

# Search

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## The Basic Model of Sequential Search

Discrete Time Model

Continuous Time Model

# The Set Up of Discrete Time Model

- ▶ The decision maker samples sequentially from the distribution  $F$ . I.e., he faces a sequence  $\{x_i\}$  of i.i.d. random variables with cumulative distribution function  $F$
- ▶ Each observation costs  $c$ .
- ▶ He can stop search at any time  $n$ .
- ▶ Let  $y_n$  be the prize received if search is stopped after  $n$  samples.

For search with recall,  $y_n = \max\{x_1, x_2, \dots, x_n\}$ .

For search without recall,  $y_n = x_n$

- ▶ Payoff for the decision maker stopping after  $n$  samples is denoted  $Y_n$  where

$$Y_n = y_n - nc$$

If the decision maker discount the the sampling time with discount factor  $\delta < 1$ , then the payoff is

$$Y_n = \delta^{n-1}y_n - \sum_{i=1}^n \delta^{i-1}c = \delta^{n-1}y_n - \frac{1 - \delta^n}{1 - \delta} \cdot c.$$

# The Problem

- ▶ A **stopping rule** prescribes a rule after what sequences of observations to stop the sampling.
- ▶ A **stopping time** resulting from a stopping rule is the integer  $n$  after which the sampling will stop, if that stopping rule is invoked. *The stopping time is a random variable* (because samples are random variables).
- ▶ A stopping rule,  $S$ , determines a random stopping time  $N(S)$ .
- ▶ The problem the decision maker is facing is to find and use a stopping rule,  $S$ , that maximizes its expected payoff,

$$E[Y_{N(S)}].$$

- ▶ The solution to this problem (a stopping rule), is called an **optimal stopping rule**.

# An Optimal Stopping Rule is a Reservation Value Rule

- ▶ A **reservation value rule** is a stopping rule which prescribes to stop sampling after  $n$  observations if and only if  $y_n \geq y$ . The level  $y$  is called the reservation value.
- ▶ It turns out that under certain conditions, a optimal stopping rule is a reservation value rule.

## Proposition

*If  $E[x_i^2]$  is finite, then there exists an optimal stopping rule. It is a reservation value rule.*

See Maurice DeGroot (1970, chapter 13.9) for the proof.

## Some Properties of the Reservation Value Rule (I)

- ▶ Let  $N(y)$  be the stopping time associated with the reservation value rule with reservation value  $y$ . Since probability of  $N(y)$  being  $i$  is  $F(y)^{i-1}(1 - F(y))$ , we can calculate the expected value of  $N(y)$  as follows;

$$E[N(y)] = \sum_{i=1}^{\infty} iF(y)^{i-1}(1 - F(y)) = 1/(1 - F(y)). \quad (1)$$

- ▶ Let  $V(y)$  be the expected value of reservation value rule with reservation value  $y$ . I.e.,  $V(y) \equiv E[Y_{N(y)}]$ . This can be calculated as follows;

$$\begin{aligned} V(y) = E[Y_{N(y)}] &= E[y_n - N(y)c] = E[x|x \geq y] - cE[N(y)] \\ &= \int_y^{\infty} x \frac{dF(x)}{1 - F(y)} - \frac{c}{1 - F(y)}. \end{aligned} \quad (2)$$

# Some Properties of the Reservation Value Rule (II)

## Proposition

*If  $F$  has a connected support, then the optimal policy has a reservation value  $y^*$  such that  $y^* = V(y^*)$ .*

- ▶ The optimal reservation value  $y^*$  solves the equation,

$$c = \int_{y^*}^{\infty} (x - y^*) dF(x) \quad (3)$$

Comparative statics results;

- (1) If the search cost  $c$  is higher, then the reservation value  $y^*$  becomes lower.
- (2) If cumulative distribution function for samples  $G$  is a mean preserving spread of  $F$ , the reservation value under  $G$  is higher than that under  $F$ . ( $G$  is mean preserving spread of  $F$  if (i) they have the same mean and (ii)  $\int_{-\infty}^y G(x) - F(x) dx \geq 0$  for all  $y$ )

# Homework I

## Homework I

- ▶ Prove comparative statics results in the previous slide.
- ▶ Derive expression for  $V(y)$  (equivalent of expression (2)) if decision maker discounts time with discount factor  $\delta < 1$ .
- ▶ Using above result, perform a comparative static of change in  $\delta$  on the optimal reservation value.