

Optimal Search for the Best Alternative

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Outline

- 1 Intro
- 2 Example
- 3 Model
- 4 The Optimal Strategy
- 5 Applications

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- Sources are sampled sequentially, in whatever order is desired.
- When it has been decided to stop searching, only one opportunity is accepted, the maximum sampled reward.
- What sequential search strategy maximizes expected present discounted value?

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- Two substitute technologies are being considered.
- No more than one technology would actually be used even if both were developed.
- Research and development must be done to remove the uncertainty, and it is costly.

Example

TABLE I

Project	α		ω	
Cost	15		20	
Duration	1		2	
Reward	100	55	240	0
Probability	$1/2$	$1/2$	$1/5$	$4/5$

Example

- The expected value of researching α is:

$$-15 + \left(\frac{1}{1.1}\right) \times \left(\frac{1}{2}(100) + \frac{1}{2}(55)\right) = 55.5$$

- The expected value of researching ω is:

$$-20 + \left(\frac{1}{1.1}\right)^2 \times \left(\frac{1}{5}(240) + \frac{4}{5}(0)\right) = 19.7$$

Example

Suppose α is developed first.

- If the payoff turns out to be 55, the expected value of researching ω is:

$$-20 + \left(\frac{1}{1.1}\right)^2 \times \left(\frac{1}{5}(240) + \frac{4}{5}(55)\right) = 56$$

- If the payoff turns out to be 100, the expected value of researching ω is:

$$-20 + \left(\frac{1}{1.1}\right)^2 \times \left(\frac{1}{5}(240) + \frac{4}{5}(100)\right) = 85.8$$

Example

- The expected value of an optimal policy beginning with developing α is:

$$-15 + \left(\frac{1}{1.1}\right) \times \left[\frac{1}{2}(100) + \frac{1}{2} \left(-20 + \left(\frac{1}{1.1}\right)^2 \times \left(\frac{1}{5}(240) + \frac{4}{5}(55) \right) \right) \right] = 55.9$$

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- ω is researched first. (a counter-intuitive property)

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Model

- There are n closed boxes, and $\mathcal{I} = \{1, 2, \dots, n\}$.
- Box i contains a potential reward x_i with pdf $F_i(x_i)$ independent of the other rewards.
- It cost c_i to open box i .
- Box i become know only after a time lag t_i .
- An initially amount x_0 is available.
- All costs and benefits are converted to present values by the discount rate r .

Model

- Let $\mathcal{I}$ be partitioned into any set S such that:
 - $S \cup \bar{S} = \mathcal{I}$
 - $S \cap \bar{S} = \emptyset$
- The variable y will represent the maximum sampled reward:

$$y = \max_{i \in S \cup \{0\}} \{x_i\}$$

- The state of the system at any time is given by the statistic $(\bar{S}, y)$.
- Define $\Psi(\bar{S}, y)$ as the expected present discounted value of following an optimal policy. Ψ must satisfy the fundamental recursive relation:

Model

$$\Psi(\bar{S}, y) = \max \left\{ y, \max_{i \in \bar{S}} \left\{ -c_i + \beta_i \left[\int_{-\infty}^y \Psi(\bar{S} - \{i\}, y) dF_i(x_i) + \int_y^{\infty} \Psi(\bar{S} - \{i\}, x_i) dF_i(x_i) \right] \right\} \right\}$$

- $\Psi(\emptyset, x) = x$
- $\beta_i = e^{-rt_i}$

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Suppose for the moment there are just two boxes (one is the closed box i).

- If the searcher elects not to open box i , she receives:

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- The critical number z_i is called the *reservation price* of box i .

The Optimal Strategy

Proposición 1: Pandora's Rule

SELECTION RULE: *If a box is to be opened, it should be that closed box with highest reservation price.*

STOPPING RULE: *Terminate search whenever the maximum sampled reward exceeds the reservation price of every closed box.*

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First Application

Gambling

- Suppose that box i contains one of two outcomes: either zero reward (“failure”) with probability $1 - p_i$, or positive reward R_i (“success”) with probability p_i .
- There’s no discounting factor ($\beta_i = 1$), and $p_i R_i - c_i > 0$. Then:

$$z_i = R_i - \frac{c_i}{p_i}$$

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- The searcher moves down the list until a success is encountered ;)

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- The optimal policy is to exploit next that unopened mine with maximum gold per probability of machine breakdown, a classic result.