

WHEN THE TWO-PERIOD OPTIMAL POLICY IS OPTIMAL  
OVER AN INFINITE HORIZON: A NOTE

Roy Mendelsohn  
Southwest Fisheries Center  
National Marine Fisheries Service, NOAA  
Honolulu, HI 96812

February 1978

DRAFT FOR COMMENT

Problems arising in managing renewable resources, particularly for yield, often take the form:

$$\max E \left\{ \sum_{t=1}^T \alpha^{t-1} P \cdot (x_t - y_t) \right\}$$

$$\text{s.t. } x_{t+1} = s[y_t, D_t]$$

$$0 \leq y_t \leq x_t$$

where  $x$  is the state,  $y$  is the decision, and the  $D_i$ 's are independent, identically distributed random variables. The planning horizon  $T$  may be either finite or infinite.

If  $T$  is finite, Mendelssohn and Sobel (1977) derive the finite dynamic program:

$$f_0(\cdot) \equiv 0$$

$$f_n(x) = \max_{0 \leq y \leq x} \left\{ P \cdot (x-y) + \alpha E f_{n-1}(s[y, D]) \right\} \quad (1)$$

and over an infinite horizon:

$$f(x) = \max_{0 \leq y \leq x} \left\{ P \cdot (x-y) + \alpha E f(s[y, D]) \right\} \quad (2)$$

In a series of recent papers (Mendelssohn 1978a, b, c) I show how to greatly reduce the effort involved in solving (2). In this paper, I show that for a special case, an optimal policy in (1) for  $n = 2$  is optimal in (2).

Consider the following assumption:

- (i)  $f(\cdot)$  is concave, continuous
- (ii)  $E_s[y, D]$  is unimodal and differentiable with respect to  $y$ .

Conditions which are sufficient for (ii) to be valid are given in Mendelssohn and Sobel (1977). Let  $y_2^*$  be the solution to the following equation:

$$E\left\{s^{[1]}[y, D]\right\} = \frac{1}{\alpha} \quad (3)$$

Mendelssohn and Sobel (1977) show that both (1) and (2) have a base stock policy as an optimal policy. That is, there is a  $y^*$  such that an optimal policy is to choose:

$$\text{minimum } (x, y^*)$$

Clearly  $y_2^*$  is the base stock size at  $n = 2$ . Theorem 1 proves  $y_2^*$  is the base stock size in (2) also.

Theorem 1. Assumptions (i)-(ii) imply  $y_2^*$  is an optimal base stock size in (2).

Proof. It is straightforward to show that at  $y^*$ ,

$$E\left\{s[y^*, D]\right\} \geq y^*$$

(see, for example, Mendelssohn and Sobel 1977).

From equation (1), this implies for  $w = E\{s[y^*, D]\}$ :

$$f(w) = p \cdot (w - y^*) + \alpha E f(s[y^*, D])$$

Applying Jensen's inequality yields:

$$p \cdot (w - y^*) + \alpha E f(s[y^*, D]) \leq p \cdot (w - y^*) + \alpha f(Es[y^*, D])$$

which implies:

$$p \cdot (w - y^*) + \alpha E f(s[y^*, D]) \leq p \cdot (w - y^*) + \alpha f(w) = p \cdot (w - y^*) + \alpha E f(s[y^*, D]) \quad (4)$$

Equation (4) implies at  $y^*$ ,  $\alpha E f(s[y^*, D]) = \alpha f(Es[y^*, D])$ .

At  $y = y_2^*$ ,  $p \cdot (w - y) + \alpha f(w)$  achieves a maximum, which implies  $p \cdot (w - y) + \alpha E f(s[y, D])$  also achieves a maximum at  $y_2^*$ . Since a base stock policy is optimal,  $y^* = y_2^*$  is the base stock size.

□

What is convenient about theorem 1 is that equation (3) can be solved on nothing more than a pocket calculator. It also underlines a very real problem in using expected value as a criterion for optimization. That is,  $y_2^*$  is optimal no matter what the variance of  $D$ , so long as the expectation on  $D$  is the same. This suggests that when going from deterministic to stochastic models, the expectation of the deterministic objective most likely is not the proper objective function for the stochastic model. There are several ways around this problem. The first is to use utility theory or other related methods to determine the decisionmaker's attitude towards risk. The second is to include smoothing costs of the form:

$$\begin{aligned} &\varepsilon \left( (x_{t-1} - y_{t-1}) - (x_t - y_t) \right) && \text{if } (x_{t-1} - y_{t-1}) > (x_t - y_t) \\ &\gamma \left( (x_t - y_t) - (x_{t-1} - y_{t-1}) \right) && \text{if } (x_t - y_t) > (x_{t-1} - y_{t-1}) \end{aligned}$$

In a future paper, I will show that for  $\varepsilon = \gamma$ , this is equivalent to weighting the mean return against the variance of the return. By parameterizing on  $\varepsilon$  (therefore  $\gamma$ ), it is then possible to explore the mean-variance tradeoff.

I suspect, but have not been able to prove, that if assumptions (i)-(ii) are valid, then a two-period optimal policy to the smoothing cost problem is a good approximation to a true infinite horizon optimal policy. This will be explored numerically.

References

- Mendelssohn, R. (1978a) Increasing computational efficiency for semi-separable Markov decision processes. U.S. Dep. Commer., Natl. Mar. Fish. Serv., SWFC Admin. Rep. 2H, 1978, 16 p.
- \_\_\_\_\_. (1978b) Recovering optimal policies and qualitative properties of optimal policies for semi-separable Markov decision processes. U.S. Dep. Commer., Natl. Mar. Fish. Serv., SWFC Admin. Rep. 3H, 1978, 7 p.
- \_\_\_\_\_. (1978c) Aggregation and other reductions for semi-separable Markov decision processes: The linear case. U.S. Dep. Commer., Natl. Mar. Fish. Serv., SWFC Admin. Rep. 4H, 1978, 11 p.
- Mendelssohn, R. and M. J. Sobel (1977) Capital accumulation and the optimization of renewable resource models. Submitted to J. Econ. Theory.



**U.S. DEPARTMENT OF COMMERCE**  
**National Oceanic and Atmospheric Administration**  
NATIONAL MARINE FISHERIES SERVICE  
Southwest Fisheries Center  
Honolulu Laboratory  
P. O. Box 3830  
Honolulu, Hawaii 96812

February 23, 1978

Dr. Matthew J. Sobel  
College of Industrial Management  
Georgia Institute of Technology  
225 North Ave., N.W.  
Atlanta, Georgia 30332

Dear Matt:

Enclosed are the revised versions of the papers I sent you. Also included are several new papers which extend these results in specific instances. Your comments, corrections, questions, etc., will be appreciated. If you have some good applications for this stuff let me know. I am trying to find just how well some of this performs.

Keep in touch--spring will come sooner than you know.

With regards,

Roy Mendelssohn  
Operations Research Analyst

Enclosure (SWFC Admin. Rep. 2H, 3H, 4H, 9H, 10H, 1978)

RM:ey  
cc: Mendelssohn  
HL

Identical letter to:

Dr. Paul Zipkin  
Graduate School of Business  
Uris Hall 416  
Columbia University  
New York, N.Y. 10027

Professor Annie Thomas  
Graduate School of Management  
University of Rochester  
Rochester, N.Y. 14627