OPTIMAL ATTACK AGAINST AN AREA DEFENSE
PROTECTING MANY TARGETS

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ABSTRACT

Many separated targets, of only a few differing values, are subject to a simultaneous attack. The area defenses considered have (a) impact point prediction (IPP) and full coordination, or (b) no IPP and full coordination, or (c) no IPP and partial coordination. For a given attack, the defense wishes to allocate its interceptors to maximize the expected total survival value of the targets. For a given attack size, and with knowledge of the defense's capabilities, the offense seeks a strategy to minimize expected total survival value against best defense. We present algorithms to determine optimal attack and defense strategies and the optimal value of the min-max problem, and we show how to take computational advantage of the relatively few unique target values. Illustrative computational results are provided.
1. INTRODUCTION

Consider an area to be defended that consists of T separated point targets, and associate with
target \( i \) a value \( v_i, \ i = 1, \ldots, T \). Without loss of generality, we assume that the \( v_i \) satisfy \( v_1 \geq v_2 \geq \ldots \geq v_T > 0 \). We denote the number of attackers directed against target \( i \) by \( a_i \) and the attack against the T targets by \( \mathbf{a} = [a_1, \ldots, a_T] \), with \( \sum_{i=1}^{T} a_i = A \) and each \( a_i \in \mathbb{N} \), the set of nonnegative integers. The defense has D interceptors, each of which may be assigned to intercept any attacker in a
given subset, possibly the entire set, of attackers. The objective of the defense is to allocate its
interceptors against attackers so as to maximize expected total survival value. In his first move, the
attacker seeks an attack allocation which will minimize expected total survival value against best
defense.

The specific assumptions we make are as follows:

1. The attack is simultaneous, i.e., the defense sees the overall size of the attack, \( A \).
2. The defense has the “last move.” Only after the attack is observed does the defense
   need to allocate its D interceptors.
3. All A attackers are identical. An attacker will penetrate to its intended target if
   not successfully intercepted. Such a penetrator will destroy the target with probability \( \alpha \), \( \alpha \in (0, 1] \).
   If multiple attackers penetrate to a target, they act independently.
4. All D defenders (interceptors) are identical. An interceptor will destroy the
   attacker at which it is directed with probability \( \delta \), \( \delta \in (0, 1] \). If several defenders are aimed at a single
   attacker, they act independently.
5. Once an attack or defense is allocated, there is no adjustment or modification. For
   example, attackers may not be diverted from their initial destinations nor may the defense use a
   “shoot-look-shoot” option.
6. The attacker knows the overall size of the defense, \( D \).
7. The adversaries have common knowledge of the target values.
(8) If the defense has impact point prediction (IPP), it correctly identifies the intended target of each attacker.

(9) If the defense is (fully) coordinated, its resources are allocated by a single controller. If different subsets of the interceptors are allocated by independent controllers, we say the defense is partially coordinated.

(10) The defense has area defenders only; none of the defensive interceptors are point defenders assigned to protect specific targets.

A brief outline of the paper is as follows. In the remainder of this introductory section we briefly survey the literature dealing with problems of attack and defense of a set of separated point targets. In section 2, we introduce our problem functions and formulate the defender's problem and the attacker's problem associated with each of the three defense capabilities considered: (1) the defense has IPP and is coordinated; (2) the defense has no IPP but is coordinated; and (3) the defense has no IPP and is only partially coordinated. For each of the three cases, we provide (or reference) appropriate algorithms to solve both the defender's problem and the attacker's problem and provide the optimal min-max value. In section 3 we show how the efficiency of each of the three algorithms, for the respective versions of the attacker's problem, can be greatly increased by taking advantage of the relatively few distinct values among the T target values. Section 4 provides illustrative computational results, and section 5 contains concluding remarks.

A good deal of missile allocation work deals with preallocated preferential defense of a set of identically valued targets, using a game theoretic approach by both attacker and defender. Early work on this problem was by Matheson (1966, 1967) and recently Bracken, Brooks and Falk (1987) and Bracken, Falk and Tai (1987) have examined robustness issues connected with the assumption that the defender knows the attack size A. They provide references to the intervening work in this area.

In their seminal work, Eckler and Burr (1972) examined a number of missile allocation
problems under a variety of assumptions and thoroughly surveyed the literature.

For targets of different value, Karr (1981) analyzed several problems under the assumption of sequential attack (the defense sees the attackers one at a time and must make an engagement decision at that time without knowledge of the overall attack size) and developed continuous (i.e., noninteger) solutions on the basis of Prim-Read deployments.

There is a body of work in which the defender’s objective is to minimize the cost of the defense while insuring prescribed levels of survival value as a function of attack size. Under the assumption of a linear bounding function, Burr, Falk and Karr (1985) generated integer solutions for a defense against a sequential attack. Falk (1985b) extended this concept to a larger class of bounding functions, and McGarvey (1987) added an overview with mixed strategies. Karr (1982) and Soland (1987a) developed solutions for defenses against simultaneous attacks.

Finn (1986) examined the value of the additional knowledge provided by IPP for the case of perfect weapons (i.e., $a = \delta = 1$) with an area defense and terminal defenses at the targets.

2. PROBLEM FORMULATIONS AND SOLUTION ALGORITHMS

2.1 Coordinated Defense with IPP

2.1.1 Problem Formulation

If $d_i$ interceptors are allocated for the defense of target $i$ under an attack of size $a_i$, then an even-as-possible distribution of these $d_i$ defenders against the $a_i$ attackers is optimal. [That such a quasi-uniform defense of target $i$ is optimal is shown by Soland (1987b)]. Specifically, for $a_i > 0$, the defense allocates

$$[d_i/a_i] \text{ interceptors to each of } r_i = [d_i - [d_i/a_i]a_i] \text{ attackers}$$

and

$$k_i = [d_i/a_i] \text{ interceptors to each of } a_i - r_i \text{ attackers.}$$

[Notation: We use $[x]$ for the integer part of $x$ and $[x] = \min \{n \geq x \mid n \text{ integer}\}$.]
If we define the probability that a target survives an attacker facing \( m \) defenders as

\[
\pi_m = 1 - \alpha (1-\delta)^m
\]

for \( m = 0, 1, \ldots \),

then, with \( k_i \) and \( r_i \) as defined above, we may write the survival probability at target \( i \) as

\[
\phi(a_i, d_i) = \begin{cases} 
\pi_{k_i}^{a_i-r_i}, & a_i = 1, 2, \ldots, A \\
1, & a_i = 0.
\end{cases}
\]

Let \( d = [d_1, \ldots, d_T] \). We now define the defender's problem for a given attack \( a \) as

\[
\mathcal{P}_1(a) \quad \text{Maximize} \quad \Phi(a, d) = \sum_{i=1}^{T} v_i \phi(a_i, d_i)
\]

subject to

\[
\sum_{i=1}^{T} d_i \leq D
\]

\[
d_i \in \mathcal{N}, \quad i = 1, \ldots, T.
\]

We denote the optimal value of problem \( \mathcal{P}_1(a) \) by \( v[\mathcal{P}_1(a)] \). [Note: For perfect weapons (\( \alpha = \delta = 1 \)), the defender's problem is a 0-1 knapsack problem. Otherwise, it is a nonlinear integer programming problem.] We may then formulate the attacker's problem as

\[
\mathcal{P}_1 \quad \text{Minimize} \quad v[\mathcal{P}_1(a)]
\]

subject to

\[
\sum_{i=1}^{T} a_i \leq A
\]

\[
a_i \in \mathcal{N}, \quad i = 1, \ldots, T.
\]

Inasmuch as our solution algorithm for problem \( \mathcal{P}_1 \) (itself a nonlinear integer programming problem) is of the branch-and-bound type, we would greatly prefer to limit the number of attacks that we need to examine. The following Nonincreasing Attack Theorem (NAT) enables us to do so. The NAT for the case of IPP and perfect weapons was first proved by Falk (1985a), whereas a proof with IPP and imperfect weapons (i.e., \( \alpha < 1 \) and \( \delta < 1 \)) was suggested by Falk (1987) and given by O'Meara (1988). The proof for the case of no IPP and coordinated defense was given by O'Meara and Soland (1988b).
Theorem (Nonincreasing Attack Theorem)

There is an optimal attack, say \( a^* \), with \( a_i^* \geq a_{i+1}^* \), \( i = 1, \ldots, T - 1 \). We call \( a^* \) a nonincreasing attack.

2.1.2 Algorithm for Problem \( \mathcal{P}_1 \)

The algorithm we use to solve problem \( \mathcal{P}_1 \) is a branch-and-bound scheme. During the course of the algorithm, it is often necessary to determine the "best defense" against some given attack \( a \), that is, to solve problem \( \mathcal{P}_1(a) \) (or to determine some bounds on its optimal value). Details of the algorithm are provided by O'Meara and Soland (1988a, 1988c).

2.2 Coordinated Defense with no IPP

2.2.1 Problem Formulation

To avoid any cumbersome notation, we shall examine the cases of perfect weapons (\( \alpha = 1, \delta = 1 \)) and imperfect weapons (\( \alpha < 1, \delta < 1 \)) separately. Results for the mixed cases follow without difficulty.

From the point of view of the attack, with perfect weapons there is a logical constraint of allocating no more than \( D + 1 \) attackers against any target since that number will assure target destruction. The assumption of no IPP means that the defense has no real decision since all the attackers "look the same." Accordingly, the defense allocates its \( D \) defenders against \( D \) of the attackers. Since the weapons are perfect, \( (A - D) \) attackers will penetrate to their intended targets. For nontrivial problems we have \( D + 1 \leq A < T(D + 1) \).

We observe that target \( i \) survives only if all \( a_i \) attackers are intercepted. To determine this probability, we consider the \( D \) attackers to be intercepted as a sample drawn at random (without replacement) from the population of \( A \) attackers. If we think of this as a population dichotomized into attackers destined for target \( i \) and those that are not, a hypergeometric distribution is appropriate.
Accordingly, we define the survival probability at target \( i \) against attack \( a_i \) as \( \psi(a_i) \) and compute it as

\[
\psi(a_i) = \begin{cases} 
\left( \frac{a_i}{A - a_i} \right) \left( \frac{D - a_i}{D} \right), & a_i = 0, 1, \ldots, D \\
0, & a_i > D.
\end{cases}
\]

We then compute the expected survival value at target \( i \) as

\[ v_i \psi(a_i) \]

and the expected total survival value for attack \( a \) as

\[ \Psi(a) = \sum_{i=1}^{T} v_i \psi(a_i). \]

With imperfect weapons, any number of attackers may be directed against any target, and any number of interceptors may be allocated against any attacker. Of course, the overall attack and defense size constraints are still active.

As above, the lack of impact point prediction means that the attackers appear identical to the defense. Intuition suggests that the best defense is to spread the interceptors as evenly as possible among the attackers. This has been shown by Soland (1987b) to be an optimal defense in that it stochastically minimizes the number of attackers that penetrate the defense. Specifically, the defense allocates

\[ \lceil D/A \rceil \text{ interceptors to each of } R = \lfloor D - \lceil D/A \rceil A \rfloor \text{ attackers} \]

and

\[ \lfloor D/A \rfloor \text{ interceptors to each of } A - R \text{ attackers.} \]

With system reliabilities \( \alpha, \beta \in (0, 1) \), the parameter range for nontrivial problems is \( A > 0 \) and \( D > 0 \).

In this case the uncertainty about survival value can be attributed to both system (un)reliability and the lack of IPP. As might be expected, the survival probability function has a more
involved structure. For target not under attack, we have \( \psi(a_i) = 1 \). For \( a_i > 0 \), conditioning arguments [see O’Meara and Soland (1988b)] yield the following:

(i) if \( D < A \),

\[
\psi(a_i) = \frac{\sum_{l=0}^{\bar{n}} (1-\alpha)^l \sum_{n=\hat{n}}^{\bar{n}} \left( a_i - n \right) (1-\delta)^{l-n} \delta^{a_i-l} \left( \frac{a_i}{A-D-n} \right) \left( \frac{A-a_i}{A-D} \right)}{\sum_{m=0}^{\bar{m}} \sum_{n=\hat{n}}^{\bar{n}} \left( \frac{a_i}{A-D-n} \right) (1-\delta)^{l-m} \left( \frac{a_i}{A-D-n} \right) \left( \frac{A-a_i}{A-D} \right)}
\]

where \( \bar{n} = \max \{ a_i - D, 0 \} \) and \( \bar{n} = \min \{ l, A - D \} \);

(ii) if \( D \geq A \),

\[
\psi(a_i) = \frac{\sum_{l=0}^{\bar{n}} (1-\alpha)^l \sum_{m=\hat{m}}^{\bar{m}} \left( \frac{a_i}{A-D-n} \right) (1-\delta)^{l-m} \left( \frac{a_i}{A-D-n} \right) \left( \frac{A-a_i}{A-D} \right)}{\sum_{m=0}^{\bar{m}} \sum_{n=\hat{n}}^{\bar{n}} \left( \frac{a_i}{A-D-n} \right) (1-\delta)^{l-m} \left( \frac{a_i}{A-D-n} \right) \left( \frac{A-a_i}{A-D} \right)}
\]

where \( \bar{n} = \max \{ a_i - R, 0 \} \) and \( \bar{n} = \min \{ a_i - (l - m), A - R \} \).

We formally define the attacker’s problem as

\[
\Phi_2 \quad \text{Minimize} \quad \Psi(a) = \sum_{i=1}^{T} v_i \psi(a_i)
\]

subject to

\[
\sum_{i=1}^{T} a_i \leq A \]

\( a_i \in \mathcal{N}, \quad i = 1, \ldots, T. \)
While it is fairly straightforward to show that, for perfect weapons, problem $\mathcal{P}_2$ is a convex programming problem in the integer sense that the function $\psi$ has nondecreasing increments, the complexity of $\psi$ for imperfect weapons has not yet permitted a similar theoretical assertion. However, all computational experience thus far has not uncovered a nonconvex case. The algorithm we use (see below) is of the "reduction" type for convex problems. A dynamic programming algorithm is also available for any nonconvex instances, if necessary. [See O'Meara and Soland (1988b)].

2.2.2 Algorithm for Problem $\mathcal{P}_2$ ($\psi$ Convex)

Lafon and Lahrichi (1988) describe the following algorithm for a class of problems containing problem $\mathcal{P}_2$ when $\psi$ has nondecreasing increments, that is, when

$$
\psi(a_i + 1) - \psi(a_i) \geq \psi(a_i) - \psi(a_{i-1}), \quad a_i = 1, 2, ..., A.
$$

Algorithm [Lafon and Lahrichi (1988)]:

Step 0 (Initialization):

$$
a_i = 0, \quad i = 1, ..., T; \quad A_1 = A;
$$

$$
k = 1; \quad \text{go to step } k.
$$

Step $k$ (Allocation):

If $A_k = 0$, stop ($a$ is optimal for problem $\mathcal{P}_2$);

else $x_k = \lceil A_k / T \rceil$;

$$
i^* = \min \left\{ i \mid v_i[\psi(a_i + x_k) - \psi(a_i)] = \min_{j=1, ..., T} \{ v_j[\psi(a_i + x_k) - \psi(a_j)] \} \right\}, \quad i = 1, ..., T;
$$

$$
a_i^* := a_i^* + x_k;
$$

$$
A_{k+1} := A_{k+1} - x_k;
$$

$$
k := k + 1; \quad \text{go to step } k.$$

8
Observe that the allocation \( a = [a_1, \ldots, a_T] \) at the start of any step \( k \geq 1 \) solves

\[
\text{Minimize } \sum_{i=1}^{T} v_i \psi(a_i) \\
\text{subject to } \\
\sum_{i=1}^{T} a_i = A - A_k \\
a_i \in \mathbb{N}, i=1, \ldots, T.
\]

### 2.3 Partially Coordinated Defense with no IPP

#### 2.3.1 Problem Formulation

For the case of partial coordination and no IPP, suppose the \( D \) interceptors are divided into \( N \) (battle) groups of \( n \) interceptors each, so \( Nn = D \). Suppose each group of \( n \) interceptors is fully coordinated in the sense that it is centrally controlled. We assume, however, that different groups act independently and do not communicate with each other.

Suppose each of the \( N \) groups is physically able to engage each of \( m \) different attackers with each of its \( n \) interceptors, where \( 1 \leq m \leq A \).

Assume, finally, that each attacker may be defended against by \( r \) different groups, and thus by a total of \( nr \) interceptors. Hence, there are \( Anr \) possible pairings of attackers and interceptors. This number is also equal to \( mD \), since each interceptor can be allocated to any one of \( m \) different attackers. Hence \( Anr = mD \), so \( n/m = D/rA \). For example, \( A=D=1000 \), \( n=10 \) and \( m=20 \) lead to \( r=2 \).

Now consider an arbitrary one of the \( a_i \) attackers directed at target \( i \), and consider one of the \( r \) groups of interceptors able to fire at this attacker. Since the defense lacks IPP, the controller of this group can do no better than to spread his \( n \) interceptors as uniformly as possible among the \( m \) attackers he can fire at; this stochastically minimizes [see Soland (1987b)] the number of attackers (out of the \( m \)) that, in the eyes of this controller, penetrate his defense. Thus the controller selects \( n - m\lfloor n/m \rfloor \) of the \( m \) attackers to receive \( \lfloor n/m \rfloor = \lceil \eta \rceil \) interceptors each, where \( \eta = n/m = D/rA \), and
m - n + m\lfloor n/m \rfloor of the m attackers to receive \lfloor \eta \rfloor interceptors each. Our particular attacker therefore survives the defense of this group with probability

$$\rho = (1 - \delta)^{\lfloor \eta \rfloor} \frac{(n - m\lfloor n/m \rfloor)}{m} + (1 - \delta)^{\lfloor \eta \rfloor} \frac{(m - n + m\lfloor n/m \rfloor)}{m}$$

$$= (1 - \delta)^{\lfloor \eta \rfloor} (\eta - \lfloor \eta \rfloor) + (1 - \delta)^{\lfloor \eta \rfloor} (1 - \eta + \lfloor \eta \rfloor),$$

and it survives the defenses of all \( r \) groups that can fire upon it with probability \( \rho^r \) because the groups act independently.

The probability our particular attacker fails to destroy target \( i \) is then

$$\gamma = 1 - \alpha \rho^r,$$

where, it turns out, \( \gamma < 1 \) if \( \delta < 1 \) and/or \( D/\tau A < 1 \).

It is certainly not true that the \( a_i \) events \{attacker \( j \) fails to destroy target \( i \}\}, \( j = 1, \ldots, a_i \), are independent. However, if the \( a_i \) attackers are cross-targeted (i.e., launched from different places), and if \( a_i/A \) is small, then, to a very good approximation, these \( a_i \) events may be treated as independent. This is akin to approximating sampling without replacement by sampling with replacement.

Using the approximation just described, we conclude that target \( i \) survives with probability \( \gamma^{a_i} \). The total expected survival value for attack \( a \) is then \( \Gamma(a) = \sum_i v_i \gamma^{a_i} \), so that the attacker's problem against a defense with partial coordination is formulated as

\[ \text{Minimize } \Gamma(a) = \sum_{i=1}^{T} v_i \gamma^{a_i} \]

subject to

\[ \sum_{i=1}^{T} a_i \leq A \]

\[ a_i \in \mathbb{N}, \quad i=1, \ldots, T. \]

For \( \gamma \in (0, 1) \) the function \( \gamma^{a_i} \) for \( a_i \in \mathbb{N} \) is easily shown to be strictly decreasing with strictly increasing increments. This is sufficient to conclude that the NAT holds for problem \( \mathcal{P}_3 \).
Note that \( r \), rather than \( N \), is the appropriate measure of the defense's partial coordination and that \( m \) and hence \( r \) depend on geometrical considerations. Since \( \gamma^A \) is a decreasing function of \( r \), larger values of \( r \) indicate less defensive coordination; \( r=1 \) indicates perfect defensive coordination and yields solutions to \( \mathcal{P}_3 \) that are almost identical to the corresponding solutions to \( \mathcal{P}_2 \). The range of meaningful values of \( r \) is \( 1 \leq r \leq D \).

2.3.2 Algorithm for Problem \( \mathcal{P}_3 \)

The principal motivation behind the algorithm for Problem \( \mathcal{P}_3 \) is that we can take advantage of the exponential property of the survival probability function. The change in expected survival value at target \( i \) when the attack changes from \( a_i \) to \( a_i + \Delta \) is \( v_i \gamma^A(\gamma^\Delta - 1) \), \( \Delta = 1, 2, \ldots \). The algorithm is of the "greedy" type. That is, the allocation of each successive attacker is made to that target offering the greatest immediate decrease in expected survival value, given the allocation to that point.

Algorithm:

Step 1 (Initial Phase):

\[
a_1 = 0, \; t=1, \ldots, T;
\]

\[
determine \; T' = \max_{t=1,\ldots,T} \left\{ t \mid \sum_{i=1}^{t-1} \left[ 1 + \ln(v_i/v_1)/\ln(\gamma) \right] \leq A \right\};
\]

\[
a_i = \left[ 1 + \ln(v_i/v_i)/\ln(\gamma) \right], \; i = 1, \ldots, T'; \quad A' = A - \sum_{i=1}^{T-1} a_i;
\]

\[
a_{T'} = \min\{1, A'\}; \quad A' = A' - a_{T'};
\]

if \( A' = 0 \), stop (\( a \) is optimal for \( \mathcal{P}_3 \)); else go to step 2.

Step 2 (Intermediate Phase):

\[
a_i = a_i + \left[ A'/T' \right], \; i=1, \ldots, T'; \quad A' = A' - T'[A'/T'];
\]

if \( A' = 0 \), stop (\( a \) is optimal for \( \mathcal{P}_3 \)); else go to step 3.
Step 3 (Final Phase):

Let \( i_1, \cdots, i_T \) be a reordering of \( 1, \cdots, T' \) such that

if \( 1 \leq m < n \leq T' \), then \( v_m \gamma^{a_m} \geq v_n \gamma^{a_n} \);

\[ a_k := a_k + 1, \quad k = 1, \cdots, i_A; \]

stop \((a)\) is optimal for \( \mathcal{P}_3 \).)

3. GROUPS OF EQUAL-VALUED TARGETS

If there are relatively few distinct values among the target values \( v_1, \cdots, v_T \), the algorithms of section 2 can be modified to run much faster. The basis for the increase in speed is the following

**Quasiuniform Attack Theorem (QAT).**

**Theorem** (Quasiuniform Attack Theorem)

With an appropriate survival function (including those of problems \( \mathcal{P}_1, \mathcal{P}_2 \) and \( \mathcal{P}_3 \)) and equal-valued targets, there is a nonincreasing optimal attack \( a^* \) with the property \( 0 \leq a^*_1 - a^*_T \leq 1 \).

Proofs of the QAT for problems \( \mathcal{P}_1 \) and \( \mathcal{P}_2 \) are given by O'Meara (1988). The proof for problem \( \mathcal{P}_3 \) follows directly from the convexity of the survival probability function \( \gamma^{a_1} \).

3.1 Modification of the Algorithm for Problem \( \mathcal{P}_1 \)

To make the necessary branching modification to the algorithm described by O'Meara and Soland (1988a, 1988c), we introduce the following notation. Order the \( G \) different values among \( v_1, \cdots, v_T \) as

\[ v_1 > \cdots > v_g > \cdots > v_G \]

with \( n_g \) targets having value \( v_g, \quad g = 1, \cdots, G \). Furthermore, define

\[ T_g = \{i|v_i = v_g, \quad i = 1, \cdots, T\}, \]

\[ A_g = \sum_{i \in T_g} a_i \text{ and } \bar{A}_g = \sum_{i \in T_g} \bar{a}_i. \]
When a given node $S = \{a | a \leq \alpha \leq \overline{a}\}$ is presented for branching, we now define the index

$$g^* = \min\left\{ g | 0 < \overline{A}_g - A_g = \min\{ \overline{A}_j - A_j > 0 | j \in \{1, \ldots, G\}\} \right\}. $$

We then define the limiting aggregate values of $S^L$ as

$$\overline{A}_g^L = \overline{A}_g \forall g = 1, \ldots, G \text{ and } \overline{A}_g^R = \begin{cases} \lfloor (\overline{A}_g + A_g)/2 \rfloor, & g = g^* \\ \overline{A}_g, & g \neq g^* \end{cases}$$

and the limiting aggregate values of $S^R$ as

$$\overline{A}_g^R = \overline{A}_g \forall g = 1, \ldots, G \text{ and } \overline{A}_g^R = \begin{cases} \lfloor (\overline{A}_g + A_g)/2 \rfloor + 1, & g = g^* \\ \overline{A}_g, & g \neq g^* \end{cases}.$$

To maintain nonincreasing limiting attacks, we tighten according to the following rules. In succession, for $g = g^*, \ldots, G$, we tighten

$$\overline{A}_g^L := \min\left\{ \overline{A}_g^L, \frac{\overline{A}_g}{n_g} \right\},$$

and for $g = g^* - 1, \ldots, 1$, we tighten

$$\overline{A}_g^R := \max\left\{ \overline{A}_g^R, \frac{\overline{A}_g}{n_g} \right\}.$$

Define $N_1 = 0$ and $N_g = \sum_{i=1}^{g-1} n_i$ for $g = 2, \ldots, G$. Then with $\hat{i}_g \equiv \overline{A}_g - n_g \lfloor \overline{A}_g/n_g \rfloor$, we may now write the limiting attacks of node $S^L$ as

$$\underline{a}^L = a$$

and

$$\overline{a}_{N_g+i}^L = \begin{cases} \lfloor \overline{A}_g/n_g \rfloor, & i \leq \hat{i}_g^* \\ \lfloor \overline{A}_g/n_g \rfloor, & i = 1, \ldots, n_g; \ g=1, \ldots, G \end{cases}.$$
With \( i^* = \Delta_s - n_s[\Delta_s/n_s] \), the limiting attacks of node \( S^R \) are

\[
\overline{a}^R = \overline{a}
\]

and

\[
\overline{a}^R_{N_{s+1}} = \begin{cases} 
[\Delta_s/n_s], & i \leq i^*_s, \\
[\Delta_s/n_s], & i > i^*_s
\end{cases}, \quad i=1, \ldots, n_s; \quad g=1, \ldots, G.
\]

3.2 Modification of the Algorithm for Problem \( \mathcal{P}_2 \)

In the algorithm of section 2.2.2, each iteration of the allocation step (step \( k \)) caused only the attack at the "critical" target \( i \) to be changed. In the modified algorithm below, a "critical" group is identified with an allocation to all targets within that group. This faster consumption of resources (attackers) reduces the number of time-consuming probability calculations that needs to be made. We use the same group notation as introduced above.

Algorithm for Problem \( \mathcal{P}_2 \) with Groups (\( \psi \) convex):

Step 0 (Initialization):

\[
a_i = 0, \quad i=1, \ldots, T; \quad A_1 = A;
\]

\[
k = 1; \quad \text{go to step } k.
\]

Step \( k \) (Allocation):

If \( A_k = 0 \),

stop (\( a \) is optimal for problem \( \mathcal{P}_2 \));

else

\[
x_k = [A_k/T];
\]

\[
g^* = \min \left\{ g \mid \psi \left( a_{N_{s+1}} + x_k \right) - \psi \left( a_{N_{s+1}} \right) \right\} = \min \left\{ v_j \left[ \psi \left( a_{N_{j+1}} + x_k \right) - \psi \left( a_{N_{j+1}} \right) \right] \mid j=1, \ldots, G \right\}, \quad g = 1, \ldots, G;
\]
if $n_g^* x_k \leq A_k$ then

$$a_{N_g}^* + i := a_{N_g}^* + i + \Delta_k, \quad i = 1, \ldots, n_g^*;$$

$$A_{k+1} := A_k - n_g^* x_k;$$

$$k := k + 1; \quad \text{go to step } k.$$

else

$$a_{N_g}^* + i := a_{N_g}^* + i + x_k, \quad i = 1, \ldots, A_k - n_g^* (x_k - 1);$$

$$a_{N_g}^* + i := a_{N_g}^* + i + x_k - 1, \quad i = A_k - n_g^* (x_k - 1) + 1, \ldots, n_g^*;$$

stop ($a$ is optimal for $\Phi_2$).

3.3 Modification of the Algorithm for Problem $\Phi_3$

When Problem $\Phi_3$ has groups of equally valued targets, we may modify the algorithm of section 2.3.2 to achieve some computational efficiencies. As in the preceding section, allocation of resources at each step now applies to all targets within a group rather than to a single target. Of course, a group may consist of only one target.

Algorithm for Problem $\Phi_3$ with Groups:

Step 1 (Initial Phase):

$$a_i = 0, \quad i = 1, \ldots, T;$$

determine $G' = \max \left\{ g \mid \sum_{i=1}^{g-1} n_g [1 + \ln \left( \frac{v_g}{v_i} \right) / \ln \gamma] \leq A \right\};$

$$a_{N_g} = \left[ 1 + \ln \left( \frac{v_G}{v_j} \right) / \ln \gamma \right], \quad j = 1, \ldots, n_g; \quad i = 1, \ldots, G' - 1;$$

$$A' = A - \sum_{i=1}^{G'-1} n_i a_{N_g} + 1;$$

$$a_{N_g} = 1, \quad i = 1, \ldots, \min\{A', n_g\};$$

$$A' := A' - \min\{A', n_g\};$$

if $A' = 0$, stop ($a$ is optimal for $\Phi_3$); else go to step 2.
Step 2 (Intermediate Phase):

\[ a_{N_i+j} := a_{N_i+j} + \left[ A' \left/ \sum_{i=1}^{G'} n_i \right. \right] \quad j=1, \ldots, n, \quad i=1, \ldots, G'; \]

\[ A' := A' - \left( \sum_{i=1}^{G'} n_i \right) \left[ A'/\sum_{i=1}^{G'} n_i \right]; \]

if \( A' = 0 \), stop (a is optimal for \( \mathcal{P}_3 \)); else go to step 3.

Step 3 (Final Phase):

Let \( i, \ldots, i_{G'} \) be a reordering of \( 1, \ldots, G' \) such that

if \( 1 \leq j < k \leq G' \), then \( v_j \gamma a_{N_j+1} \geq v_k \gamma a_{N_k+1} \);

\[ g^* = \max_{g=1, \ldots, G'} \left\{ g \mid N_{ig} \leq A' \right\}; \]

\[ a_{N_{ig}+h} = a_{N_{ig}+h} + 1, \quad h = 1, \ldots, n_{ig}, \quad g=1, \ldots, g^* - 1; \]

\[ A' := A' - N_{ig}; \]

\[ a_{N_{ig}+h} := a_{N_{ig}+h} + 1, \quad h=1, \ldots, A'; \]

stop (a is optimal for \( \mathcal{P}_3 \)).

4. COMPUTATIONAL EXPERIENCE

All algorithms, both initial and modified, have been implemented in FORTRAN with no
special compiler options invoked. All computations have been performed on the VAX 11/780 of the

4.1 Problem \( \mathcal{P}_1 \)

The algorithm for problem \( \mathcal{P}_1 \) is the most computationally intense of the three in that it
involves two levels of optimization (both attacker and defender), each of which involves a branch-and-
bound scheme. Accordingly, only problems of modest size have reasonable run times. Table 1,
however, shows that we can consider solving much larger problems in a timely manner when there are
groups of equally valued targets. The CPU times shown in the "singly" column are for the basic
algorithm in which targets are considered individually. The times shown in the "Groups" column are
for the modified algorithm. The problem data are: (T =) 20 targets, (G =) 3 groups (target values:
2@110, 12@60, 64@10), attack and defense reliabilities \( a = \delta = 0.9 \), and equal numbers of attackers
and defenders (A = D).

Table 1. Run Times for Problem \( \mathcal{P}_1 \)

<table>
<thead>
<tr>
<th>A(=D)</th>
<th>CPU Time (secs)</th>
<th>Reduction Factor</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Singly</td>
<td>Groups</td>
</tr>
<tr>
<td>60</td>
<td>844.03</td>
<td>2.53</td>
</tr>
<tr>
<td>70</td>
<td>1178.67</td>
<td>3.30</td>
</tr>
<tr>
<td>80</td>
<td>1637.69</td>
<td>4.10</td>
</tr>
<tr>
<td>90</td>
<td>1425.23</td>
<td>4.75</td>
</tr>
<tr>
<td>100</td>
<td>2091.03</td>
<td>8.15</td>
</tr>
<tr>
<td>Total Time</td>
<td>7176.65</td>
<td>22.83</td>
</tr>
</tbody>
</table>

The significant reduction in CPU time is due to the fact that the attacker's branch-and-bound
scheme operates in G-space in the modified algorithm rather than in T-space, as in the basic
algorithm.

4.2 Problem \( \mathcal{P}_2 \)

For the fully coordinated defense with no impact point prediction, we can solve larger problems
since only the attacker has a real decision on allocating resources. In Table 2 the problem data are:
(T =) 100 targets, (G =) 3 groups (target values: (10@22, 60@12, 30@2), attack and defense
reliabilities $\alpha = \delta = .99$, and equal numbers of attackers and defenders ($A = D$).

Table 2. Run Times for Problem $\mathcal{P}_2$

<table>
<thead>
<tr>
<th>$A (= D)$</th>
<th>CPU Time (secs)</th>
<th>Reduction Factor</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Singly Groups</td>
<td>(Singly ÷ Groups)</td>
</tr>
<tr>
<td>200</td>
<td>35.42 .15</td>
<td>236</td>
</tr>
<tr>
<td>300</td>
<td>112.94 .28</td>
<td>403</td>
</tr>
<tr>
<td>400</td>
<td>262.67 .48</td>
<td>547</td>
</tr>
<tr>
<td>500</td>
<td>506.78 .82</td>
<td>618</td>
</tr>
<tr>
<td>600</td>
<td>866.44 1.30</td>
<td>666</td>
</tr>
<tr>
<td>Total Time</td>
<td>1784.25 3.03</td>
<td>588</td>
</tr>
</tbody>
</table>

Here the reduction in run time can be attributed to the much smaller number of survival probability calculations that need to be made. This is a direct result of the Quasiuniform Attack Theorem.

4.3 Problem $\mathcal{P}_3$

The exponential structure of the objective function in the model formulation for Problem $\mathcal{P}_3$ permits solution of problems much larger than either of the others. Notwithstanding the small run times of the basic algorithm, there is a substantial reduction in run time when advantage is taken of equally valued targets. The data for Table 3 are: fixed attack size of ($A =$) 6000, ($T =$) 100 targets, ($G =$) 3 groups (target values: 10@22, 60@12, 30@2), attack and defense reliabilities $\alpha = .40$, $\delta = .80$, and coordination factor ($r =$) 3.
Table 3. Run Times for Problem \( P_3 \)

<table>
<thead>
<tr>
<th>D</th>
<th>CPU Time (secs)</th>
<th>Reduction Factor</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Singly</td>
<td>Groups</td>
</tr>
<tr>
<td>14,000</td>
<td>.31</td>
<td>.02</td>
</tr>
<tr>
<td>15,000</td>
<td>.27</td>
<td>.02</td>
</tr>
<tr>
<td>16,000</td>
<td>.31</td>
<td>.02</td>
</tr>
<tr>
<td>17,000</td>
<td>.29</td>
<td>.02</td>
</tr>
<tr>
<td>18,000</td>
<td>.29</td>
<td>.02</td>
</tr>
<tr>
<td>Total Time</td>
<td>1.47</td>
<td>.10</td>
</tr>
</tbody>
</table>

5. CONCLUDING REMARKS

In this paper we have emphasized the min-max modeling and, especially, the solution efficiency of the attacker/defender problem. We have done so for three different cases that model alternate degrees of defender capability with respect to impact point prediction and coordination. By tailoring the algorithms to the situation in which only a few different target values are present, we are able to solve problems of realistic size at reasonable cost. Multiple computer runs are thus feasible, and so it becomes practical to examine sensitivity and trade-off issues by solving many problem variations.

One such issue we have already begun to examine is the sensitivity of results to the attacker’s perception of the defender’s capabilities [see O’Meara and Soland (1988d)]. For example, an attack that is optimal against a defender possessing IPP is not, in general, optimal against the same defender lacking IPP. With efficient algorithms for the three cases of defensive capability examined above, it is quite easy to determine the penalties paid by the attacker for misjudging the capability of the defense. The attacker optimization models developed, and the defender optimization model for the case of IPP, can also be used to examine the sensitivity of results to other parameters. For example, the attacker’s
knowledge of $a$, $D$, $\delta$ and/or $r$ may be imprecise. Similarly, the defender's knowledge of $a$, $\delta$ and/or $r$ may be imprecise. Examination of the effects of such imprecision may be carried out in a straightforward manner. It is also possible to compare, as a function of the true state of affairs and the respective states perceived by attacker and defender, the actual and anticipated expected total survival values.

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REFERENCES


