

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

LITERATURE

► 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 $z_0 = 0$

STATE VARIABLES

- ▶ 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. π 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**:

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

$$\rightarrow = \pi\delta + (1 - \pi)\pi \cdot \delta^2 + (1 - \pi)^2\pi \cdot \delta^3 + \dots = \frac{\pi\delta}{1 - (1 - \pi)\delta}$$

• the search cost is paid half the time in expectation

► π 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 Δt 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 π 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 π_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**

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

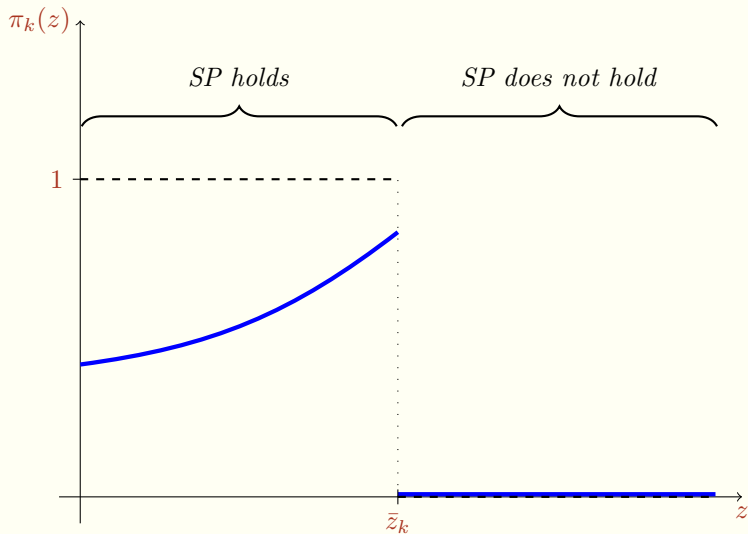
where

◇ 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 π_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

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