5530 | . 2089 | . 2783 | 12.5519 |
|  | $\mathrm{y}_{\mathrm{k}}^{*}, \mathrm{k}=\mathrm{a}, 1,2$ | . 0000 | . 0000 | . 0000 | . 0000 | . 0000 |
|  | $z_{k}^{*}=r_{k}^{*}, k=a, 1,2$ | 16.6993 |  | . 0044 | 16.6949 |  |

TABLE 2: Consistent Aggregation Operators

| Aggregation Operators | Firm 1 |  | Firm 2 |  |
| :---: | :---: | :---: | :---: | :---: |
| $\mathrm{W}_{\mathrm{k}}, \mathrm{k}=1,2$ | 3/7 | 0 | 5/8 | 0 |
|  | -- | -- | 0 | 5/4 |
| $\mathrm{D}_{\mathrm{k}}, \mathrm{k}=1,2$ | 3/7 | 0 | 4/7 | 0 |
|  | --- | --- | 0 | 1 |
| $\mathrm{R}_{\mathrm{k}}, \mathrm{k}=1,2$ | 2/3 | -1/2 | 5/4 | 3/8 |
|  | --- | --- | 0 | 1 |
| $S_{k}, \mathrm{k}=1,2$ | 2/3 | -1/4 | 8/7 | 6/35 |
|  | --- | --- | 0 | 4/5 |
| $G_{k}, \mathrm{k}=1,2$ | 1 | 0 | 1 | 0 |
|  | --- | --- | 0 | 1 |
| $\mathrm{P}_{\mathrm{k}}, \quad \mathrm{k}=1,2$ | 1344/1801 | 0 | 1960/1801 | 0 |
|  |  | --- | 0 | 4/5 |

flexibility of the microtechnologies. For example, we hypothesize that the matrix $U^{*}=\left[\begin{array}{ll}S & 0 \\ 0 & R\end{array}\right]$ is an alternative aggregation operator fulfilling the same role as the matrix $U$ previously defined. To be such an operator, the submatrices $S$ and $R$ must satisfy conditions (21) or, more explicitly

$$
\begin{align*}
\mathrm{EW} & =S E_{a}  \tag{32}\\
\mathrm{~A}^{\prime} \mathrm{W} & =R A_{a}^{\prime}  \tag{33}\\
\mathrm{AD} & =S A_{a}  \tag{34}\\
\mathrm{QD} & =R Q_{a} \tag{35}
\end{align*}
$$

and condition (12), or

$$
\begin{align*}
& W^{\prime} S=I_{h}  \tag{36}\\
& D^{\prime} R=I_{k} . \tag{37}
\end{align*}
$$

Notice that if relations (32) and (34) are premultiplied by $W^{\prime}$ and relations (33) and (35) by $D$, the same conditions (28) through (31) are obtained. Table 2 presents a set of aggregation operators which satisfy conditions (32) - (37). Table 1 exhibits the structural elements and optimal primal and dual solutions of the micro and macroproblems relative to three different sets of $b$ and $c$ vectors. These values were selected to represent the following cases: (a) boundary solution for firm 2 and interior solution for firm 1; (b) boundary solution for both firms; (c) interior solution for both firms. It can be verified that the aggregation is exact in all three cases for both primal and dual variables.

## gg

## References

[1] R. W. Cottle, "Symmetric Dual Quadratic Programs", Quarterly of Applied Mathematics, 21 (1963) 237-243.
[2] A. Guccione and N. Oguchi, "A Note on the Necessary Conditions for Exact Aggregation of Linear Programing Models", Mathematical

Programming, 12 (1977) 133-135.
[3] M. Hatanaka, "Note on Consolidation Within a Leontief System", Econometrica, 20 (1952) 301-303.
[4] Y. Ijiri, "Fundamental Queries in Aggregation Theory", Journal of The American Statistical Association, 66 (1971) 766-782.

$$
\begin{aligned}
& \cdot \\
& \cdot \\
& \cdots
\end{aligned}
$$

