

# COLLABORATIVE SEARCH FOR A PUBLIC GOOD

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# MOTIVATION

- ▶ **group of agents** often searches for possible solutions to a given problem
- ▶ resulting solution, as well as the information gathered during search, are often a **public good**
- ▶ examples of **collaborative search for a public good**:
  - ◇ consumer search
  - ◇ search for investment opportunities
  - ◇ adoption of new technologies
  - ◇ research and development

## MODELING CHOICES

- ▶ I extend the sequential search model of **Weitzman (1979)** to 2 searchers
- ▶ each public good (project) is represented by a **box**:
  - ◊ uncertain **reward** revealed upon paying a **search cost**
- ▶ once the search process is over, the best uncovered project is implemented

## QUESTIONS ASKED

- ▶ What is the optimal **search order** among risky alternatives?
- ▶ What are the **incentives to free ride** on colleague's search efforts?
- ▶ How does collaborative search by a group of people compare to the **(socially optimal) individual search**?

## PREVIEW OF THE RESULTS

- ▶ **search order** and **stopping rule** are **that of a social planner**:
  - ◇ same project is implemented in the end
  - ◇ same information is gathered in the same order
- ▶ there is **delay at each stage of search process**
  - ◇ each agent free rides in hopes that her colleague will pay the search cost
- ▶ overall, collaborative search is **inefficient**, but **preferred by each individual agent** to searching alone

## ► collective experimentation:

- ◇ Bolton and Harris (1999), Keller et al. (2005), Keller and Rady (2010), Bonatti and Rantakari (2016)

**what I do:** consider multi-armed bandit and study the *order* and *stopping rule*

## ► collaboration in teams:

- ◇ Bonatti and Hörner (2011), Campbell et al. (2014), Georgiadis (2015)

**what I do:** agents choose the order in which to search and decide when to stop.

## ► dynamic provision of public goods:

- ◇ Fershtman and Nitzan (1991), Marx and Matthews (1991), Admati and Perry (1991), Compte and Jehiel (2004), Bowen et al. (2019)

**what I do:** study *search* for a public good

## MODEL

## SETUP

- ▶ 2 players:
  - ◇ risk-neutral
  - ◇ maximize expected present value of best uncovered reward (**free recall**)
  - ◇ discount time at exponential rate  $\delta = e^{-r\Delta t}$
- ▶ each period, one player is randomly (with prob. 1/2) **chosen** to perform search
- ▶ game ends if either
  - ◇ there are no options left to search among
  - ◇ players agree **unanimously** to terminate search process



## ACTIONS

- ▶ when player  $i$  is **chosen**, she can
  - ◇ open exactly one box of her choice
  - ◇ do nothing
  - ◇ propose to terminate the game
- ▶ in the latter case, her **opponent** (player  $j$ ) can
  - ◇ accept the offer
  - ◇ reject it

- ▶  $N$  unopened **boxes**. Box  $b_k \equiv (c_k, F_k(\cdot))$ 
  - ◇ contains an uncertain **reward**  $x_k \sim F_k(\cdot)$  (independent)
  - ◇  $c_k$  is **search cost** paid to learn contents of the box
  - ◇ reward is drawn in the following period
- ▶ initially, there is a fallback reward  $z_0 = 0$

- ▶ at each stage, **state**  $s = (z, \mathcal{B}^c)$  of the problem is
  - ◇ **current best option**  $z$ ,
    - e.g. at  $t = 0$  it is  $z_0 = 0$
  - ◇ **set of unopened boxes**  $\mathcal{B}^c$

# MARKOV PERFECT EQUILIBRIUM

- ▶ let  $\Phi_i^{ch}(s)$  and  $\Phi_i^{op}(s)$  be **discounted continuation payoff**, depending on player  $i$ 's role in state  $s$
- ▶ let  $\alpha_i(s) \equiv (\alpha_i^{ch}(s), \alpha_i^{op}(s))$  be a **stationary Markov strategy**

## Theorem

A pair of strategies  $(\alpha_1(s), \alpha_2(s))$  is an **MPE** if  $\forall i, \forall j \neq i, \forall s$

$$\alpha_i^{ch}(s) = \arg \max_{\hat{\alpha}_i^{ch}(s)} \Phi_i^{ch}(s), \quad \alpha_i^{op}(s) = \arg \max_{\hat{\alpha}_i^{op}(s)} \Phi_i^{op}(s)$$

given  $(\alpha_2^{ch}(s), \alpha_2^{op}(s))$  and subject to

$$\Phi_i^{ch}(z, \emptyset) = \Phi_i^{op}(z, \emptyset) = z$$

ONE BOX

## SOCIAL PLANNER: WEITZMAN (1979)

- ▶ **social planner** solves **individual search problem**
- ▶ if there is only one box *left*, the SP opens it iff

$$-c_k + \delta S(z, F_k) \geq z \quad (\text{SP})$$

where

$$S(z, F_k) \equiv \mathbb{E}[\max\{z, x_k\}] = z \int_{-\infty}^z dF_k(z) + \int_z^{+\infty} x dF_k(x)$$

## RESERVATION VALUE OF A BOX

- let  $\bar{z}_k$  solve

$$-c_k + \delta S(\bar{z}_k, F_k) = \bar{z}_k$$

- then,

$$-c_k + \delta S(z, F_k) \geq z \iff \bar{z}_k \geq z \quad (\text{SP})$$

- $\bar{z}_k$  is **reservation value** of box  $b_k$  that  
**contains all relevant information about this box**

### Theorem

Social Planner **opens** box  $b_k$  iff **box is good enough** i.e. when reservation value of this box is higher than the current best option


## 2 AGENTS, 1 BOX: OPPONENT

► when opponent receives a termination offer, he can

◇ **accept**, get  $z$  immediately

◇ **reject**, **eventually** open the box, and get

$$\frac{\delta}{2-\delta} \cdot [-c_k + \delta S(z, F_k)]$$


$$= \frac{1}{2}\delta + \left(\frac{1}{2}\delta\right)^2 + \dots = \frac{\delta}{2-\delta}$$

► offer is **rejected** if and only if

$$z \leq \frac{\delta}{2-\delta} \cdot [-c_k + \delta S(z, F_k)] \iff z \leq z_k^{IR} \quad (IR)$$



## 2 AGENTS, 1 BOX: CHOSEN PLAYER

- ▶ consider problem of **chosen** player
- ▶ if  $z > \bar{z}_k$ , **proposing termination** is strictly dominant
  - ◊ this offer is always accepted since  $z_k^{IR} < \bar{z}_k$
- ▶ if  $z \leq \bar{z}_k$ , chosen player can do better by mixing between
  - ◊ **opening the box**
  - ◊ **doing nothing**

## EQUILIBRIUM IN MIXED STRATEGIES

- ▶ suppose each player, when chosen, **opens box** with prob.  $\pi$  and **does nothing** with prob.  $(1 - \pi)$ .
- ▶ in equilibrium, chosen player must be indifferent btw
  - ◊ **opening herself**:  $-c_k + \delta S(z, F_k)$
  - ◊ **someone opening it in the future**:

$$\frac{\pi\delta}{1 - (1 - \pi)\delta} \cdot \left[ -\frac{c_k}{2} + \delta S(z, F_k) \right]$$
$$= \pi\delta + (1 - \pi)\pi \cdot \delta^2 + (1 - \pi)^2\pi \cdot \delta^3 + \dots = \frac{\pi\delta}{1 - (1 - \pi)\delta}$$

- the search cost is paid half the time in expectation

- ▶  $\pi$  is obtained from the indifference condition

► chosen player

- ◇ if  $z \leq \bar{z}_k$ , **opens the box**  $b_k$  with prob.

$$\pi_k = \begin{cases} \frac{2(1-\delta)}{\delta c_k} [-c_k + \delta S(z, F_k)] < 1 \text{ if } c_k > S(z, F_k) \cdot \frac{2\delta(1-\delta)}{2-\delta} \\ 1 \text{ otherwise} \end{cases}$$

and **does nothing** with prob.  $1 - \pi_k$

- ◇ if  $z > \bar{z}_k$ , proposes to terminate the game

► opponent

- ◇ **accepts** termination proposal if  $z > z_k^{IR}$
- ◇ **rejects** proposal if  $z \leq z_k^{IR}$

- ▶ on equilibrium path, box is opened *eventually* if  $z \leq \bar{z}_k$ 
  - ◊ this is **socially optimal** cutoff
- ▶ for *large* search costs, box is opened with a **delay**
  - ◊ whenever  $\pi_k < 1$ , chosen player is **free riding**
  - ◊ if  $\Delta t$  is time interval between periods, then **expected delay** is  $\Delta t \cdot \frac{1-\pi_k}{\pi_k}$
- ▶ each agent pays search cost half of the time on average

## Corollary

Higher  $\pi$  means less delay

- ▶ for **very low** values of  $c_k$ , there is **no delay** because it is strictly dominant to open box right away
- ▶ otherwise,  $\pi_k(z)$  is **increasing** and convex in  $z$ .
- ▶ **comparative statics**:  $\pi_k(z)$  is **increasing** in the reservation value of the box, i.e. as
  - ◇ search cost  $c_k$  decreases
  - ◇ distribution of rewards gets “better” (in terms of FOSD or MPS)

MANY BOXES

## Weitzman (1979)

- ▶ **selection rule:** if a box is to be opened, it should be that closed box with *highest reservation value*
- ▶ **stopping rule:** terminate search whenever best sampled reward exceeds reservation value of every closed box

- ▶ let  $\bar{z}_k = \max_{b_l \in \mathcal{B}^c} \bar{z}_l$
- ▶ chosen player
  - ◇ if  $z \leq \bar{z}_k$ , **opens the box**  $b_k$  with prob.  $\tilde{\pi}_k \in (0, 1]$  and **does nothing** with prob.  $1 - \tilde{\pi}_k$
  - ◇ if  $z > \bar{z}_k$ , proposes to terminate the game
- ▶ opponent, upon receiving a termination offer
  - ◇ **accepts** termination proposal if  $z > \tilde{z}_k^{IR}$
  - ◇ **rejects** proposal if  $z \leq \tilde{z}_k^{IR}$



- ▶ **search order** and **termination rule** are myopic
  - ◇ only depend on highest reservation value  $\bar{z}_k$
  - ◇ **socially optimal** on equilibrium path
- ▶ **prob. of opening the box**  $\tilde{\pi}_k(s)$  is NOT myopic
  - ◇ can only be estimated numerically
  - ◇ known lower bound  $\pi_k$  (from the one box case)
  - ◇ less than one for large enough search costs  $\implies$  **delay at each stage of the learning process**

► How does the delay change as they search?

◇ the more boxes are opened, **the better the uncovered reward**, so

$z \uparrow \implies \pi \uparrow$ , the delay decreases

◇ the more they search, **the worse boxes are left** so

$\bar{z}_k \downarrow \implies \pi \downarrow$ , the delay increases

## DISCUSSION

- ▶ all results still **hold** if
  - ◇ there are  $N$  players
  - ◇ players alternate or are chosen with unequal probability
  - ◇ there is no explicit option to do nothing
- ▶ results **do not hold** if players value boxes differently:
  - ◇ best uncovered reward is not a public good
  - ◇ they have different discount factors
  - ◇ players have diferent costs of opening the same box

## CONCLUSION

- ▶ this paper examines a model of **sequential search** for a **public good** by a **group of agents**
- ▶ I find that
  - ◇ **search order** and **stopping rule** are **socially optimal**
  - ◇ **delay** occurs **at every stage of the search process** because agents free ride
  - ◇ each agent prefers to search in group rather than by herself

**Thank You!**

► **Bellman equation is**

$$\Phi(s) = \max \left\{ \boxed{z}, \max_{b_k \in \mathcal{B}^c} \left\{ -c_k + \delta \Phi(s^{-b_k}) \right\} \right\}$$

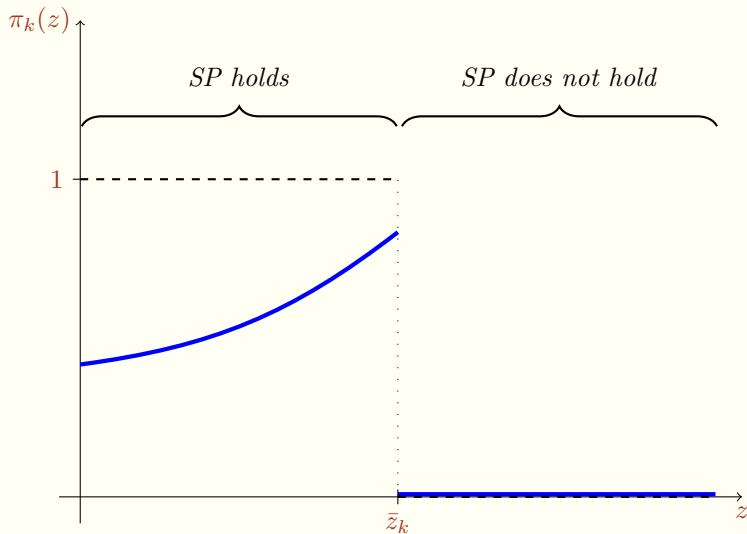
where

◇ payoff if she quits

◇ cont. value of opening the *best* box

- the new state  $s^{-b_k} \equiv \left( \mathbb{E}[\max\{z, x_k\}], \mathcal{B}^c \setminus \{b_k\} \right)$

## PROPERTIES OF $\pi_k$



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# BELLMAN EQUATIONS FOR 2 SEARCHERS

► let  $\bar{\Phi}_i = 1/2\Phi_i^{ch}(s) + 1/2\Phi_i^{op}(s)$  be **average discounted continuation payoff**

► when player  $i$  is **chosen**, her Bellman equation is

$$\Phi_i^{ch}(s) = \max_{\alpha_i^{ch}} \left\{ \alpha_j^{op}(s) \cdot z, \delta \bar{\Phi}_i(s), \max_{b_k \in \mathcal{B}^c} \left\{ -c_k + \delta \bar{\Phi}_i(s^{-b_k}) \right\} \right\}$$

► when player  $i$  is **opponent**, her Bellman equation is

$$\begin{aligned} \Phi_i^{op}(s) &= \max_{\alpha_i^{op}} \left\{ \mathbb{1}_{\{\alpha_j^{ch}(s)=T\}} \cdot r_i \cdot z, \delta \bar{\Phi}_i(s') \right\} \\ \text{s.t. } s' &= \begin{cases} s & \text{if } \alpha_j^{ch}(s) = T, r_i = 0 \text{ or } \alpha_j^{ch}(s) = \emptyset \\ s^{-b_k} & \text{if } \alpha_j^{ch}(s) = b_k \end{cases} \end{aligned}$$