

# Production planning, activity periods and passivity periods

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# Production planning, activity periods and passivity periods

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## Abstract

Consider a company which produces and sells a certain product on a market with highly variable demand. Since the demand is very high during some periods, the company will produce and create a stock in advance before these periods. On the other hand it costs money to hold a big stock, so that some balance is needed for optimum. The demand is assumed to be known in advance with sufficient accuracy. We use a technique from optimal control theory for the analysis, which leads to so-called activity periods. During such a period the stock is positive and the production is maximal, provided that the problem starts with zero stock, which is the usual case. Over a period of one or more years, there will be a few activity periods. Outside these periods the stock is zero and the policy is to choose production = the smaller of [demand, maximal production]. The “intrinsic time length” is a central concept. It is simply the maximal time a unit of the product can be stored before selling without creating a loss.

**Remarks:** The author realizes that this is a simplified model, since for instance the production cost and the price are both assumed constant. We think nevertheless that the structure theorem (Th.2) and the complete solution of the seasonal problem should be of some interest. The reader may like to take a quick look at the Appendix 2 in the end of this paper.

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# 1. Definition of the problem. Notation and solution concepts.

The company produces and sells its product on a market with variable demand over a fixed time interval  $0 \leq t \leq T$ . The demand  $d(t)$  is supposed to be known in advance. The problem is to determine the production  $u = u(t)$  over the time interval so that the result

$J = \text{income} - \text{costs}$ , becomes maximal. The following parameters and variables will be used:

$t = \text{time}$  ;

$u(t) = \text{production per time unit}$  ,  $0 \leq u \leq U$ , where the maximal production rate  $U$  is constant;  $u$  is the control variable;

$\beta = \text{the production cost for one unit of product}$ , also constant;

$d(t) = \text{demand for product}$ ,  $0 \leq d(t)$ , is the maximal product quantity that can be sold per unit time

$x = x(t) \geq 0$  is the stock of product available at time  $t$ ;

$h > 0$  is the storage cost per time unit and unit of stock

$c_1 = \text{price per unit of sold product}$

$S(x,d,u) = \text{selling function}$ , which means that  $S(x(t),d(t),u(t))$  units are sold per time unit at time  $t$ . Various choices for  $S$  are possible.

All this leads to the "state equation"  $\dot{x} = u(t) - S(x(t), d(t), u(t))$  ;

i.e.: increase of stock per time unit = production minus selling.

The objective functional is therefore given by

$$J = \int_0^T [c_1 S(x(t), d(t), u(t)) - \beta u(t) - h x(t)] dt + c_0 x(T).$$

It is to be maximized by clever choice of the control function  $u(t)$ , when  $x(0)$  is given and  $x(T)$  is free or prescribed. In the case that  $x(T)$  is not prescribed there is a rest value  $c_0$  ascribed to the product. It is understood from the beginning that  $c_0 \leq c_1$  for obvious reasons.

The demand function  $d(t)$ ,  $0 \leq t \leq T$ , is assumed to be "known" and assumed to be piece-wise constant, having a finite number of jump discontinuities.

It is also clear that there can be no selling if there is no demand,  $d(t) = 0$ . As a basic approximation we assume that the selling is proportional to  $d(t)$  for any given  $x > 0$ .

The dependence on  $x$  can certainly be modeled in various ways.

The given problem can be discussed under different mathematical ambitions:

## A. The initial problem formulation

The concrete optimization problem for the company considered is called the initial problem. It normally runs over a period of one or more years. The demand, in particular, may vary considerably for a "seasonal" product during such a period, for example being considered constant each month, or week. It is understood that the control  $u(t)$  should

be piece-wise continuous, and the stock  $x(t)$  should have a piece-wise continuous derivative, allowing for a finite number of discontinuities of the derivative. Clearly,  $x(t)$  is supposed to be non-negative. The selling may simply be the demand  $d(t)$  times some “suitable” function of  $x > 0$ , or some more complicated function of  $d$ ,  $x$  and  $u$ , when  $x=0$ . In this context we are not aiming at complete mathematical rigour.

## B. A strict mathematical solution, satisfying the side condition

Here, a mathematically correct solution  $\{x(\cdot), u(\cdot)\}$  is wanted, satisfying the side condition  $x(t) \geq 0$ , plus restrictions stated below. It will be called a proper solution.

**Remark.** The fact that the function  $d(t)$  in our problem has a finite number of jump discontinuities means no severe difficulty.

In this context the control function  $u(t)$  must be Lebesgue measurable and satisfy  $0 \leq u(t) \leq U$ . The stock  $x(t)$  must be absolutely continuous and satisfy the state equation a.e. The side condition  $x(t) \geq 0$  must be satisfied in the whole domain of definition. The stock  $x(t)$  must satisfy initial and final conditions.

It remains to specify the selling function and the state equation.

It is assumed, for  $x > 0$  only, that the selling function can be written

$S(x, d) = d \varphi(x)$ , where the function  $\varphi(x)$  is continuously differentiable and non-decreasing for  $x \geq 0$ . Further,  $\varphi(0) = 1$ .

Also, for  $x > 0$ , the state equation is simply  $\dot{x} = u(t) - S(x(t), d(t))$ , i.e.

$$\dot{x} = u(t) - d(t) \varphi(x(t)).$$

Define  $E = \{t: x(t) = 0\}$ . Since  $x(t)$  is differentiable a.e., and since almost all points of  $E$  are points of accumulation, it follows that  $\dot{x}(t) = 0$  a.e. on  $E$ .

Thus, a specific formula for  $\dot{x}$  on  $E$  is simply not needed.

It is required, however, that  $\dot{x} = u(t) - \text{selling function} = 0$  on  $E$ . Thus

$\text{selling} = u(t) \leq U$ , and by definition  $\text{selling} \leq d(t)$ .

Consequently,  $u(t) = \text{selling} \leq \min\{U, d(t)\}$  on  $E$ . Clearly,  $\min\{U, d(t)\}$  is the biggest possible value for production and selling at  $t$  that does not increase the stock. It is therefore “locally” optimal management.

The continued analysis will be based on the understanding, or condition that

$$u(t) = \text{selling} \leq \min\{U, d(t)\} \text{ for almost all } t \in E.$$

The results obtained in this paper refer to case B. They can, however, easily be interpreted in the context of case A.

The question of the existence of a proper solution will be resolved in §9.

In the following analysis it is simply assumed that we have an optimal element.

## 2. Using the control maximum principle

In this situation the usual maximum principle (MP) by Boltyanski and Pontryagin in optimal control can be applied, paying due attention to the condition  $x(t) > 0$ . We will adhere to the presentation by Evans [4].

Rigorous versions of the maximum principle are found in [L-M], pp. 318-321, or [M-S], pp. 126-127, but the most useful version for the present text is found in [4] by L.C. Evans, pp. 110 – 118. We will refer to [4] and use the same notation, as far as possible.

### A brief background comment

The standard method for handling side conditions like  $x(t) \geq 0$  is to introduce multipliers for the side conditions, multiply and add to the so-called Hamiltonian function, which gives the Lagrangian. Then a modified maximum principle is supposed to hold for the Lagrangian. Further, complementary slackness conditions enter the picture and make it more complicated. For this more traditional approach, see [S-T], pp. 98 ff. All this is avoided here, thanks to our “activity period” approach.

We are now facing a problem in optimal control of Bolza’s type, over a given time interval. The problem is non-autonomous, since the demand depends on time. It is shown in [4] how the Bolza problem can be rewritten as a problem of Mayer’s type, which is convenient.

It was above assumed that the selling function is written  $S(x, d) = d \cdot \varphi(x)$ , where the function  $\varphi(x)$  is continuously differentiable and non-decreasing for  $x \geq 0$ . Further,  $\varphi(0) = 1$ . In a while it will be assumed that  $\varphi(x) \equiv 1$ , but

$\varphi(x)$  will be kept until further for reference purposes.

Consider for a while an element  $(x(t), u(t))$ , optimal on some interval  $[0, T]$ , such that  $x(t) > 0$  on the whole interval. Some notation must be changed in order to adapt to [4].

First, the notation  $x$  for the stock is changed to  $x_1$ . Next,  $x_2$  will be the integral found in the definition of  $J$ , but now taken from 0 to  $t$ . In other words, we have

$$\dot{x}_1 = u - \varphi(x_1)d(t) \equiv f_1(t, x_1, u),$$

$$\dot{x}_2 = c_1 \varphi(x_1)d(t) - \beta u - h x_1 \equiv f_2(t, x_1, u).$$

Introduce the Jacobian matrix  $A(t)$  as in [4], p. 115:

$$A(t) = \left( \frac{\partial f_i}{\partial x_k} \right) = \begin{bmatrix} -\varphi'(x_1)d(t) & 0 \\ c_1 \varphi'(x_1)d(t) - h & 0 \end{bmatrix}. \text{ Consider the adjoint system}$$

$$\dot{\eta} = -(\eta_1, \eta_2) A(t), \text{ i.e.}$$

$$\dot{\eta}_1 = \eta_1 \cdot \varphi'(x_1) d(t) - \eta_2 \cdot [c_1 \cdot \varphi'(x_1)d(t) - h],$$

$$\dot{\eta}_2 = 0.$$

Here,  $\eta = (\eta_1, \eta_2)$  is the so-called adjoint state variable.

In the case that the end-point is free, then a so-called transversality condition is available, giving  $\eta_1(T) = c_0$  and  $\eta_2(T) = 1$ . (See [4], p. 117.) This may in some cases simplify the continued analysis. It will not be needed in this work.



In the case that the end-point is prescribed, much less information is obtained. One only gets the information that  $\eta_2(T) \geq 0$ . (See [4], p. 123.)

According to the maximum principle (MP), [4] p.116 -118, we form the Hamiltonian function  $H$  from the state equation and the objective functional  $J$  as follows:

$$H = (u - d(t)\varphi(x_1)) \cdot \eta_1 + (c_1 \cdot d(t)\varphi(x_1) - \beta \cdot u - h \cdot x_1) \cdot \eta_2 .$$

Observe here that  $x$  satisfies the dual equation  $\dot{X} = \frac{\partial H}{\partial \eta}$ . In the lucky case that

$\eta_2 > 0$ , we can replace  $\eta_2$  by 1 without losing generality, and then

the only terms in  $H$  which contain  $u$  are  $u \eta_1 - \beta u = u(\eta_1 - \beta)$ . According to MP, this expression is maximized by  $u(t)$  along an optimal trajectory, for almost all  $t$ . Since the control variable  $u$  is restricted by  $0 \leq u \leq U$ , it follows that (except for a null set)

$$u(t) \text{ is } \begin{cases} U & \text{if } \eta_1 > \beta \\ 0 & \text{if } \eta_1 < \beta . \\ \text{unspecified} & \text{if } \eta_1 = \beta \end{cases}$$

All this is in agreement with the “standard” deterministic maximum principle.

For obvious reasons, we make the following general assumption:  $c_1 > \beta > 0$ .

The interpretation of the adjoint variable  $\eta_1$  as a “shadow value” for the state variable  $x_1$  is well known. It works best for the case that the end-point is free, in which it follows from the derivation of the maximum principle from the solution of a Mayer problem. It requires, however, some regularity of the so-called value function  $V(t, x_1)$ , which is not always satisfied. See [4], p.85. This interpretation is nevertheless very helpful for intuitive understanding of the result. The above rule says that if this “shadow value” is higher than the cost of production, then produce at maximal rate; otherwise do not produce at all.

The concept of a switch must be explained. Let  $\eta_1(t) < \beta$  for  $t < t'$ , and

$\eta_1(t) > \beta$  for  $t > t'$ . Then, by the above rule for  $u(t)$ , it follows that  $u(t) = 0$  for  $t < t'$ , and  $u(t) = U$  for  $t > t'$ . This is an off-to-on switch. The meaning of an on-to-off switch is now obvious. The possibility of different behavior of  $\eta_1$  at some  $t$ -value, other than a simple sign change of the crucial quantity  $(\eta_1 - \beta)$  is not a priori excluded.

We want to obtain a better understanding of an optimal control. An important step will be to understand the number and position of switches.

**Definitions:** A demand period is an open interval on the  $t$ -axis, where  $d(t)$  is constant. The interval is always assumed maximal. Other parameters like  $U, h, \beta, c_1$  are also assumed constant during a demand period.

The global problem is the just the previous optimization problem, now considered over an interval consisting of a finite number of demand periods. Other parameters than  $d(t)$  do not differ between the demand periods, unless otherwise is said.

### 3. Business periods. Guiding function

Clearly, the side condition  $x_1 \geq 0$  is an important feature of our problem. Now consider a “candidate” solution element  $x_0(t)$ , not necessarily optimal for the initial problem. Suppose that  $x_0(t) > 0$  on an interval  $L$ , and  $x_0(t) = 0$  at the endpoints of  $L$ . Assume that  $x_0(t)$  is optimal, considered on  $L = [T_0, T_1]$ , with prescribed boundary values zero, and under the restriction  $x_1 > 0$  (except for endpoints). We then say that  $x_0(t)$ , considered on  $L$ , defines a true active period. This will turn out to be a very useful concept. We may later also consider active periods, starting by a positive stock, or active periods ending by a positive stock, or both. But for the moment all active periods considered will be “true”.

The Boltyanski-Pontryagin maximum principle is applicable to  $x_0(t)$ , as was briefly explained in §2. For more details, we refer to [4], Chapter 4, and in particular the Appendix. The control system here has the form ( $n = 1$ )

$$\dot{x}_1 = u - \varphi(x_1)d(t) \equiv f_1(t, x_1, u),$$

$$\dot{x}_2 = c_1 \varphi(x_1)d(t) - \beta u - hx_1 \equiv f_2(t, x_1, u).$$

For technical reasons, first consider  $x_0(t)$  on  $[T_0, T']$  for some  $T' < T_1$ . Clearly,  $x_0(t)$  is optimal over this interval too, and the end-point is not on the boundary of the allowed domain. The maximum principle is clearly applicable.

The basic statement of the principle is found in [4], p.123. The question as to whether the situation is “abnormal” or not will be resolved below. The adjoint variable is here written  $\eta(t) = (\eta_1(t), \eta_2(t))$ , instead of  $p^*(t)$  as in [4].

The perturbation cone  $K(T')$  (in our case just a sector in the plane) is well defined and plays a central role. The unit vector  $e_2$  cannot be interior to the cone  $K$  (see [4], p.122), because that would quickly lead to a contradiction, namely a better element could then be constructed, i.e. an element with the same value for  $x_1$ , and a bigger value for  $x_2$ . Hence, there exists a separating vector  $w = (w_1, w_2)$ , such that  $w \cdot z \leq 0$  for all  $z \in K(T')$ , and  $w \cdot e_2 = w_2 \geq 0$ . The terminal condition for the adjoint vector is  $\eta(T') = w$ , and so  $\eta_2(T') = w_2$ . The maximization statement is the same as in [4], p.123-124, although in slightly different notation. This argument is much the same as in [1], pp. 247-254 (much more in detail). See also [2], pp. 108-109, or [7], pp. 92-107.

Two different cases must be considered:  $w_2 > 0$ , or  $w_2 = 0$  (abnormal case).

1. Let  $w_2 > 0$ . As explained in §2, the adjoint variable  $\eta(t)$  will satisfy the system

$$\dot{\eta}_1 = \eta_1 \cdot \varphi'(x_1) d(t) - \eta_2 \cdot [c_1 \cdot \varphi'(x_1)d(t) - h],$$

$$\dot{\eta}_2 = 0.$$

From now on, we specialize on the case  $\varphi(x) \equiv 1$ .

Divide  $\eta(t)$  by  $w_2$  to get  $\eta_2(t) \equiv 1$ , and simplify the first equation into

$$\dot{\eta}_1 = h, \text{ after observing that } \varphi'(x_1) \equiv 0.$$

Therefore, only a switch “off to on” is a priori possible. But  $x_0(t)$  starts from zero and immediately becomes positive, so the only possibility is that  $u(t) = U$  on the whole  $[T_0, T']$ . Further,  $U > d(t)$  must hold on  $L$  near the left end-point.

2. Let  $w_2 = 0$ . Also here,  $\dot{\eta}_2 = 0$ , and so  $\eta_2(t) \equiv 0$ . Clearly,  $\eta_1(t) \equiv \text{const.} \neq 0$ . The Hamiltonian reduces to  $H = \eta_1(t) \cdot (u - d(t))$ , to be maximized a.e. along the optimal

$x_0(t)$ . This implies  $u(t) \equiv 0$  or  $u(t) \equiv U$  on  $[T_0, T']$ . The case  $u(t) \equiv 0$  is clearly impossible, and thus  $u(t) \equiv U$ .

But  $T'$  was arbitrary, except that  $T_0 < T' < T_1$ .

Consequently  $u(t) \equiv U$  on  $L$  in any case. To summarize:

### Theorem 1

An optimal solution  $x_0(t)$  must satisfy  $u(t) \equiv U$  a.e. during a true active period.

The question is now: does the same result hold for any active period, which starts from zero? Consider a problem, where the end-point is free, and a term  $c_0 \cdot x(T_1)$  is added to  $J$ . Consider an optimal solution  $x_0(t)$ , starting from zero, and ending with  $x_0(T_1) > 0$  (otherwise we are back in the previous case). Then the previous argument is applicable to  $x_0(t)$ , and so  $u(t) \equiv U$  a.e. for  $T_0 \leq t \leq T_1$ . But in this case more information is available, namely certain end-point conditions. These will be considered in §5.

### Theorem 1'

An optimal solution  $x_0(t)$ , starting from zero, must satisfy  $u(t) \equiv U$  a.e. during a final active period, where the endpoint is free.

Also in this case,  $U > d(t)$  must hold on  $L$  near the left end-point.

The equation governing the development of the stock  $x_1(t)$  on  $L$  has the form

$\dot{X} = U - d(t)$ , where  $d(t)$  is non-negative and constant on each demand period, i.e. piece-wise constant. The value of  $d(t)$  at an endpoint of a demand period has no importance. The form of the equation implies that all possible solutions during an active period will be restrictions or "vertically" translated restrictions of one and the same piece-wise linear solution  $X(t)$ , as long as

$x_0(t) > 0$ . The function  $X(t)$  will be called the guiding function of the problem.

Thus  $X(t) = \int_0^t (U - d(s)) ds$ , for  $0 \leq t \leq T$ .

### Corollary of Theorem 1

An optimal solution  $x_0(t)$  is a restriction

(+ a constant) of the guiding function during a true active period.

This will be the key to a numerical solution.

Clearly, an active period for  $x_0(t)$  consists of a finite number of demand periods, or parts thereof, such that  $x_0(t)$  is linear on each demand period. It starts from zero and ends non-negative. In the situation that  $x_0(t)$  ends by a positive value further information can be obtained, as seen in §5. Clearly,  $x_0(t)$  will have jump discontinuities only at endpoints of demand periods. Further,  $x_0(t)$  may start from zero at an interior point of a demand period, or from an end-point, and similarly for the ending.

Before summarizing the results so far: assume that  $x(t) = 0$  on some interval  $L'$ . As was observed in §1, the best our company can do during that time is to choose  $u(t) = \min[d(t), U]$ , i.e. produce and sell what can be sold without creating a stock. We can now summarize:

## Theorem 2

**Main structure theorem:** If  $x(t)$  is optimal on the “global” interval  $L$ , then  $L$  is decomposed into a finite number of activity periods: positive stock,  $u(t) = U$ ; and a finite number of “passive” periods: zero stock,

$$u(t) = \min[d(t), U].$$

It can certainly occur that one of these categories is empty.

## Corollary

If  $x(t)$  is optimal, then  $x(t)$  is piece-wise linear on  $[0, T]$  and its derivative has a finite number of discontinuities. These discontinuities occur at points, where  $d(t)$  has a discontinuity, and at points where an activity period begins or ends.

#### 4. Useful technical results. Inner and outer derivatives.

We begin here by observing that the whole problem is trivial if  $d(t) \geq U$  always, or if  $d(t) \leq U$  always. In any of these two cases the choice

$u^*(t) = \min(U, d(t))$  is clearly optimal. Thus, assume from now on that  $\max(d(t)) > U > \min(d(t))$ .

Consider again the initial problem with arbitrary boundary data. The following simple fact will be useful.

##### Lemma 1

Let  $x^*(t) \geq 0$  be a proper solution to our basic problem. Let

$d(t) < U$  immediately to the left of some point  $t_0$  and  $d(t) > U$  immediately to the right of  $t_0$ . Then  $x^*(t_0) > 0$ .

Proof: This goes by contradiction. Assume that  $x^*(t_0) = 0$ . Then a better element can be constructed. First, it is clear that  $x^*(t) = 0$  must hold on some interval to the right of  $t_0$ , because of the dominating demand. Let  $s > 0$  be a parameter at our disposal. Let  $y(t)$  be an admissible candidate for our optimization problem, obtained by replacing  $u^*(t)$  by  $U$  on the interval  $t_0 - s \leq t \leq t_0$ . Put  $\lambda = y(t_0) > 0$ . Clearly,  $y(t) = x^*(t)$  will hold for  $t > t_0 + a \cdot \lambda$ , for some positive constant  $a$ . Now, if  $s$  goes to 0, then so does  $\lambda$ .

Compare the merits of  $x^*(t)$  and  $y(t)$ . Clearly, the difference in storage cost will be  $o(\lambda)$ . The difference in selling returns will be  $c_1 \cdot \lambda$  in favour of  $y(t)$ . The difference in production cost will be  $\beta \cdot \lambda$  in favour of  $x^*(t)$ . Now, since  $c_1 > \beta$ , it follows that  $y(t)$  is better than  $x^*(t)$ , for  $s$  small enough, contradicting the optimality of  $x^*(t)$ . This completes the proof.

Terminology. Let  $t'$  be a point where  $d(t) - U$  changes sign from strictly negative to strictly positive. Then  $t'$  is called a check-point. (Just to have a convenient name.) So an optimal element is positive at each check-point.

The concept of intrinsic time length for our problem will be useful later. It is defined as  $l_0 = \frac{1}{h}(c_1 - \beta)$ . Clearly, it is a measure of the possible profit of one unit of product versus the storage cost; not very surprising! It has indeed the dimension of time.

Again, let  $x^*(t) \geq 0$  be a proper solution, such that  $x^*(t) = 0$  at the endpoints of the basic interval. Then, provided that there is at least one check-point, then there must be at least one true active period. Clearly, there may be several active periods. Some of these may be isolated, and others may be adjacent. Each active period must be internally optimal, and so an analysis for these is valid, whereas a little more can be proved for "isolated" active periods. Certainly, there may occur a "non-true" initial active period or a "non-true" final active period.

In order to find a way of computing a proper solution, it is clearly needed to look at the functional  $J$ , evaluated for a suitable family of restrictions, or "vertically" translated restrictions, of the guiding function  $X(t)$ . The idea of the construction is that the wanted function element  $x_0(t)$  is imbedded and "trapped" in a one-parameter family of restrictions of  $X(t)$ ; then to be identified from its optimizing property by using an appropriate derivative of  $J$ .

Some simple technical preparations are needed.

Let  $X(t)$  be increasing and linear on some interval  $t' \leq t \leq t''$ ;

decreasing and linear on some interval  $t^{**} \leq t \leq t^*$ .

Assume further that  $t'' < t^{**}$ ,  $X(t'') = X(t^{**})$  and  $X(t') = X(t^*)$ .

Assume finally that  $X(t) > X(t'')$  for  $t'' < t < t^{**}$ .

We now define a one-to-one correspondence  $t_1 \leftrightarrow t_2$  between the intervals  $[t', t'']$  and  $[t^{**}, t^*]$  by requiring that  $X(t_1) = X(t_2)$ . Clearly, this correspondence is linear. Write  $d(t) = D_1$  for  $t' < t < t''$ , and  $d(t) = D_2$  for  $t^{**} < t < t^*$ . Clearly,  $D_1 < U < D_2$ .

Now put  $x(t) = X(t) - X(t_1)$ . This defines a true active period for  $t_1 < t < t_2(t_1)$ , not necessarily optimal.

## Lemma 2

Consider the above situation. A basic formula, describing the result of an infinitesimal positive shift of the active period is needed; in other words, a one-sided derivative of  $J$  with respect to  $t_1 \in [t', t'']$ . The result is:

$$\frac{dJ}{dt_1} = (U - D_1)[h(t_2 - t_1) - (c_1 - \beta)].$$

This is called an “inner” derivative for reasons which will be explained after the proof.

Proof: To begin with, it is enough to consider the contribution to the overall functional  $J$  from the interval  $[t', t^*]$ , since  $t_1$  has no influence outside it. But it is not enough to look at  $\int_{t_1}^{t_2(t_1)} \dots$ . It is necessary to specify the situation outside  $[t_1, t_2]$ , when  $t_1$  and  $t_2$  are slightly perturbed. As mentioned,

$x(t) = X(t) - X(t_1)$  holds for  $t_1 \leq t \leq t_2$ . Further, the following convention will be used:

for  $t' < t < t_1$  and for  $t_2 < t < t^*$ , it is assumed that  $x(t) = 0$ , and

$u(t) =$  selling function  $= \min[U, d(t)]$ . Clearly, the company wants to produce and sell, also outside the activity period without creating a stock there.

We look for a derivative of  $J = \int_0^T (c_1 d(t) - \beta u(t) - hx(t)) dt$ . To find  $\frac{dJ}{dt_1}$  it is sufficient to consider  $\int_{t_1}^{t_2(t_1)} \dots$  and to remember the convention near  $t_1$  and  $t_2$ . This will be clearly seen below.

Observe that there are no restrictions on  $x(t)$  outside the interval  $[t', t^*]$ . In the following derivation nothing happens outside this interval.

Let  $t_1$  be an arbitrary point on  $(t', t'')$ . The corresponding  $x$ -trajectory ( $x(t) = X(t) - X(t_1)$ ) reaches zero at  $t_2(t_1)$ , but not earlier. The demand near  $t_1$  is  $D_1 < U$ , and the demand near  $t_2$  is  $D_2 > U$ . For comparison, let another trajectory start at  $t_1 + \delta$ , where  $\delta > 0$  will soon be sent to zero. The perturbed trajectory reaches zero at time  $t_2 - \delta'$ . Because of volume conservation these quantities are linked by the simple relation  $\delta(U - D_1) = \delta'(D_2 - U) =$  volume perturbation. The “merits” of the trajectories must be compared:

Change of production cost  $= \delta \cdot (U - D_1) \cdot \beta > 0$ ; in favour of perturbed curve.

Change of selling income:  $c_1 \cdot \delta' \cdot (D_2 - U) > 0$ ; in favour of unperturbed curve.

Change of storage cost:  $h \cdot \delta \cdot (U - D_1)(t_2 - t_1) + o(\delta) > 0$  ; in favour of perturbed curve.

It is understood here that during the interval  $[t_1, t_1 + \delta]$  it holds  $u(t) = D_1$  for the perturbed curve, and  $u(t) = U$  for the unperturbed curve. The selling is  $D_1$  for both curves.

During the interval  $(t_2 - \delta', t_2)$  the selling is  $D_2$  for the unperturbed curve, and  $U$  for the perturbed curve. There is no change of production cost here.

The change of storage cost should be obvious, so the quantities are clear.

Adding things together, we get

$$\Delta J = h \cdot \delta \cdot (U - D_1)(t_2 - t_1) + o(\delta) + \delta \cdot (U - D_1) \cdot \beta - c_1 \cdot \delta' \cdot (D_2 - U).$$

Dividing by  $\delta$  and sending it to zero gives

$$\frac{dJ}{dt_1^i} = h \cdot (U - D_1)(t_2 - t_1) + \beta(U - D_1) - c_1 \cdot \frac{\delta'}{\delta} \cdot (D_2 - U),$$

where the  $i$  sign (inner) indicates that this is a one-sided derivative. From above we have the relation  $\frac{\delta'}{\delta} = \frac{U - D_1}{D_2 - U}$ . Inserting this into the above formula, we find, as was claimed

$$\frac{dJ}{dt_1^i} = (U - D_1)[h(t_2 - t_1) - (c_1 - \beta)].$$

Observe that  $\frac{dJ}{dt_1^i} > 0$  if  $(t_2 - t_1) >$  the intrinsic length  $l_0$ ; that was expected.

This one-sided derivative is called inner because it is always defined, independently of whether the activity period in question is isolated or not.

No “free space” is needed. But the corresponding one-sided derivative for decreasing  $t_1$  clearly needs some free space at both ends. It is called the outer derivative.

### Corollary 1

The corresponding “outer” derivative can be derived in a completely analogous way, if  $t_1 \in (t', t'']$ . The result is:

$$\frac{dJ}{dt_1^o} = (U - D_1)[h(t_2 - t_1) - (c_1 - \beta)].$$

### Corollary 2

Assume that  $t' < t_1 < t''$ . Then the derivatives are clearly equal and continuous in a neighbourhood of  $t_1$ .

Let  $t_1$  vary over  $(t', t'')$ . Then  $t_2$  is a linear function of  $t_1$ , and

$\frac{dJ}{dt_1} = (U - D_1)[h(t_2 - t_1) - (c_1 - \beta)]$ . Thus,  $J(t_1) \in C^1(t', t'')$ , and a second derivative is easily computed. We have already  $\frac{dt_2}{dt_1} = \frac{D_1 - U}{D_2 - U} < 0$ , and a simple calculation gives

$$\frac{d^2J}{dt_1^2} = -(U - D_1) \cdot h \cdot \frac{D_2 - D_1}{D_2 - U} = \text{const.} < 0.$$

Thus,  $J$  is just a polynomial of degree 2, with negative leading coefficient, as long as  $t_1 \in (t', t'')$ .

It seems suitable to finish this section by two simple results:

### Theorem 3

A true optimal activity period cannot last longer than the intrinsic length  $l_0$ .

Proof: The inner derivative is always well defined and given by

$$\frac{dJ}{dt_1} = (U - D_1)[h(t_2 - t_1) - (c_1 - \beta)]. \text{ This is enough.}$$

### Theorem 4

Let  $(t_1, t_2)$  be any true optimal activity period. Assume that  $d(t) < U$  on some interval immediately to the left of  $t_1$  and  $d(t) > U$  on some interval immediately to the right of  $t_2$ .

Then  $t_2 - t_1 = l_0 =$  the intrinsic length.

Proof: Look at the outer derivative! Because of the assumptions concerning  $d(t)$ , the endpoints are free to move (no adjacent activity periods!) so that the interval can be expanded. If  $t_2 - t_1$  were less than  $l_0$ , then this would imply a negative outer derivative. Then a slight expansion would give an improvement, contradicting optimality. Together with Theorem 3, this completes the proof.

Remark: This argument is correct also if  $d(t)$  should happen to be discontinuous at  $t = t_1$ . Just approximate  $t_1$  from the left and perform a simple limiting procedure.



## 5. On problems with a seasonal demand. Solution of Case 1

For problems where the demand function has a simple seasonal structure it is possible to use our preceding results to design a rather simple computational method to find an optimal solution. The idea is to exploit some helpful monotonicity properties of the guiding function. It is assumed that the demand for the product is low in the first part of the basic interval  $[0, T]$ , then higher in the middle part and then low again.

The solution will look differently, depending on the parameter values and the choice of end conditions. The beginning condition is  $x(0) = 0$ , unless otherwise is said. It seems advisable to look first at the simpler cases.

To be concrete, let the basic interval  $[0, T]$  be divided into open sub-intervals  $I_k$ ,  $k = 1, 2, \dots, M + N + P$ , such that  $d(t)$  is constant and non-negative on each sub-interval. It is further assumed that  $d(t) < U$  on  $I_k$  for  $k = 1, \dots, M$ ;  $d(t) > U$  on the following  $N$  intervals and finally  $d(t) < U$  on the last  $P$  intervals. Write  $I_k = (t_k^-, t_k)$  for all  $k$ . Clearly,  $X(t)$  is strictly increasing on  $[0, t_M]$ , strictly decreasing on  $[t_M, t_{M+N}]$ , and strictly increasing on  $[t_{M+N}, t_{M+N+P}]$ . Consider  $X(t)$  on the interval  $[0, t_{M+N}]$ . It has a strict maximum at  $t = t_M$ . Further,  $t_M$  is a check-point and  $x_0(t_M) > 0$  for any optimal  $x_0(\cdot)$ , as proved above. Obviously, there exists an activity period, starting at some  $t_0$  on the interval  $[0, t_M]$ . Let it be represented by  $x_0(\cdot)$ . Write  $T = t_{M+N+P}$ .

1. Assume to begin with that  $X(t_{M+N}) \leq 0 = X(0)$ .

This is the simplest case and it makes sense to consider it first.

This means that the stock goes down to zero and the activity period ends before the transition to low season, also in the case of the maximal solution.

The optimization over  $[t_{M+N}, T]$  can be left aside, until further.

In this case the activity period must be a "true" one, and so Theorem 1 is applicable, according to which the active trajectory  $x_0(\cdot)$  must be a restriction of the guiding function (+ a constant), as long as  $x_0(t) > 0$ . This also means that  $x_0(t') = 0$  must hold for some  $t' \leq t_{M+N}$ . Thus,  $x_0(t) > 0$  on the interval  $t_0 < t < t'$  and  $x_0(t) = 0$  for  $t' \leq t \leq t_{M+N}$ . The problem is that  $t_0$  and  $t'$  are both unknown.

Put  $t^* = \min \{t; t > 0, X(t) \leq 0\}$ . Thus,  $t_{M+N} \geq t^* > t_M$  and

$X(t^*) = 0$ . Because of the nice continuity and monotonicity properties of  $X(t)$  on the interval  $[0, t^*]$ , the relation  $X(t_1) = X(t_2)$  defines a piece-wise linear strictly monotone mapping  $t_1 \rightarrow t_2$ , for  $t_1 \in [0, t_M]$ , suitably stored by the computer. The basic interval for optimization and identification of  $t_0$  is now

$[0, t_M]$ . The idea of the construction is now that the wanted function element  $x_0(\cdot)$ , on  $[t_0, t']$  is imbedded and "trapped" in a one-parameter family of restrictions of  $X(t)$ ; then to be identified from its optimizing property. It will be carried out in some detail below, after a short digression on case 2. Note that the one-to-one correspondence  $t_1 \leftrightarrow t_2$  here is similar to the one in §4, but now expanded considerably.

**Definition:** a break point is a value for  $t_1$ , where  $d(t)$  has a discontinuity, or such that  $d(t)$  has a discontinuity at  $t_2(t_1)$ .

There will be a finite number of break points on  $[0, t_M]$ , easily identified. The number  $t^*$  and all break points are easily recorded by the computer. The correspondence  $t_1 \leftrightarrow t_2$  is linear between break points.

2. The case  $X(t_{M+N}) > 0$  will be considered in §7 as a part of the whole solution of our problem. It will, among other things, produce another version of the preceding case 1.

Return to case 1 for a closer description. Clearly,  $J(t_1) = \int_0^{t^*} \dots dt$  is continuous for  $t_1 \in [0, t_M]$ , and, according to §4, reduces to a polynomial of order two on each of the intervals, written  $A_j, j = 1, 2, \dots, J$ , between the break points. For each interval  $A_j$  there exists a "twin interval"  $B_j$  to the right of  $t_M$ , defined by the one-to-one correspondence  $t_1 \leftrightarrow t_2$ . The demand on  $A_j$  is denoted  $D_{1,j}$  and the demand on  $B_j$  is written  $D_{2,j}$ .

Further,  $t_2 - t_1$  is a

continuous, strictly decreasing function of  $t_1$ , and it is readily seen from the formula

$$\frac{dJ}{dt_1} = (U - D_{1,j})[h(t_2 - t_1) - (c_1 - \beta)]$$

that this derivative can change sign only once; from + to -. The factor  $(U - D_{1,j})$  can be discontinuous, but does not change sign (positive) at a break point. Therefore,  $J(t_1)$  has a unique maximum on  $[0, t_M]$ , for identifying the point  $t_0$ .

Observe first that  $t_1 = t_M$  is an impossible maximum point, since the above derivative would then be negative.

Then observe that for  $t_1 = 0$ , we get  $\frac{dJ}{dt_1} = (U - D_{1,1})[ht^* - (c_1 - \beta)]$ . Thus, if  $t^* > l_0$ , then  $t_1 = 0$  cannot be optimal, because of  $\frac{dJ}{dt_1} > 0$ .

If  $t^* \leq l_0$  (cheap storing!), then  $t_1 = 0$  is optimal, because of  $\frac{dJ}{dt_1} \leq 0$  on the whole interval.

So assume now  $t^* > l_0$ ; the more interesting case.

Since all constants involved are known, we can easily sketch a suitable numerical procedure for finding  $t_0$ , which also gives the wanted activity period. It seems that the simplest way would be to find  $t_1$  such that  $t_2(t_1) - t_1 = l_0$ ; now a routine matter.

The procedure will be as follows:

1. Identify and list all break points on  $[0, t_M]$ .
2. For each break point, find  $t_2 - t_1$
3. On some sub-interval  $I^*$  between break points,  $(t_2 - t_1) - l_0$  will change sign, unless we already have found a zero for this quantity.
4. For this particular sub-interval  $I^*$ , find the value of  $\frac{dt_2}{dt_1} - 1 = \frac{D_{1,j} - D_{2,j}}{D_{2,j} - U} = \text{const.} < 0$ . Observe that  $D_{1,j} < U$  and  $D_{2,j} > U$ .
5. Find  $t_1$  on  $I^*$ , such that  $t_2 - t_1 = l_0$ . This value of  $t_1$  defines the optimal activity period. So this value of  $t_1$  is the wanted starting point  $t_0$ .

## 6. Some general observations for the case of a free end-point

For the moment, we make no assumptions concerning seasonal demand. Consider an optimal trajectory  $x_0(\cdot)$ , starting from zero at time  $t = t_1$ . Let  $x_0(\cdot)$  be the last active period. No restriction yet concerning the end-point. Theorem 1' is applicable, according to which the trajectory must be a restriction of the guiding function. But here, in contrast to §5, a final activity period can occur close to  $t = T$ , depending on the final "payment"  $c_0$ .

### Lemma 3

("Closing" lemma). Let  $c_0 \leq \beta$ . Then  $x_0(T) = 0$ .

Proof: Assume that  $x_0(T) > 0$ . We will look at a derivative of  $J$  with respect to the starting point of the trajectory  $x_0(\cdot)$ , much as in §4. As in §4, the trajectory is imbedded in a family of restrictions (+ constant) of  $X(t)$ . Let  $x_0(t)$  start from zero at  $t = t_1 < T$ . The demand immediately to the right of  $t_1$  is denoted by  $D_1 < U$ . Find an inner derivative, as in §4.

For comparison, let another trajectory start at  $t_1 + \delta$ , where  $\delta > 0$  will soon be sent to zero. In other words, the higher production starts  $\delta$  time units later. Clearly,

$x_0(t_1 + \delta) - x_0'(t_1 + \delta) = \delta(U - D_1)$ , where  $x_0'(\cdot)$  is the perturbed trajectory. Then, by volume conservation, we also have

$$x_0(T) - x_0'(T) = \delta(U - D_1).$$

The merits of the trajectories are now compared, using the convention in §4:

Change of production cost:  $\beta \cdot \delta \cdot (U - D_1) > 0$  ; in favour of perturbed curve.

Change of storage cost:  $h \cdot \delta \cdot (U - D_1)(T - t_1) + o(\delta) > 0$  ; in favour of perturbed curve.

Change of final payment:  $c_0 \cdot \delta \cdot (U - D_1) \geq 0$  ; in favour of unperturbed curve.

No change of selling near  $t_1$ !

Adding things together, we get

$\Delta J = \delta \cdot (U - D_1)[h(T - t_1) + \beta - c_0] + o(\delta)$  ; in favour of perturbed curve.

Thus,  $\frac{dJ}{dt_1} = (U - D_1)[h(T - t_1) + \beta - c_0]$ , where "i" again stands for "inner".

Consequently, if  $x_0(\cdot)$  is an optimal trajectory, then

$h(T - t_1) + \beta - c_0 \leq 0$ , which immediately gives  $T - t_1 \leq \frac{1}{h}(c_0 - \beta) \leq 0$ . The contradiction proves the lemma.

Observe that this is correct also if  $d(t)$  should happen to be discontinuous at  $t_1$ .

Assume now that  $c_0 > \beta$ . Various situations can still occur, but the following is clear:

### Lemma 4

Let  $c_0 > \beta$  and let  $d(t) < U$  near  $t=T$ . Then  $x_0(T) > 0$  at optimum.

Proof: This is very similar, though not identical, to the proof of Lemma 1. We leave the details to the reader.

Consider again an optimal trajectory  $x_0(\cdot)$ , starting at  $t = t_1$ , and such that  $x_0(T) > 0$ . The argument in the proof of Lemma 3 is still valid and gives

$T - t_1 \leq \frac{1}{h}(c_0 - \beta)$ . In order to apply an outer derivative, we must assume:

1.  $t_1 >$  the left endpoint of the basic interval
2. the demand immediately to the left of  $t_1$  is  $D'_1 < U$ . ( $D'_1 = D_1$  is not needed).

Now an outer derivative can be involved, as before. The result is

$\frac{dJ}{dt_1^0} = (U - D'_1) \cdot [h(T - t_1) + \beta - c_0]$ . The optimality clearly implies that

$T - t_1 \geq \frac{1}{h}(c_0 - \beta)$ . Combine with the previous result and summarize:

### Theorem 5

Consider an optimal trajectory  $x_0(\cdot)$ , starting from zero at  $t = t_1$ , and ending with  $x_0(T) > 0$ .

Then  $T - t_1 \leq \frac{1}{h}(c_0 - \beta)$ .

If, furthermore,  $d(t) < U$  on some interval immediately to the left of  $t_1$ , then

$T - t_1 = \frac{1}{h}(c_0 - \beta)$ . Notation: put  $l_0' = \frac{1}{h}(c_0 - \beta)$ .

This result should be compared to Theorem 4.

Observe that no assumptions were made above (in case B) concerning seasonal demand.

## 7. Solving the one-year seasonal problem with free end-point

Now make the same assumptions on seasonal demand as in the beginning of §5. It makes sense first to look at optimization over the interval  $I = [0, t_{M+N}]$ , then to look at  $II = [t_{M+N}, T]$ , and finally somehow combine the results. It will be seen below that it is only in one case, called  $2B\alpha$ , that it can happen that  $x_0(t_{M+N}) > 0$  at optimum. The solution in this case is specified. In all other cases it holds that  $x_0(t_{M+N}) = 0$  at optimum, leading to separate optimization over I and II. Also these solutions are described.

Easy fact from §5, concerning “main” case 1:

Let  $X(t_{M+N}) \leq 0$ . Then every possible trajectory  $x(t)$  satisfies  $x(t_{M+N}) = 0$ .

Numerical method for I was specified. Separate optimization over intervals I and II.

The following subcases will be used in this section:

A) the end-condition is  $x(T) = 0$ , or  $x(T)$  free end and  $c_0 \leq \beta$ .

First, let the end-condition be  $x(T) = 0$ . We also have  $x(t_{M+N}) = 0$ .

Then,  $x_0(t) \equiv 0$  and  $u_0(t) \equiv d(t)$  on II for any optimal pair  $(x_0(\cdot), u_0(\cdot))$ . This is seen as follows: for any admissible pair  $(x, u)$  we have:

$$J = \int c_1 S(t) dt - \int \beta u(t) dt - h \int x(t) dt.$$

But the boundary conditions imply  $\int u(t) dt = \int S(t) dt$ , and so

$$J = (c_1 - \beta) \int S(t) dt - h \int x(t) dt \leq (c_1 - \beta) \int d(t) dt.$$

Equality can hold here if and only if  $S(t) = d(t)$  a.e. on II, and  $x(t) \equiv 0$  on II, i.e.  $u(t) = S(t) = d(t)$  a.e. on II.

In the second case,  $x(T)$  free end and  $c_0 \leq \beta$ , invoking Lemma 3 is enough. Thus  $x_0(t) \equiv 0$  on II.

B)  $x(T)$  free, and  $c_0 > \beta$ .

Here,  $x_0(t)$  is not  $\equiv 0$  on interval II because of Lemma 4.

Optimizing over  $[t_{M+N}, T]$ , under the condition  $x_0(t_{M+N}) = 0$ .

This case was essentially solved in §6.

Cases A and B, as above.

In case B,  $x_0(T) > 0$ , and it only remains to determine the length

$T - t_1$  of the final activity period.

We still have the derivative

$$\frac{dJ}{dt_1} = (U - D_1(t_1))h[(T - t_1) - l_0'], \text{ and this expression is clearly positive for}$$

$t_1 < T - l_0'$  and negative for  $t_1 > T - l_0'$ . Clearly, for maximizing,  $t_1$  must be chosen as close as possible to  $T - l_0'$ .

Thus, if  $t_{M+N} \leq T - l_0'$ , then  $t_1 = T - l_0'$  is optimal.

If  $t_{M+N} \geq T - l_0'$ , then clearly  $t_1 = t_{M+N}$  is optimal. (Note that  $t_{M+N}$  is the smallest possible value for  $t_1$ .)

Thus, the “sub-problem” is solved, at least in principle.

Analysis of the case  $X(t_{M+N}) > 0$ . This is the “main” case number 2. Also here we will use subcases A and B.

First, some terminology and easy observations concerning trajectories.

Put  $t^* = \min \{t; t \geq 0, X(t) = X(t_{M+N})\}$ . (Note that this is not the same  $t^*$  as in §5.)

Clearly, a trajectory  $\Gamma(t_1)$  starting at some  $t_1 \in [0, t^*)$ , written  $X(t) - X(t_1)$ , will “arrive” at  $t_{M+N}$  with  $x(t_{M+N}) > 0$ . Let  $\Gamma(t_1)$  include its continuation up to  $t = T$ .

The limit trajectory,  $X(t) - X(t^*)$ , is denoted by  $\Gamma^*$  and is the limit of  $\Gamma(t_1)$ , when  $t_1$  approaches  $t^*$  from below. It is defined for  $0 \leq t \leq T$ .

(The reader should draw a simple figure.)

Thus, split the trajectories into two classes, long (L) and short(S).

Long trajectories start at  $t_1$ , where  $0 \leq t_1 \leq t^*$ .

Short trajectories start at  $t_1$ , where  $t^* \leq t_1 \leq t_M$ . These are only defined on I.

$\Gamma^*$  plays the role of a separatrix on  $(t^*, t_{M+N})$ . It can be seen as long or short.

Now look at the subcases of case 2!

A) as above. In this case, trajectories (L) are immediately excluded from optimum by Lemma 3. This leads to the only possibility  $x(t) = 0$  on interval II at optimum. (Note that the restriction of  $\Gamma^*$  to interval I is not excluded.) Thus separate optimization! To optimize over I, consider curves (S), with starting point  $t_1$ , where  $t^* \leq t_1 \leq t_M$ . This is exactly the same kind of problem as was solved in §5; only notation is different. The numerical procedure is the same.

B)  $x(T)$  free, and  $c_0 > \beta$ . Here,  $x(t)$  is not  $\equiv 0$  on interval II.

We need to optimize over (L), i.e. with respect to  $t_1$ ;  $0 \leq t_1 \leq t^*$ .

$\Gamma(t_1)$  must be compared to  $\Gamma^*$ . Consider the outer derivative of J:

$$\frac{dJ}{dt_1^0} = (U - D_1(t_1))h[(T - t_1) - l_0'], \text{ for } 0 < t_1 \leq t^*. \quad (*)$$

For  $t = t^*$  this derivative becomes  $(U - D_1(t^*))h[(T - t^*) - l_0']$ .

Again, two subcases:

$\alpha$ ) Assume that  $l_0' > T - t^*$ .

Then  $\frac{dJ}{dt_1^0}(t^*) < 0$ . Thus, there exists some  $t^{**}$ ,  $0 < t^{**} < t^*$ , such that

$$J(t^{**}) > J(t^*).$$

Furthermore:  $(T - t_1 - l_0')$  increases, if  $t_1$  decreases. Thus, in order to reach optimum, decrease  $t_1$  until we get  $T - t_1 - l_0' = 0$ , or  $t_1 = 0$ ; choose the alternative that happens first. Consequently,

Choose  $t_1 = 0$ , if  $T - l_0' \leq 0$  (extremely cheap storage!), and

choose  $t_1 = T - l_0'$ , if  $0 \leq T - l_0' < t^*$ .

This choice gives a unique optimum within the class (L) of curves, and

$J(t_1) > J(\Gamma^*)$ . Now, does this procedure give the absolute optimum? It is also needed to verify that  $\Gamma^*$  is at least as good as all separated solutions, over interval I and interval II combined. But this is correct, for the following reasons:

$l_0' > T - t^*$  also implies  $l_0' > T - t_{M+N}$ , i.e.  $\Gamma^*$  is optimal, considered on II.

It also follows that  $l_0 > t_{M+N} - t^*$ , which means that  $\Gamma^*$  is optimal on I. Thus, the absolute optimum has been identified, and it is unique.

$\beta$ ) Assume that  $l_0' \leq T - t^*$ . It follows from (\*) that  $\frac{dJ}{dt_1^0} > 0$ , if  $t_1 < t^*$ .

Hence  $J(t_1) < J(t^*)$ , if  $t_1 < t^*$ .

Clearly, all trajectories (L) can be excluded from the optimization, except possibly  $\Gamma^*$ . Therefore, necessarily  $x_0(t_{M+N}) = 0$ , and the optimization must be done separately. This is similar to 2A above. Thus, consider now only the case  $t^* \leq t_1 \leq t_M$ .

Finally, observe here that  $\Gamma^*$  actually delivers the absolute optimum if (but not only if)  $l_0' = T - t^*$ . This is so, because it immediately follows that

$l_0' \geq T - t_{M+N}$  and  $l_0 \geq t_{M+N} - t^*$ , since  $l_0 \geq l_0'$ .

A final observation: in case 2A the long curves are discarded because of the boundary conditions at  $t=T$  and the maximum principle; in case 2B $\beta$  most long curves are discarded because of being uneconomical!

## 8. On the single demand period problem. A solution formula

We will consider here the same kind of problem as before, but with two changes: the demand is assumed constant, denoted  $D$ , over the interval  $T_1 \leq t \leq T_2$  considered, and non-negative boundary values  $X_1, X_2$  are prescribed. As before, it is assumed that  $U \neq D$ , and  $D \geq 0$ . This is a simpler problem than the preceding, but still of some interest and importance.

The admissible domain  $\Omega$  is simply the set in the  $(t,x)$ -plane, where  $(t,x(t))$  can occur for an admissible trajectory. The forward set  $S_1$  consists of all points  $(t,x)$  which can be reached, starting from  $(T_1, X_1)$ , ignoring the second boundary condition. It is often called the reachable set. There is also a backward set  $S_2$ , analogously defined. The admissible domain is the intersection of  $S_1$  and  $S_2$ . It is a compact set, not always convex, as will be seen below.

Clearly,  $\Omega$  can be the empty set, the meaningless case.

It can also consist of only a straight line segment, the trivial case.

1) To get more insight, it makes sense to start with the easier case,  $U > D \geq 0$ .

Define the upper function  $F_1(t) = X_1 + (U - D)(t - T_1)$ , and

the lower function  $F_2(t) = \max[X_1 - D(t - T_1), 0]$ . The geometrical meaning of these functions is obvious.

Now, an admissible trajectory  $x(t)$  exists if and only if  $F_2(T_2) \leq X_2 \leq F_1(T_2)$ .

Assume that this condition is satisfied. In this fortunate case, an optimal element can be specified. It is simply

$$m(t) = \max[X_1 - D(t - T_1), 0, X_2 - (U - D)(T_2 - t)]. \quad (*)$$

This is clearly the point-wise smallest of all admissible elements. The reader should verify that it has the prescribed boundary values. It also gives the smallest possible storage cost. That this is the only optimal element will now be verified.

Assume first that  $m(t) > 0$  for  $T_1 \leq t \leq T_2$ , so the side condition  $x(t) \geq 0$  is never active.

For any admissible pair  $(u(t), x(t))$  it holds that  $\dot{x} = u(t) - S(t)$ , a.e., where  $S(t)$  is the rate of selling. Therefore,

$$\int u(t) dt - \int S(t) dt = X_2 - X_1 = \Delta X. \text{ Further, this immediately gives}$$

$$J = c_1 \int S dt - \beta \left( \int S dt + \Delta X \right) - h \int x(t) dt.$$

In the present case,  $S = D$  a. e. for any  $(u(t), x(t))$ , giving

$$J = c_1 D (T_2 - T_1) - \beta [D (T_2 - T_1) + \Delta X] - h \int x(t) dt,$$

$$\text{i.e. } J = (c_1 - \beta) D (T_2 - T_1) - \beta \Delta X - h \int x(t) dt \quad (**)$$

The first two terms here are independent of  $(u,x)$  and the last term is minimized by  $m(t)$ . Thus,  $m(t)$  is clearly the one and only optimal element.

Next, let  $m(t) = 0$  on some interval. In the present case,  $U > D$ , the relation



$S = D$  still holds everywhere, including this particular interval, because of the conventions in §1. So, (\*\*) still holds, and the optimality of  $m(t)$  again follows.

Uniqueness of the optimal element also follows.

2) It remains to consider the case  $D > U > 0$ .

Define the upper function  $G_1(t) = \max[X_1 - (D - U)(t - T_1), 0]$ , and the lower function  $G_2(t) = \max[X_1 - D(t - T_1), 0]$  (identical with  $F_2(t)$ ). Also in this case, an admissible trajectory exists if and only if  $G_2(T_2) \leq X_2 \leq G_1(T_2)$ . Assume that this condition is satisfied.

The reachable set is confined by the trajectories of  $G_1$  and  $G_2$ , and parts of the  $t$ -axis and/or the line  $t = T_2$ . It need not be convex.

Assume first that  $X_2 > 0$ . Since each trajectory is non-increasing it follows that  $x(t) \geq X_2 > 0$  for any admissible trajectory. This means that the relation (\*\*) is valid for any admissible pair, which in turn means that the point-wise smallest trajectory is optimal, as before. And this trajectory is in this case

$m_1(t) = \max[X_1 - D(t - T_1), X_2 + (D - U)(T_2 - t)]$ . The reader should verify that this is correct! It is also seen that this solution is unique.

It remains to consider the case  $X_2 = 0$ .

Observe first that if  $X_1 = 0$ , then  $x(t) \equiv 0$  for any admissible element. So let  $X_1 > 0$ .

Next, it makes sense to see what the maximum principle may give for an optimal pair  $(u_0(\cdot), x_0(\cdot))$ , starting at  $(T_1, X_1)$ , where  $X_1 > 0$ . It must necessarily reach zero at some  $t' \leq T_2$ , or "survive" positive up to  $t = T_2$ . We can apply the maximum principle to this optimal pair as in §3, pages 10-11 but now over  $[T_1, t']$  or  $[T_1, T_2]$ . The situation is not exactly as on page 10. Here, the trajectory considered does not start from zero, and therefore the conclusion must be modified.

We have the control system (state equation):

$\dot{x}_1 = u - D$ ,  $\dot{x}_2 = c_1 D - \beta u - h x_1$ , and the adjoint system

$\dot{\eta}_1 = h \eta_2$ ,  $\dot{\eta}_2 = 0$ .

Now apply the MP on  $[T_1, T']$ , for some  $T' < t'$  (or  $T_2$ ), like on pp.10-11. The relevant part of the Hamiltonian is just  $u(\eta_1 - \beta \eta_2)$ , to be maximized a.e. along the optimal trajectory. The possible cases for  $w_2 (= \eta_2(T'))$  are as before  $w_2 > 0$  or  $w_2 = 0$ .

1) Let  $w_2 > 0$ . Then  $\eta_1 = h w_2 > 0$  and  $\eta_2 = 0$ . Therefore, it follows that  $u = 0$  as long as we have  $\eta_1 < \beta \eta_2 = \text{const.}$ , and  $u = U$ , if  $\eta_1 > \beta \eta_2$ . Thus, only a switch from off to on is possible, or no switch.

2) Let  $w_2 = 0$ , i.e.  $\eta_2(T') = 0$ . But  $\dot{\eta}_2 = 0$ , so  $\eta_2(t) \equiv 0$ . Clearly, we must have

$\eta_1 = \text{const.} \neq 0$ . Then MP implies that  $u(t) \equiv 0$  or  $u(t) \equiv U$  on  $[T_1, T']$ . But  $T'$  was arbitrary except that  $T_1 < T' < t'$  (or  $T_2$ ).

Therefore, the only alternatives for  $u_0$  are:

A)  $u_0 \equiv 0$  a.e. on  $(T_1, t'$  (or  $T_2))$ .

B)  $u_0 \equiv U$  a.e. on  $(T_1, t'$  (or  $T_2))$ .

C)  $u_0$  makes a switch from 0 to  $U$  at some time  $t_1$  between  $T_1$  and  $t'$  (or  $T_2$ ).

Motivated by this result, we consider two straight lines, emanating from  $(T_1, X_1)$ ; one with slope  $-D$  and the other with slope  $U-D$ . Each of these lines is continued up to the line  $t = T_2$ , or until it hits the  $t$ -axis. These lines are now seen as trajectories, denoted as  $\Gamma_0$  and  $\Gamma_U$ , respectively, for obvious reasons.

A third trajectory  $\Gamma_1$  will also be considered. It starts like  $\Gamma_0$  and then has a switch to  $u(t) = U$  at some time  $t_1$ , until further undetermined.

Before looking at some specific geometric situations, it makes sense to look at a “pilot case” and solve that. This solution will later be invoked in several similar cases.

It is now assumed that  $\Gamma_0$ ,  $\Gamma_1$  and  $\Gamma_U$  all hit the  $t$ -axis at times  $t_0$ ,  $t_2$  and  $t_U$ . For geometric reasons it follows that  $T_1 \leq t_1 \leq t_0 \leq t_2 \leq t_U \leq T_2$ . (The reader should draw a figure!) Now,  $t_1$  and  $t_2$  are unknown, and  $t_2$  is a strictly decreasing function of  $t_1$ .

Consider a situation where the trajectory  $\Gamma_1$  is disturbed by applying the switch  $\delta$  time units later. The perturbed trajectory will hit the  $t$ -axis  $\delta'$  time units earlier. The “merits” of the two trajectories must be compared, in analogy with similar derivations in §4 and §6. (The reader should draw another simple figure!)

Change of storage cost:  $U\delta h(t_2 - t_1) + o(\delta)$ , in favour of perturbed curve.

Change of production cost:  $\beta U \delta$ , in favour of perturbed curve.

Change of selling:  $\delta' c_1 (D - U)$ , in favour of unperturbed curve.

Thus,  $\Delta J = \delta' c_1 (D - U) - \beta U \delta - \delta U h(t_2 - t_1) + o(\delta)$ , in favour of unperturbed curve.

Volume conservation obviously implies that  $\delta' (D - U) = \delta U$ , and so

$\Delta J = \delta U [c_1 - \beta - h(t_2 - t_1)] + o(\delta)$ , in favour of unperturbed curve.

This leads to the derivative

$$\frac{dJ}{dt_1} = U[h(t_2 - t_1) - c_1 + \beta], \text{ more conveniently written as}$$

$$\frac{dJ}{dt_1} = Uh[t_2 - t_1 - l_0].$$

Like in similar earlier situations, it is clear that  $\frac{d}{dt_1}(t_2 - t_1) < \text{const.} < 0$ . It is therefore clear that  $t_1$  should be chosen so that  $t_2 - t_1$  comes as close as possible to  $l_0$  (the intrinsic length), given possible geometric restrictions. Thus,

in particular,  $t_2 - t_1 > 0$  at optimum. The solution in the pilot case is now determined, but we have no explicit formula for the solution.

Now, the trajectories  $\Gamma_0$  and  $\Gamma_U$  can be end in various positions relative to the point  $(T_2, 0)$ . One possibility is that  $\Gamma_0$  ends in  $(T_2, 0)$ . Then  $\Gamma_0$  is the only admissible trajectory, and thus also optimal. In the other cases the solution is obtained by appropriately adapting the result from the pilot case. We leave the details of that procedure. It is still the case  $X_2 = 0$  in focus. We will now perform a phase-plane analysis of the admissible domain  $\Omega$  for unspecified starting point and determine all optimal trajectories ending in  $(T_2, 0)$ .

To begin with, it is clear that each point  $(t, x)$  such that  $x > D(T_2 - t)$  is not admissible. (The stock is simply too big and cannot be sold out in “due” time!)

Moreover, the starting point  $(T_1, D(T_2 - T_1))$  defines a unique admissible trajectory, which is optimal by definition.

The further determination of all optimal trajectories starting on the "axis"  $t = T_1$  and ending at  $(T_2, 0)$  depends on the intrinsic length  $l_0$ .

Assume first that  $l_0 \geq T_2 - T_1$ ; the simplest case. The straight line segment

$L: \{(t, x): T_1 \leq t \leq T_2, x = (D - U)(T_2 - t)\}$  divides  $\Omega$  into two parts, the upper and the lower. Consider an optimal trajectory  $\Gamma$  starting in the upper part! Two values for the slope are possible; first  $-D$  and then  $U-D$  after a switch.

It is clear from the geometry that there cannot be any switch before  $\Gamma$  hits  $L$ . Then it follows from the pilot case that  $\Gamma$  must follow  $L$  to the end-point  $(T_2, 0)$ .

Thus, in the upper part of  $\Omega$  all trajectories are straight lines with slope  $-D$ .

For the lower part a similar argument invoking the pilot case and the fact that  $l_0 \geq T_2 - T_1$  shows that the slope is  $U-D$  throughout. All trajectories here hit the  $x$ -axis and then follow it. All optimal trajectories are now determined, in this case. Observe that the whole line  $L$  is a switching locus. The  $x$ -axis also serves as a switching locus, but of a different nature. The reader should draw a figure!

The case  $l_0 < T_2 - T_1$  is slightly different. Here, the part of  $L$  (the same  $L$  as above), corresponding to  $t_2 - l_0 \leq t \leq t_2$ , is still a switching locus, but now continued to the left by a horizontal segment. Thus, there is a "broken" switching locus. The  $x$ -axis is also here a switching locus. Details are left to the reader. The reader should again draw a figure!

Finally, the case of a fixed starting point  $(T_1, X_1)$  where  $X_1 > 0$ , and the end-point free also leads to an interesting family of optimal trajectories. The phase-plane analysis will be somewhat similar to the preceding case, but certainly not identical. Two "empty zones" can occur, namely if  $l_0$  is less than a certain geometric quantity, defined from  $X_1, (T_2 - T_1), U$  and  $D$ . (It is still assumed that  $D > U$ .)

A closer study is left to the interested reader.

## 9. On the existence of an optimal element

Return to the general optimization problem as it was stated in §1, on an interval  $[0, T]$ . An initial value is prescribed, the final value may be prescribed or free. No seasonal conditions are imposed. The following result is basic:

### Theorem 6

If there exists an admissible element, then there also exists an optimal element.

Proof: This is based on the classical technique of selecting convergent subsequences, to produce a "good" limit element. Start by an admissible optimizing sequence  $\{(u_k(\cdot), x_k(\cdot))\}$ ,  $k = 1, 2, \dots$ . There is also an associated sequence  $\{(S_k(\cdot))\}$ . The functions  $x_k(\cdot)$  are clearly uniformly bounded and equicontinuous on  $[0, T]$ . Thus there is a uniformly convergent subsequence on  $[0, T]$ , converging to some absolutely continuous (in fact Lipschitzian) function  $x_0(\cdot)$ . There is no need to change notation for the sequence in question. Starting from this subsequence we make another selection of a subsequence so that the new sequence of control functions converges weakly in  $L^2[0, T]$  to a limit  $u_0(\cdot)$ . A final selection gives a subsequence  $\{S_k(\cdot)\}$ , also weakly convergent in  $L^2[0, T]$  to a limit  $S_0(\cdot)$ . All this produces a convergent sequence of consistent triples  $\{(u_k, x_k, S_k)\}$  on  $[0, T]$ . Now take an arbitrary  $t \in [0, T]$ .

Then

$$x_k(t) = x_k(0) + \int_0^t (u_k(s) - S_k(s)) ds. \text{ Passage to the limit gives}$$

$x_0(t) = x_0(0) + \int_0^t (u_0(s) - S_0(s)) ds$ . Therefore  $\dot{x}_0(t) = u_0(t) - S_0(t)$  for almost all  $t \in [0, T]$ , so the state equation is valid in the limit. Further, the passage to the limit in the linear functional  $J$  is no problem because of the weak convergence. Thus, the triple  $(u_0, x_0, S_0)$  delivers the optimal value of  $J$ , provided that all side-conditions in §1 are satisfied.

Now,  $0 \leq u_k(t) \leq U$  for all  $k$  and  $t$ . The weak convergence implies the same bounds for  $u_0(t)$ . Further,  $0 \leq S_k(t) \leq d(t)$  for all  $k$  and  $t$ , and this holds on each demand period. The weak convergence gives the same bounds for  $S_0(t)$ .

Next, consider  $E = \{t: x_0(t) = 0\}$ . Almost every point of  $E$  is an accumulation point for  $E$ , so that  $\dot{x}_0(t) = 0$  a.e. on  $E$ , which implies that  $u_0(t) = S_0(t)$  a.e. on  $E$ .

Finally, consider  $F = \{t: x_0(t) > 0\}$ . Take an arbitrary point  $t' \in F$ , not a discontinuity for  $d(t)$ . Take an open interval  $I'$  so that  $t' \in I' \subset F$ , and such that

$d(t) = \text{constant} = d'$  on  $I'$ . Let  $x_0(t) > \delta > 0$  on  $I'$ . According to §1 each  $S_k(t) = d'$  a.e. on  $I'$  for  $k$  big enough, because of the uniform convergence to  $x_0(t)$ . Because of the weak convergence of  $S_k(t)$  to  $S_0(t)$  it follows that  $S_0(t) = d'$  a.e. on  $I'$ . The discontinuities for  $d(t)$  can be ignored here.

Thus, the triple  $(u_0, x_0, S_0)$  satisfies the side conditions of §1, so the proof is complete.

Remark: This existence theorem can, as far as we can see, be proved using results from [C] and some extra arguments.

## 10. Two examples involving seasonal demand

### 1. A very simple example with seasonal demand

Using the notation of §5, put  $M = N = P = 1$ . This means low season, demand  $D_1 < U$ ; then high season, demand  $D_0 > U$ ; finally low season, demand  $D_2 < U$ . The length of these periods need not be equal. Keep the notation  $t_M, t_{M+N}, T$ . Clearly, the guiding function  $X(t)$  is linear and increasing on

$[0, t_M]$ , linear and decreasing on  $[t_M, t_{M+N}]$ , finally linear and increasing on  $[t_{M+N}, T]$ . As before,  $t_M$  is a check-point and  $x_0(t_M) > 0$  if  $x_0(\cdot)$  is optimal. The sign of  $X(t_{M+N})$  depends on the parameters, but let us assume that  $X(t_{M+N}) \leq 0$ , for simplicity of presentation.

The end condition is  $x(T) = 0$ . Thus, only one activity period for  $x_0(\cdot)$ .

As before (§5), we have  $t^* = \min \{t; t > 0, X(t) \leq 0\}$ . Thus,  $t_{M+N} \geq t^* > t_M$  and  $X(t^*) = 0$ . Further, the relation  $X(t_1) = X(t_2)$  defines a linear strictly monotone mapping  $t_1 \rightarrow t_2$ , for  $t_1 \in [0, t_M]$ , and where  $t_2 \in [t_M, t^*]$ .

Assume first that  $l_0 \geq t^*$ , i.e. it is very cheap to store. It then follows easily from Lemma 2 that the optimal solution  $x_0(t)$  is given by  $X(t)$  itself for

$0 \leq t \leq t^*$ , and by zero for  $t \geq t^*$ . There is a similar argument in §5.

Assume then that  $t^* > l_0 > 0$ , i.e. not very cheap to store. Then there exists a unique couple  $t_1, t_2$  such that  $X(t_1) = X(t_2)$ ,  $0 < t_1 < t_M$ , and  $t_2 - t_1 = l_0$ . Now, as known from §5, the couple  $t_1, t_2$  defines the unique optimal solution, i.e.  $x_0(t) > 0$  for  $t_1 < t < t_2$ , and otherwise  $x_0(t) = 0$ . Observe that if  $h$  is decreased, then  $l_0$  increases and the activity interval increases, as expected.

The important moments are:  $0 < t_1 < t_M < t_2 < t^* \leq t_{M+N} < T$ . The optimal control  $u_0(t)$  and  $x_0(t)$  can be specified as follows:

$u_0(t) = D_1$  for  $0 < t < t_1$ , passivity period;  $x_0(t) = 0$ ;

$u_0(t) = U$  for  $t_1 < t < t_2$ , activity period;  $x_0(t) > 0$ ;

$u_0(t) = U$  for  $t_2 < t < t_{M+N}$ , passivity period;  $x_0(t) = 0$ ;

$u_0(t) = D_2$  for  $t_{M+N} < t < T$ , passivity period;  $x_0(t) = 0$ .

The reader should draw a figure illustrating the situation! It is also instructive to look at the variation of the selling over the whole interval  $[0, T]$ .

Observe that the onset  $t_1$  of high production and of  $x_0(t) > 0$  can not occur earlier than  $t_M - l_0$ . It can come close to  $t_M - l_0$  if  $D_0$  is very big. All this is natural in view of the interpretation of  $l_0$ . See also Appendix 1!

### 2. Another example involving seasonal demand

Consider a producing and selling company over a period of 1 year. It is assumed that the demand function  $d(t)$  is highly seasonal. During months 1-8 the demand is step-wise increasing, and during months 9-12 it is step-wise decreasing. The demand is strictly bigger than  $U$  during months 5-10, i.e. high season; and strictly less than  $U$  the other months. Let  $d_k$  denote the demand in interval number  $k$ . The process starts with vanishing stock, and must end likewise.

The values are as follows:  $U = 4,5$ . On intervals  $I_k$ ,  $k = 1, 2, \dots, 8$  we have  $d_k = k$ . On intervals  $I_k$ ,  $k = 9, \dots, 12$  we have  $d_k = 16 - k$ . (The reader should draw a simple figure and fill in the step function  $d(t)$ , plus the maximal production level  $U$  to see the situation better.)

Further notation:  $I_k = (k - 1, k)$ . In the notation of §5 we therefore have

$M = 4$ ,  $N = 7$  and  $P = 1$ . The value of  $U$ , as well as values assigned to  $d(t)$  in the intervals are of course artificial, and chosen in order to get simple arguments.

By construction, the demand function has the following simple anti-symmetry property:  $U - d_k = d_{9-k} - U$  for  $k = 1, 2, 3, 4$ . Consequently,

$$\sum_{k=1}^4 (U - d_k) = \sum_{k=1}^4 (d_{9-k} - U) = \sum_{j=5}^8 (d_j - U) = - \sum_{j=5}^8 (U - d_j).$$

Thus,  $\sum_{i=1}^8 (U - d_k) = 0$ . This implies that  $X(8) = 0$ . In the notation of §5, this means that  $t^* = 8$ .

Let  $x^*(t) \geq 0$  be a proper solution to the problem. Clearly,  $t = 4$  is a check-point and it follows from Lemma 1 that  $x^*(4) > 0$ . Thus, there exists an activity period containing  $t = 4$ . It must start in one of the intervals 1,2,3,4.

Theoretically, it may end in any of the intervals number 5- 8, and which one is determined by the parameters of the problem. Note that the important other scalars of the problem  $c_1, h, \beta$  have not yet been involved. Assume until further that the end-point condition is  $x^*(T) = 0$ . Clearly, then there cannot be any more activity period. Let the activity period end at a point  $t_0 \leq 8$ . Then,  $x^*(t) = 0$  for  $t \geq t_0$ . The starting point of the activity period must be determined from the optimization conditions, derived in §5. To do so, a value must be assigned to the quantity  $l_0 = \frac{1}{h}(c_1 - \beta)$ , i.e. the intrinsic length of the problem.

Assume first that  $l_0 \geq 8$ . In this case it follows from the formula for interior derivative in §4 that the optimal  $x^*(t)$  is given by  $X(t)$  for  $0 \leq t \leq 8$ , and  $x^*(t) = 0$  for  $8 \leq t \leq 12$ . Note that the solution is unique.

Then assume that  $0 < l_0 < 8$ . In this case here exists a unique  $t_1 \in (0,4)$  such that  $8 - 2t_1 = l_0$ . From the analysis in §4 we have

$$\frac{dJ}{dt_1^i} = (U - D_1)[h(t_2 - t_1) - (c_1 - \beta)];$$

it follows that this inner derivative is zero, and moreover  $t_1$  defines the starting point of the optimal activity period. Also here, the solution  $x^*(t)$  is unique. It can be written as  $x^*(t) = X(t) - X(t_1)$  for  $t_1 \leq t \leq 8 - t_1$ , and otherwise  $x^*(t) = 0$ . In terms of the optimal control  $u^*(t)$  this means that

$u^*(t) = 4,5 (= U)$  for  $t_1 \leq t \leq 8 - t_1$  and  $u^*(t) = \min(U, d(t))$  otherwise.

The reader should draw a sketch of the graph of  $x^*(t)$ ! Observe that increasing  $h$  means decreasing  $l_0$ , i.e. increasing  $t_1$ , which means decreasing activity period, not too surprising!

Finally, consider the case of a free endpoint  $x^*(12)$ ! In the case  $c_0 \leq \beta$ , we know already that  $x^*(t) = 0$  for  $t \geq 8$ .

So, let  $c_0 > \beta$  and consider the “reduced intrinsic length”  $l_0' = \frac{1}{h}(c_0 - \beta) > 0$ .

According to Lemma 4 there exists a final activity period starting at some  $t_1^*$  which must necessarily satisfy  $11 \leq t_1^* < 12$ . To determine  $t_1^*$  completely recall the formula from

§6:  $\frac{dJ}{dt_1} = (U - D_1)[h(T - t_1) + \beta - c_0]$ . It can now be written as

$\frac{dJ}{dt_1} = (4,5 - 4) h (T - t_1 - l_0')$ . Thus, this expression is positive for  $t_1 < T - l_0'$  and negative for  $t_1 > T - l_0'$ . For optimum,  $t_1^*$  must clearly be chosen as close as possible to  $T - l_0'$ . Consequently, if  $11 \leq T - l_0'$ ,

then  $t_1^* = T - l_0'$ , and otherwise  $t_1^* = 11$ . The optimal solution is now completely determined and unique.

## References

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## Appendix 1

For the interested and practically oriented reader we give here a description, as simple as possible, of the optimal solution of the season problem in the main Case 1. Both cases 1A and 1B are covered.

All notation needed is found in §§ 1, 3 and 5.

Consider the guiding function  $X(t) = \int_0^t (U - d(s)) ds$  on the interval

I:  $0 \leq t \leq t_{M+N}$ , i.e. from start of the process up to the end of the high season. In Case 1 we have  $X(t_{M+N}) \leq 0$ . This means that the maximal productivity  $U$  is not higher than the average of the demand  $d(t)$  over this period of time. (A few words on Case 2 at the end below.)

The guiding function has a very simple structure on the interval I: it starts from zero at  $t = 0$  and increases strictly up to its maximum, denoted  $X_0$ , at  $t = t_M$ , after which it decreases strictly to zero at some time  $t^* \leq t_{M+N}$ .

Consider any level  $y$  such that  $0 < y < X_0$ . Clearly, there exist times

$t_1, t_2$ , such that  $0 < t_1 < t_2 < t^*$ , and  $X(t_1) = X(t_2) = y$ . Denote these times by  $t_1(y)$  and  $t_2(y)$ .

Assume now that  $t^* > l_0 =$  the intrinsic time length, a characteristic quantity for the problem. It was introduced in § 4 and defined as

$$l_0 = \frac{1}{h}(c_1 - \beta).$$

Clearly, there exists a level  $y_0$ , such that  $t_2(y_0) - t_1(y_0) = l_0$ .

The value  $t_1(y_0)$  is the starting time for the wanted optimal activity period, and it ends at time  $t_2(y_0)$ . During this time  $u(t) \equiv U$  and  $x(t) > 0$ .

On the other hand, if  $t^* \leq l_0$ , then  $t = 0$  is the optimal starting time and  $t^*$  is the appropriate end time.

The remaining intervals  $0 \leq t \leq t_1(y_0)$  and  $t_2(y_0) \leq t \leq t_{M+N}$  are passivity periods, i.e.  $x(t) = 0$  and  $u(t) = \min(U, d(t))$  here, hence no stock and selling directly from the production.

This specifies the optimal solution up to the end of the high season. The optimum over the final low season depends on the end condition as follows:

if the end condition is  $x(T) = 0$ , or if  $c_0 \leq \beta$  (rest value less than production cost), then  $x(t) \equiv 0$  for  $t_{M+N} \leq t \leq T$ , i.e. a passivity period.

But,

if  $x(T)$  is free and  $c_0 > \beta$ , then there is a final active period, starting at time  $t_3 = \max(t_{M+N}, T - l_0')$ . The quantity  $l_0'$  was defined in §6 as

$$l_0' = \frac{1}{h}(c_0 - \beta). \text{ Simply: produce what is motivated by the rest value!}$$

The interval  $t_{M+N} \leq t \leq t_3$  is a passivity period.

Finally, a few words on the main Case 2. It can be a little more complicated, depending on end conditions and rest value, but under end conditions A, the same as above, one

gets  $x(t) \equiv 0$  for  $t_{M+N} \leq t \leq T$  and one activity period “somewhere” on the interval  $0 \leq t \leq t_{M+N}$ . It is found by the same type of geometric procedure as in Case 1. Details in §7.

## Appendix 2

We include some more or less philosophical comments on our results.

It should be clear that extreme accuracy is not expected or demanded in this kind of economic optimization. It is certainly not like computing basic energy levels in atoms. The results in this work concerning activity periods and passivity periods (Theorem 2) should be seen as qualitative information on the nature of an optimal solution. The intrinsic time length is a natural and useful concept for understanding solutions. The complete solutions of the “seasonal” problem should be seen as useful suggestions for a good solution. The starting times computed cannot be more accurate than the given data for the demand periods, i.e. rather inaccurate. The fact that  $u(t) \equiv U$  during an activity period starting from zero gives some insight. For an activity period starting with a positive stock it holds that  $u(t) = 0$  up to a switch moment  $\lambda$ , after which  $u(t) = U$  holds. (It may occur that  $\lambda = 0$ , or  $\lambda = T$ , i.e. no switch.) For the single demand period problem there is an explicit solution formula available in most cases, but not in all. This can be helpful for understanding in general. The concept of a check-point gives qualitative information, which has been useful here. It is a kind of dimensional effect.

Many questions remain to be answered, for instance solving a 2-year problem.