Rutgers Business School
Newark and New Brunswick

# An Introduction to Search Games 

Thomas Lidbetter<br>Department of Management Science and Information Systems<br>Monday 25 th July 2022

## Part I: Isaac's problem and Gal's solution Hide and seek on a network

- Rufus Isaacs (1965) Differential Games
- Shmuel Gal (1979) Search Games with Mobile and Immobile Hider
- Shmuel Gal (2000) On the Optimality of a Simple Strategy for Searching Graphs
- Steve Alpern (2011) A new approach to Gal's Theory of Search Games on Weakly Eulerian networks
- Steve Alpern, Thomas Lidbetter (2020) Search and delivery man problems: when are depth-first paths optimal?


## Search for Immobile Hider on a Network

- Every arc $a$ of a network $Q$ has length $L(a)$
- Total length of $Q$ is $L(Q)=\mu$
- Distance function $d$ on $Q$ is the "shortest path" metric



## The game $\boldsymbol{G}=\boldsymbol{G}(\boldsymbol{Q}, \boldsymbol{O})$

- Pure strategy for Hider (maximizer): a point in $Q$ (not necessarily a node)
- Mixed strategy $h$ for Hider is a distribution over $Q$
- For $A \subseteq Q$, write $h(A)$ for the probability the Hider is in $A$
- Pure strategy for Searcher (minimizer) is a unit speed path $S(t), t \geq 0$ which covers $Q$
- Mixed strategy for the Searcher is a probability distribution over such paths
- The payoff is the search time $T=T(S, H)=\min \{t: S(t)=H\}$
- The function $T$ is only lower-semicontinuous (uniform topology) but the game has a value $V=V(Q, O)$, optimal mixed Searcher strategies and $\varepsilon$ optimal mixed Hider strategies.


## Search higher density regions first

For a fixed $Q$ and Hider distribution $h$, which has a lower expected search time: $X, A, B, Y$ or $X, B, A, Y$ ?


It turns out that the answer depends only on the search density $\rho$ of $A$ and $B$, where

$$
\rho(C)=h(C) / t(C)
$$

and $t(C)=$ time spent in $C$.

## Search higher density regions first

Search density lemma: For a fixed Hider distribution $h$ on a network $Q$, suppose $S_{1}$ is a search of $Q$ that can be written as $X, A, B, Y$ and $S_{2}$ is a search that can be written as $X, B, A, Y$, where $X, A$ and $B$ are disjoint. Then $T\left(S_{1}, h\right) \leq T\left(S_{2}, h\right)$ if and only if $\rho(A) \geq \rho(B)$, with equality if and only if the densities are equal.

Proof: Write $T_{A}$ for the expected time spent to find the Hider in $A$, assuming he is in $A$. Similarly for $T_{B}$. Then

$$
\begin{aligned}
T\left(S_{2}, h\right)-T\left(S_{1}, h\right)= & h(B)\left(t(X)+T_{B}\right)+h(A)\left(t(X)+t(B)+T_{A}\right) \\
& -h(A)\left(t(X)+T_{A}\right)-h(B)\left(t(X)+t(A)+T_{B}\right) \\
= & t(A) t(B)\left(\frac{h(A)}{t(A)}-\frac{h(B)}{t(B)}\right) .
\end{aligned}
$$

## Uniform Hider strategy

A mixed strategy always available to the Hider is the uniform strategy $h=u$ which hides in any subset $A$ of $Q$ with probability proportional to its length, that is $u(A)=L(A) / \mu$.

Lemma: For any $(Q, O)$ and any $S$,

$$
T(S, u) \geq \mu / 2
$$

Hence, $V \geq \mu / 2$.
Proof: By time $t$, max. probability $F(t)$ of finding the Hider is $t / \mu$, so

$$
T(S, u)=\int_{0}^{\infty} 1-F(t) d t \geq \int_{0}^{\mu} 1-\frac{t}{\mu} d t=\mu / 2
$$

## Chinese Postman Tours

- A covering path $S$ is called a tour if it ends back at $O$
- If a tour has minimum length, denoted $\bar{\mu}$, it is called a Chinese Postman Tour
- A tour is called Eulerian if it has length $\mu$ (traverses each edge exactly once)
- An Eulerian tour exists if all nodes have even degree (number of incident edges), in which case $Q$ is called Eulerian and $\mu=\bar{\mu}$



## Chinese Postman Tours

Example


CPT: achfbgdfebi

## Chinese Postman Tours

Lemma: Any CPT of $Q$ satisfies $\bar{\mu} \leq 2 \mu$ with equality only for trees.


Proof:

- Define $Q^{\prime}$ by doubling every arc of $Q$ (add another arc of the same length with the same endpoints)
- All nodes of $Q^{\prime}$ have even degree so there is an Eulerian tour of length $2 \mu$
- This is also a covering tour of $Q$
- If $Q$ is not a tree it contains a circuit (closed path of distinct arcs), whose arcs we do not need to double


## Random Chinese Postman Tours

Definition: Suppose that $S:[0, \bar{\mu}] \rightarrow Q$ is a CPT. Let $S^{r}$ denote its reverse, given by $S^{r}(t)=S(\bar{\mu}-t)$. A Random Chinese Postman Tour (RCPT) $s$ is an equiprobable mix of $S$ and $S^{r}$.

Lemma: Let $s$ be a RCPT on a network $Q$ with root $O$. Then for any $H \in Q$, $T(s, H) \leq \bar{\mu} / 2$. Hence $V \leq \bar{\mu} / 2$.

Proof: Let $t$ be such that $S(t)=H$. Then $T\left(S^{r}, H\right) \leq \bar{\mu}-t$. So

$$
T(s, H)=\frac{1}{2} T(S, H)+\frac{1}{2} T\left(S^{r}, H\right) \leq \frac{1}{2} t+\frac{1}{2}(\bar{\mu}-t)=\bar{\mu} / 2 .
$$

## Bounds on $V=V(Q, O)$ for a general network

Theorem: For any network $Q$ with root $O$, the value $V=V(Q, O)$ of the search game for an immobile Hider satisfies

$$
\frac{\mu}{2} \leq V \leq \frac{\bar{\mu}}{2} \leq \mu
$$

The lower bound is tight if and only if $Q$ is Eulerian. The bound $V \leq \mu$ can only be tight if $Q$ is a tree.

## Equal Branch Density (EBD) Hider Distribution for Trees

Definition: The EBD Hider distribution is concentrated on the leaf nodes and at every branch node the search density of all branches is equal.


## Depth-first search is a best response against the EBD

Lemma: Any depth-first search $S$ is a best response against the EBD distribution, $h$ and has expected search time $T(S, h)=\mu$.

## Proof

(i) Any two depth-first searches $S_{1}$ and $S_{2}$ have the same expected search time because $S_{1}$ can be transformed into $S_{2}$ by successively swapping the order of search of equal density subtrees that share a root.
(ii) If $S$ is any depth-first search and $S^{r}$ is its time reverse search then for any leaf node $v$,

$$
T(S, v)+T\left(S^{r}, v\right)=2 \mu,
$$

so

$$
T(S, h)+T\left(S^{r}, h\right)=2 \mu,
$$

so

$$
T(S, h)=T\left(S^{r}, h\right)=\mu .
$$

(iii) Proof by contradiction that any depth-first search is a best response...

## Depth-first search is a best response against the EBD

(iii) (continued) If a best response $S$ is not depth-first, it must be of the form:


## $\boldsymbol{V}=\boldsymbol{\mu}=\bar{\mu}$ for trees

Theorem: Let $Q$ be a tree with root $O$. Then $V=\mu$.

## Proof:

(i) $V \leq \bar{\mu} / 2=\mu$ (Searcher uses RCPT)
(ii) $V \geq \mu$ (Hider uses EBD distribution)

## Other networks...

Arc-adding lemma: Let $Q$ be a network and let $Q^{\prime}$ be derived from adding an arc $e$ of length $\ell \geq 0$ between points $x$ and $y$ on $Q$. Then

1. $V\left(Q^{\prime}\right) \leq V(Q)+2 \ell$ so $V\left(Q^{\prime}\right) \leq V(Q)$ if we identify $x$ and $y$ (i.e. $\ell=0$ ).
2. If $\ell \geq d_{Q}(x, y)$, then $V\left(Q^{\prime}\right) \geq V(Q)$. Any hiding strategy on $Q$ does just as well on $Q^{\prime}$.

## Other networks...

Arc-adding lemma: Let $Q$ be a network and let $Q^{\prime}$ be derived from adding an arc $e$ of length $\ell \geq 0$ between points $x$ and $y$ on $Q$. Then

1. $V\left(Q^{\prime}\right) \leq V(Q)+2 \ell$ so $V\left(Q^{\prime}\right) \leq V(Q)$ if we identify $x$ and $y$ (i.e. $\ell=0$ ).

Proof: Replace every pure $S$ used in an optimal strategy $s$ by $S^{\prime}$ which follows $S$ until it reaches $x$, then tours $e$, then follows $S$ again.

$$
T(s, z) \leq V(Q)+\ell \text { for } z \in e
$$

and

$$
T(s, z) \leq V(Q)+2 \ell \text { for } z \notin e
$$

## Other networks...

Arc-adding lemma: Let $Q$ be a network and let $Q^{\prime}$ be derived from adding an edge $e$ of length $\ell \geq 0$ between points $x$ and $y$ on $Q$. Then
2. If $\ell \geq d_{Q}(x, y)$, then $V\left(Q^{\prime}\right) \geq V(Q)$. Any hiding strategy on $Q$ does just as well on $Q^{\prime}$.

Proof: Let $h$ be optimal on $Q$. Let $h^{\prime}$ on $Q^{\prime}$ be same as $h$ (don't hide in $e$ ). Note that for $H \in Q$,

$$
T_{Q^{\prime}}\left(S^{\prime}, H\right) \geq T_{Q}(S, H)
$$

where $S$ is like $S^{\prime}$ but replacing $e$ with the shortest path from $x$ to $y$ in $Q$.

## Other networks...

Proposition: The network $Q$ drawn below has $V(Q)=\bar{\mu} / 2$.
$Q$


Proof:
$V\left(Q^{*}\right) \leq V(Q)$ by arc-adding lemma (1).
$V\left(Q^{*}\right) \geq V\left(Q^{* *}\right)$ by arc-adding lemma (2).
But $V\left(Q^{* *}\right)=\bar{\mu} / 2$ by the tree theorem, so
$Q^{* *}$
$\bar{\mu} / 2=V\left(Q^{* *}\right) \leq V\left(Q^{*}\right) \leq V(Q) \leq \bar{\mu} / 2$.


## Weakly Eulerian networks

Definition: A network is weakly Eulerian if it contains a disjoint set of Eulerian networks such that shrinking each to a point transforms the network into a tree.
Equivalently, a network is weakly Eulerian if removing all disconnecting edges leaves a network with only even degree nodes.

Example


## Weakly Eulerian networks

Definition: A network is weakly Eulerian if it contains a disjoint set of Eulerian networks such that shrinking each to a point transforms the network into a tree.
Equivalently, a network is weakly Eulerian if removing all disconnecting edges leaves a network with only even degree nodes.

Example


## Weakly Eulerian networks

Theorem: The value of the search game on a network $Q$ is $\bar{\mu} / 2$ if and only if $Q$ is weakly Eulerian.

Proof $(\Longleftarrow): \quad \bar{\mu} / 2 \geq V(Q)$


## Weakly Eulerian networks

Theorem: The value of the search game on a network $\mathbf{Q}$ is $\bar{\mu} / 2$ if and only if $\mathbf{Q}$ is weakly Eulerian.
$\operatorname{Proof}(\Longleftarrow): \quad \bar{\mu} / 2 \geq V(Q) \geq V\left(Q^{*}\right)$


## Weakly Eulerian networks

Theorem: The value of the search game on a network $\mathbf{Q}$ is $\bar{\mu} / 2$ if and only if Q is weakly Eulerian.
$\operatorname{Proof}(\rightleftharpoons): \quad \bar{\mu} / 2 \geq V(Q) \geq V\left(Q^{*}\right) \geq V\left(Q^{* *}\right)$


## Weakly Eulerian networks

Theorem: The value of the search game on a network $\mathbf{Q}$ is $\bar{\mu} / 2$ if and only if $\mathbf{Q}$ is weakly Eulerian.
$\operatorname{Proof}(\Longleftarrow): \quad \bar{\mu} / 2 \geq V(Q) \geq V\left(Q^{*}\right) \geq V\left(Q^{* *}\right)=\bar{\mu} / 2$
$Q^{* *}$


## The "Three arc" network



If the Searcher successively chooses unsearched arcs at random, then

$$
T(S, H)=\frac{1}{3}(1-x)+\frac{1}{3}(1+x)+\frac{1}{3}(3-x)=\frac{5-x}{3} \leq 5 / 3 .
$$

So $3 / 2 \leq V(Q) \leq 5 / 3$.

## The "Three arc" network



$$
\mu=3, \bar{\mu}=4
$$

Theorem (L. Pavlovic): It is optimal for the Hider to choose $x$ according to the p.d.f.

$$
f(x)=2 e^{-x}, 0<x<\ln 2 \approx 0.693
$$

It is optimal for the Searcher to go to $A$, go distance $y$ towards $O$, back to $A$, to $O$ on another arc, to $A$ on the untraversed arc, where $y$ is chosen according to the c.d.f.

$$
F(y)=\frac{1}{2}+\frac{e^{y}}{4}, 0 \leq y \leq \ln 2
$$

$V=(4+\ln 2) / 3 \approx 1.56$


## Part II: Variations to the model

- Steve Alpern (2010) Search games on trees with asymmetric travel times
- Steve Alpern \& Thomas Lidbetter (2014) Searching a variable speed network
- Steve Alpern \& Thomas Lidbetter (2013) Mining coal or finding terrorists: the expanding search paradigm
- Steve Alpern (2011) Find-and-fetch search on a tree (2011)


## Variable speed network (tree)



- Define EBD using tour times $\tau$ instead of lengths
- Define height $\delta(v)$ of a leaf node $v$ as the difference between the time from $O$ to $v$ and the time from $v$ to $O$.
- Define the incline $\Delta$ as the average height of a leaf node, weighted according to EBD.


## Variable speed network (tree)



Theorem: The value of the variable speed search game is $\frac{\tau+\Delta}{2}$. The EBD is optimal for the Hider and it is optimal for the Searcher to use a probabilitistic "branching strategy".

## Applications of variable speed: 1. Kikuta-Ruckle game

- Like the original Isaacs-Gal game, but the Hider can only hide at nodes and each node $v$ has a search cost $c_{v}$.
- Searcher can either continue without searching a node or pay the search cost to search it.
- Replace search cost of $c_{v}$ with a "variable speed" arc with outward travel time $c_{v}$ and inward travel time 0.



## Applications of variable speed: 2. Find-and-fetch

- Another variation on the classic model, where the Searcher has to return the Hider to the root (eg. search and rescue, foraging)
- Add a variable speed arc to each leaf node $v$ with outward travel time equal $d(0, v)$ and inward travel time equal $-d(0, v)$.


## Applications of variable speed: 3. Expanding search

- Searcher picks a sequence of $\operatorname{arcs} a_{1}, a_{2}, \ldots$ such that $a_{1}$ is incident to the root and each $a_{1}$ is incident to a node already reached.
- Suitable in cases where the cost to retrace your steps is negligible, eg. mining coal, searching for landmines.
- Can also model search with many searchers.
- For trees, this can be modeled by variable speed search: an arc of length $a$ can be replaced by a variable speed arc with outward travel time $a$ and inward travel time 0.


## Part III: Search games with multiple hidden objects

- Hider hides $k$ balls in $n$ boxes
- Cost of searching box $j$ is $c_{j}$
- Searcher looks in boxes one by one till finding all the balls
- Payoff is cost of finding all the balls.

Lemma: The Hider can make the Searcher indifferent between all her strategies by choosing a subset $H$ of $k$ boxes with probability $p^{*}(H)=$ $\frac{\Pi_{i \in H} c_{i}}{S_{k}}$, where $S_{k}=\sum_{|A|=k} \sum_{i \in A} c_{i}$.
All orderings have expected cost

$$
C-\frac{S_{k+1}}{s_{k}},
$$

Where $C=\sum_{j=1}^{n} c_{j}$.


Eg. $(k=3)$ This choice of $H$ is picked with probability proportional to $3 \times 3 \times 2=18$.

Proof: For the ordering $1,2, \ldots, n$, the expected cost of boxes not searched is

$$
\sum_{j=k+1}^{n} c_{j} \sum_{H \in j-1]^{(k)}} p^{*}(H)=\sum_{j=k+1}^{n} c_{j} \sum_{H \in[j-1]^{(k)}} \frac{\Pi_{i \in H} c_{i}}{S_{k}}=\frac{S_{k+1}}{S_{k}} .
$$

Theorem: The value of the game is $V=C-\frac{s_{k+1}}{s_{k}}$. It is optimal for the Searcher to start by opening a subset $H$ of $k$ boxes with probability $p^{*}(H)$ and to open the remaining boxes in a (uniformly) random order. An optimal strategy for the Hider is $p^{*}$.

## Proof:

- Restrict the Searcher to strategies of the form $s_{A}=$ "search all boxes in $A$ then search the remaining boxes in a random order", where $|A|=k$.
- Then payoff of $s_{A}$ against $B$ is same as payoff of $s_{B}$ against $A$ for $|A|=$ $|B|=k$.


Expected search cost $=(7+2+6)$ $+3+1 / 2(3)$

Expected search cost
$=(7+3+2)$
$+6+1 / 2(3)$

## In general



- All boxes in $A$ and $B$ must be searched. Remaining boxes are all searched with the same probability.
- So payoff matrix is symmetric
- Thus Searcher can use strategy $p^{*}$ to make Hider indifferent between all his strategies. Both players indifferent $\Rightarrow$ equilibrium.

