

Article

An Inventory Model with Advertisement- and Customer-Relationship-Management-Sensitive Demand for a Product's Life Cycle

Mei-Chuan Cheng ¹, Chun-Tao Chang ² and Tsu-Pang Hsieh ^{3,*} 

¹ Department of Information Management, Fu Jen Catholic University, Xinzhuang Dist., New Taipei City 242062, Taiwan; 155379@mail.fju.edu.tw

² Department of Statistics, Tamkang University, Tamsui Dist., New Taipei City 251301, Taiwan; chuntao@mail.tku.edu.tw

³ Department of Business Administration, Aletheia University, Tamsui Dist., New Taipei City 25103, Taiwan

* Correspondence: tsupang@gmail.com

Abstract: Advertisements play an important role in communicating with target customers. A higher advertisement frequency increases costs but may increase the chances of acquiring new customers. Moreover, faced with a wide-ranging array of products that might fit specific needs, customers usually buy according to expectations about value and satisfaction. When customers are satisfied with a purchasing experience, they are more likely to buy again and share their experiences with others. Hence, companies are concerned about increasing customer value and service satisfaction to develop and manage customer relationships. This maintains a company's competitive edge and can improve its market share. In this article, we incorporate the frequency of advertisements and the cost of customer relationship management (CRM) into the demand function under a product life cycle (PLC). Customers can return products in the appreciation period offered by a retailer. A profit-maximizing model is developed to analyze the joint marketing and ordering policy of each stage of a product's life cycle with a product return guarantee. We construct an algorithm to identify the optimal decisions. Finally, numerical examples are presented to illustrate the proposed model, and managerial insights are obtained from a sensitivity analysis, followed by conclusions and future research.



Citation: Cheng, M.-C.; Chang, C.-T.; Hsieh, T.-P. An Inventory Model with Advertisement- and Customer-Relationship-Management-Sensitive Demand for a Product's Life Cycle.

Mathematics **2023**, *11*, 1555. <https://doi.org/10.3390/math11061555>

Academic Editor: Yong He

Received: 9 February 2023

Revised: 17 March 2023

Accepted: 19 March 2023

Published: 22 March 2023



Copyright: © 2023 by the authors. Licensee MDPI, Basel, Switzerland. This article is an open access article distributed under the terms and conditions of the Creative Commons Attribution (CC BY) license (<https://creativecommons.org/licenses/by/4.0/>).

Keywords: inventory; marketing; appreciation period; advertisement; return products

MSC: 90B05

1. Introduction

After a new product is launched, its sales grow, mature, and decline throughout the product's life cycle. Although it is unrealistic to expect a product to sell forever, a company may nevertheless seek to earn an adequate level of profitability to afford the effort and risk. Product sales over a typical product life cycle comprise five distinct stages, as shown in Figure 1. As a result, an appropriate inventory policy at each stage of a product's life cycle should be developed according to the level of product sales.

Over recent decades, researchers have discussed different patterns of demand matching at different stages of the product life cycle. The assumption of a constant demand in the traditional economic order quantity (EOQ) model only aligns with the maturity stage of a product's life cycle. In order to reflect the reality, the earlier research has generally assumed that demand rate is a linear or exponential function, increasing during the introduction and growth stages and decreasing during the decline stage. Donaldson [1] established a classical inventory model with a linear trend in demand and no shortages over a finite time-horizon. Henery [2] further expanded the demand rate from the work

of Donaldson [1] to develop general log-concave demand functions. Teng and Yang [3] developed a replenishment policy where the demand function fluctuated with time, and it included increasing, decreasing, and log-concave demand patterns. On the other hand, demand for a product over its life cycle may increase initially and then remain constant for a certain period before finally declining. Hill [4] and Ahmed et al. [5] proposed a ramp-type demand rate using the EOQ model in which the demand increased with time up to a certain level and then became constant over the rest of the period. To depict a product demand pattern more reasonably through a life cycle, Cheng and Wang [6] extended the deteriorating inventory model to allow for a trapezoidal-type demand rate, which included the stages from market introduction to decline over a product's life cycle. Skouri and Konstantaras [7] and Lin et al. [8] explored a time-varying deteriorating inventory model with a trapezoidal-type demand rate by introducing shortages. Wu et al. [9] discussed pricing and inventory decisions in an inventory model by considering that product freshness was linked to expiration date, and the demand rate was a trapezoidal, multivariate function of price and time. Inventory control requires maintaining an adequate stock level of a product to meet customer demand and, in turn, achieving the possible lowest cost or the highest possible profit. The analysis of the inventory models that allow for a constant rate of demand to vary over time has broadened the field of inventory management in practice.

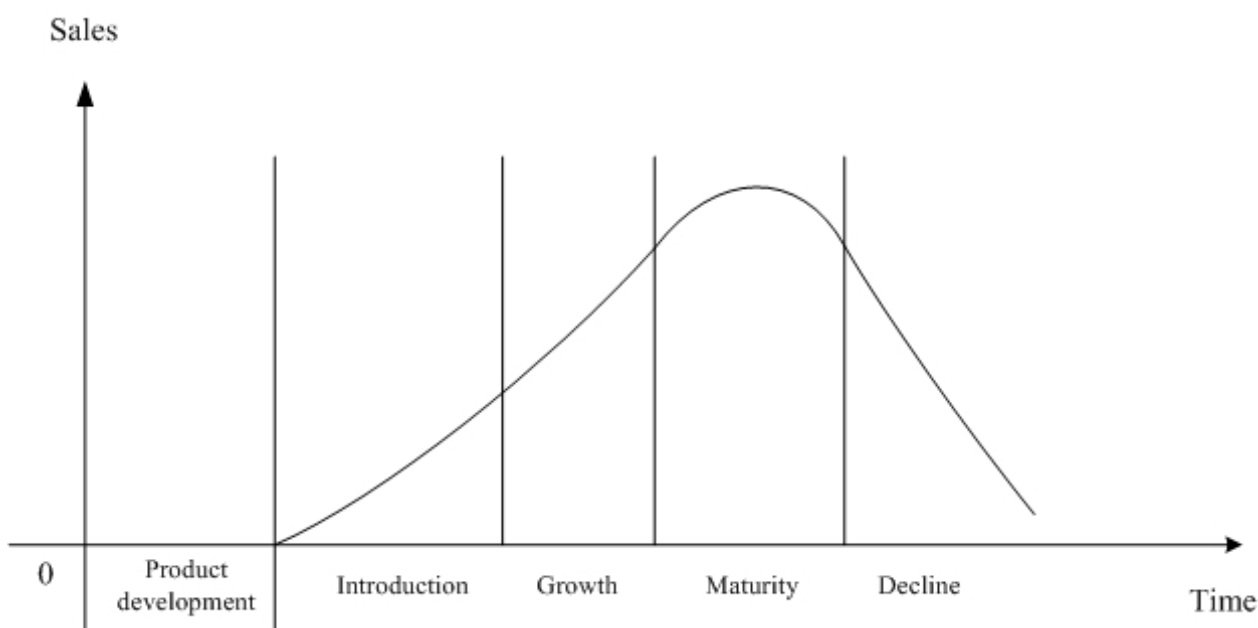


Figure 1. Sales over a product's life, from inception to decline.

In order to keep and sustain a competitive advantage, a retailer offers an appreciation period in which customers can request a return or exchange when they are dissatisfied with a product. According to online shopping surveys conducted by Pinkerton [10] and Trager [11], over 70% of the respondents evaluated the return conditions of a store before they made purchase decisions. Petersen and Kumar [12] demonstrated that a lenient return policy builds customer trust and convenience in shopping experiences, and hence, this leads to future repeat purchases. Yalabik et al. [13] introduced an optimal returns system that coordinated decisions in marketing and logistics. They identified that a refund policy decreases perceived risk and provides a protection mechanism for customers to go ahead with a purchase. Chen and Bell [14] designed a dual-channel structure with different returns policies—full-refund and no-returns—to examine the impact of customer returns on a firm's pricing and inventory decisions. They illustrated how a firm could enhance profitability using their dual-channel structure. Akçay et al. [15] considered a retailer who adopted a money-back guarantee policy and resold returned products by

discounting them as open-box items. They studied the retailer's optimal pricing, order quantities, and refunded amounts, and how the retailer's profit by the restock and resale of returned products had increased. Hu et al. [16] found that a vendor would not offer a return policy if the salvage value was zero in an inventory control problem under a consignment contract. In the research of Priyan and Uthayakumar [17], defective recoverable products were returned to a warehouse by a distributor, and the warehouse recovered them into perfect products under a recovery inventory system. Ülkü and Gürler [18] introduced a newsvendor model in which customers could decide to keep a product or exercise an abused or normal return after using it. Li et al. [19] studied a two-tier supply chain problem with a manufacturer and a retailer where the pricing and ordering decisions were made along with the retailer's product returns policy over two time periods. Assuming that customers could return a product at any stage of the replenishment cycle, Kumari and De [20] analyzed the impact of both returns and trade credit policies on the EOQ model, considering resalable returns for deteriorating items. Unlike the study of Kumari and De [20], Cheng and Ouyang [21] assumed that a retailer only offered a no-reason return period (an appreciation period) to its customers. They investigated the optimal pricing and ordering policies in an advanced sales system where a retailer could receive a supplier's trade credit. Later, considering an advance sales system within a no-reason return period and two-phase advance sales, Cheng et al. [22] established a deteriorating EOQ model to determine a retailer's optimal sales periods and selling prices.

Kotler and Armstrong [23] proposed that achieving marketing goals depends on not only attracting customers but also on maintaining and growing customer numbers. Delivering the desired satisfactions is a key influence on future buying behavior. To reach their target markets, companies rely heavily on advertising, which serves to inform, persuade, and remind consumers about their products or services. Through mass media channels, advertising effectively communicates a company's or brand's value proposition to potential customers. In addition, customer relationship management is an important business strategy for businesses looking to build strong, lasting relationships with their customers. When customers feel that a business understands and cares about their needs, they are more likely to stay loyal to that business. By doing so, advertising and customer relationship management help create a positive perception of a company or brand, which, in turn, leads to an increase in customer referrals through word of mouth. Urban [24] developed a finite replenishment inventory model by incorporating the influence of price variations and advertising expenditures on demand. Lee and Kim [25] examined the effects of integrated lot sizing and marketing expenditure decisions through an extended pricing-EOQ model context. Sadjadi et al. [26] considered an inventory model by assuming that production rate is a linear function of demand. Their model formulation determined a product's selling price, production lot size, and marketing expenditure. Wang et al. [27] established a newsvendor-type coordination model consisting of a manufacturer and a buyer by considering the combined effect of advertising expenditure and selling price on demand. Yadav et al. [28] investigated an inventory problem with backorders by assuming demand to depend on the frequency of advertisements in a fuzzy environment. Rabbani et al. [29] proposed a joint pricing and inventory control problem for non-instantaneous deteriorating items in which the demand rate was a function of both the selling price and the frequency of advertisements. Manna et al. [30] extended an imperfect production inventory model by accounting for an advertisement-dependent demand rate, which was assumed to be an increasing function of time at a decreasing rate. Shaikh et al. [31] considered the effects of selling price and frequency of advertisements on demand in a deteriorating inventory model which allowed for a permissible delay in payments and a time-proportional backlogging rate. Because advertisements can spread out the popularity of a product for all categories of customers, Khan et al. [32] expanded the deteriorating inventory model by considering an advertisement policy with an advanced payment policy. They assumed that the demand for a product with a maximum lifetime depended on the frequency of advertisements, which was confined to a positive integer, and the selling

price. San-José et al. [33] obtained an optimal advertising policy adopting a power form of demand dependent on the selling price, time, and frequency of advertisements in a lot-size inventory problem. Mandal and Pal [34] dealt with an EPQ model under an imperfect production system where demand was dependent on selling price and advertisements. In a two-level supply chain model with unreliable production process, Giri and Dash [35] derived an optimal batch shipment policy under price, advertisement frequency, and green-sensitive demand.

In reality, customer returns are the major category of returns in the supply chain. Retailers should use effective CRM to communicate their return policies across multiple channels. As a result, customer returns put pressure on a firm's costs arising from various activities, including CRMs and returns. In addition, advertising and CRMs are widely used by firms to communicate with customers in a competitive environment. Motivated by previous studies and investigations, this study aimed to address existing research gaps by developing an inventory model that links a firm's profit, incorporating the effects of advertising, with a CRM and an appreciation period. In the literature, advertising models with return policies have been extensively studied. However, there is a lack of research investigating the effects of advertising, CRMs, and appreciation periods within a product's life cycle. This study endeavored to address this gap by making contributions to three streams of literature: (1) an inventory model with a return policy, (2) an inventory model with advertisement-sensitive demand, and (3) an inventory model with time-dependent demand. In this paper, we discuss the inventory issues, including the product life cycle, and we incorporate the frequency of advertisements and the cost of customer relationship management into the demand function. In our model, a retailer offers an appreciation period in which customers can make a request to return products when they are dissatisfied with the products. A comparison of the present paper with the relevant literature is provided in Table 1 to help the readers understand the contributions of our model. We established a proper model and then provided an easy and useful algorithm to obtain the optimal ordering and advertisement policies for a retailer to achieve its maximum total profit. Finally, numerical examples are given to illustrate the solution procedure, and a sensitivity analysis is performed to investigate the effects of changes in some main parameter values on the optimal solution. This paper is organized as follows: Section 1 provides a comprehensive review of the related literature. Section 2 describes the problem and provides the developed model in detail. Section 3 presents the theoretical results and provides an algorithm for searching the optimal solutions. Section 4 illustrates numerical examples and a sensitivity analysis to provide further insights. Finally, Section 5 presents the conclusions and directions for future research.

Table 1. A comparison among the present model and related previous research.

Authors	Model Type	Return Policy	Advertisement-Dependent	CRM Cost-Dependent	Demand Rate	
					Appreciation Period-Dependent	Time-Dependent
Ahmed et al. [5]	EOQ	No	No	No	No	Ramp type
Ahmed et al. [8]	EOQ	No	No	No	No	Trapezoidal type
Wu et al. [9]	EOQ	No	No	No	No	Trapezoidal type
Li et al. [19]	A two-tier supply chain	Yes	No	No	No	No
Kumari and De [20]	EOQ	Yes	No	No	No	Exponential increasing/decreasing
Cheng et al. [22]	EOQ	Yes	No	No	No	No

Table 1. Cont.

Authors	Model Type	Return Policy	Advertisement-Dependent	Demand Rate		
				CRM Cost-Dependent	Appreciation Period-Dependent	Time-Dependent
Khan et al. [32]	EOQ	No	DV	No	No	No
San-Jose et al. [33]	EOQ	No	DV	No	No	Power pattern
Mandal and Pal [34]	EPQ	No	DV	No	No	No
Giri and Dash [35]	A two-level supply chain	No	DV	No	No	No
Present model	Single period	Yes	DV	DV	DV	Mixture

Note: “DV” represents “Decision variable”.

2. Problem Description and Model Formulation

2.1. Notation and Assumptions

The mathematical model in this paper was developed using the below notation and assumptions.

Notation:

p	selling price/unit
c	purchase cost/unit, where $c < p$
s	ordering cost/order
c_a	cost of each advertisement
h	holding cost/unit/unit time
N	appreciation period
ρ	return rate during the appreciation period, where $0 \leq \rho < 1$
$(\cdot)_i$	parameter or decision variable at the stage in the life cycle of the product, where $i = 1, 2, 3, 4$ corresponds to the four stages of introduction, growth, maturity, and decline
M_i	customer relationship management cost of stage i , $i = 1, 2, 3, 4$ /unit of time (we do not consider customer relationship management costs at the introduction stage, that is, $M_1 = 0$)
t_i	time length of stage i , $i = 1, 2, 3, 4$
T	length of a product's life cycle, $T = t_1 + t_2 + t_3 + t_4$
A_i	frequency of advertisements of stage i , $i = 1, 2, 3$ (we do not consider the frequency of advertisements at the decline stage, that is, $A_4 = 0$)
Q_i	order quantity in stage i , $i = 1, 2, 3, 4$
$Z_i(A_i, t_i, M_i)$	retailer total profit for stage i , $i = 1, 2, 3, 4$, where $M_1 = 0$ and $A_4 = 0$
Z	retailer total profit for a product's life cycle, $Z = Z_1 + Z_2 + Z_3 + Z_4$
t_i^*	optimal length of stage i for a product's life cycle, $i = 1, 2, 3, 4$
T^*	optimal product life cycle
M_i^*	optimal customer relationship management cost per unit of time for stage i , $i = 2, 3, 4$
A_i^*	optimal frequency of advertisements for stage i , $i = 1, 2, 3$
Q_i^*	optimal order quantity of stage i , $i = 1, 2, 3, 4$
Z_i^*	retailer optimal total profit for stage i , $Z_i^* = Z_i(A_i^*, t_i^*, M_i^*)$, $i = 1, 2, 3, 4$, where $M_1^* = 0$ and $A_4^* = 0$
Z^*	retailer optimal total profit for a product's life cycle, $Z^* = Z_1^* + Z_2^* + Z_3^* + Z_4^*$

Assumptions:

- (1) The demand rate is a function of time, appreciation period, customer relationship management cost, and frequency of advertisements (see, for example, Cheng and Wang [6], and Rabbani et al. [29]). The demand rates of different stages are as follows:
 - (1.1) Product introduction stage: The marketing objective is to create product awareness and trials. The demand rate is a function of time, appreciation period, and frequency of advertisements, that is, $D_1(A_1, t) = (a_1 + b_1 N t)(1 + A_1)^\lambda$, where $a_1 > 0$, $b_1 > 0$, $0 < \lambda < 1$, and $0 < t < t_1$.
 - (1.2) Product growth stage: The marketing objective is to maximize market share. The demand rate is a function of time, appreciation period, customer relationship management cost, and frequency of advertisements, that is, $D_2(A_2, t, M_2) = (a_2 + N^{b_2(t+M_2)})(1 + A_2)^\lambda$, where $a_2 > 0$, $b_2 > 0$, $0 < \lambda < 1$, and $0 < t < t_2$.
 - (1.3) Product maturity stage: The marketing objective is to maximize profit while defending market share. The demand rate is a function of the appreciation period, customer relationship management cost, and frequency of advertisements, that is, $D_3(A_3, M_3) = a_3(N + M_3)^{b_3}(1 + A_3)^\lambda$, where $a_3 > 0$, $b_3 > 0$, $0 < \lambda < 1$, and $0 < t < t_3$.
 - (1.4) Product decline stage: The marketing objective is to reduce expenditures and extract value from the brand. Sales may plunge to zero, or they may drop to a low level, where they will continue for many years. The demand rate is a function of time and customer relationship management cost. that is, $D_4(t, M_4) = a_4(N + M_4)\left(\frac{1}{b_4}\right)^t$, where $a_4 > 0$, $b_4 > 1$, and $0 < t < t_4$.
- (2) During the appreciation period N , customers can make a request to return products if they are dissatisfied with them. In such cases, the retailer will refund the money and the returned products will be discarded.

2.2. Mathematical Formulation

In this paper, we discuss the optimal ordering and marketing policies for four distinct stages of the product life cycle and the optimal ordering and marketing policies for a product's life cycle. The four distinct stages of the product life cycle are as follows: (1) introduction, (2) growth, (3) maturity, and (4) decline. We have developed retailer inventory models for these four distinct stages of the product life cycle, as follows:

Stage 1. Introduction

The demand rate is a function of time, appreciation period, and frequency of advertisements, that is, $D_1(A_1, t) = (a_1 + b_1 N t)(1 + A_1)^\lambda$, where $a_1 > 0$, $b_1 > 0$, $0 < \lambda < 1$, and $0 < t < t_1$. The order quantity in this stage is:

$$Q_1 = \int_0^{t_1} D_1(A_1, t) dt = \left(a_1 t_1 + \frac{b_1 N}{2} t_1^2\right) (1 + A_1)^\lambda.$$

The total profit consists of the following elements:

- (a) sales revenue (pQ_1)
- (b) purchasing cost (cQ_1)
- (c) returned product cost ($p\rho Q_1$)
- (d) advertisement cost ($c_a A_1$)
- (e) ordering cost (s)
- (f) holding cost, which is calculated by:

$$h \int_0^{t_1} D_1(A_1, t) t dt = h \left(\frac{a_1}{2} t_1^2 + \frac{b_1 N}{3} t_1^3 \right) (1 + A_1)^\lambda.$$

Therefore, a retailer's total profit for the introduction stage is:

$Z_1(A_1, t_1)$ = sales revenue – purchasing cost – cost of return products – advertisement cost – ordering cost – holding cost, which is determined by:

$$= (p - c - p\rho) \left(a_1 t_1 + \frac{b_1 N}{2} t_1^2 \right) (1 + A_1)^\lambda - c_a A_1 - s - h \left(\frac{a_1}{2} t_1^2 + \frac{b_1 N}{3} t_1^3 \right) (1 + A_1)^\lambda. \quad (1)$$

Stage 2. Growth

The demand rate is a function of time, appreciation period, customer relationship management cost, and frequency of advertisements, that is, $D_2(A_2, t, M_2) = \left(a_2 + N^{b_2(t+M_2)} \right) (1 + A_2)^\lambda$, where $a_2 > 0$, $b_2 > 0$, $0 < \lambda < 1$, and $0 < t < t_2$. The order quantity in this stage is:

$$Q_2 = \int_0^{t_2} D_2(A_2, t, M_2) dt = \left(a_2 t_2 + \frac{N^{b_2(t_2+M_2)}}{b_2 \ln N} - \frac{N^{b_2 M_2}}{b_2 \ln N} \right) (1 + A_2)^\lambda.$$

The total profit consists of the following elements:

- (a) sales revenue (pQ_2)
- (b) purchasing cost (cQ_2)
- (c) return products cost ($p\rho Q_2$)
- (d) advertisement cost ($c_a A_2$)
- (e) customer relationship management cost ($(1 - \rho)M_2 Q_2$)
- (f) ordering cost (s)
- (g) holding cost, which is calculated by:

$$h \int_0^{t_2} D_2(A_2, t, M_2) t dt = h \left[\frac{a_2}{2} t_2^2 + \frac{N^{b_2(t_2+M_2)}}{b_2 \ln N} t_2 - \frac{N^{b_2(t_2+M_2)}}{b_2^2 (\ln N)^2} + \frac{N^{b_2 M_2}}{b_2^2 (\ln N)^2} \right] (1 + A_2)^\lambda$$

Therefore, a retailer's total profit for the growth stage is:

$Z_2(A_2, t_2, M_2)$ = sales revenue – purchasing cost – cost of return products – advertisement cost – cost of customer relationship management – ordering cost – holding cost, which is determined by:

$$= [p - c - p\rho - (1 - \rho)M_2] \left(a_2 t_2 + \frac{N^{b_2(t_2+M_2)}}{b_2 \ln N} - \frac{N^{b_2 M_2}}{b_2 \ln N} \right) (1 + A_2)^\lambda - c_a A_2 - s - h \left[\frac{a_2}{2} t_2^2 + \frac{N^{b_2(t_2+M_2)}}{b_2 \ln N} t_2 - \frac{N^{b_2(t_2+M_2)}}{b_2^2 (\ln N)^2} + \frac{N^{b_2 M_2}}{b_2^2 (\ln N)^2} \right] (1 + A_2)^\lambda \quad (2)$$

Stage 3. Maturity

The demand rate is a function of the appreciation period, customer relationship management cost, and frequency of advertisements, that is, $D_3(A_3, M_3) = a_3 (N + M_3)^{b_3} (1 + A_3)^\lambda$, where $a_3 > 0$, $b_3 > 0$, $0 < \lambda < 1$, and $0 < t < t_3$. The order quantity in this stage is:

$$Q_3 = \int_0^{t_3} D_3(A_3, M_3) dt = a_3 (N + M_3)^{b_3} (1 + A_3)^\lambda t_3.$$

The total profit consists of the following elements:

- (a) sales revenue (pQ_3)
- (b) purchasing cost (cQ_3)

- (c) return products cost ($p\rho Q_3$)
- (d) advertisement cost ($c_a A_3$)
- (e) customer relationship management cost ($((1 - \rho)M_3 Q_3)$)
- (f) ordering cost (s)
- (g) holding cost, which is calculated by:

$$h \int_0^{t_3} D_3(A_3, M_3) t dt = \frac{h}{2} a_3 (N + M_3)^{b_3} (1 + A_3)^\lambda t_3^2.$$

Therefore, a retailer's total profit for the maturity stage is:

$Z_3(A_3, t_3, M_3)$ = sales revenue – purchasing cost – cost of return products—advertisement cost – cost of customer relationship management—ordering cost—holding cost, which is determined by:

$$= [p - c - p\rho - (1 - \rho)M_3] a_3 (N + M_3)^{b_3} (1 + A_3)^\lambda t_3 - c_a A_3 - s - \frac{h}{2} a_3 (N + M_3)^{b_3} (1 + A_3)^\lambda t_3^2. \quad (3)$$

Stage 4. Decline

The demand rate is a function of time and customer relationship management cost, that is, $D_4(t, M_4) = a_4(N + M_4) \left(\frac{1}{b_4}\right)^t$, where $a_4 > 0$, $b_4 > 1$, and $0 < t < t_4$. The order quantity in this stage is calculated by:

$$Q_4 = \int_0^{t_4} D_4(t, M_4) dt = \frac{a_4(N + M_4)}{\ln b_4} (1 - b_4^{-t_4}).$$

The total profit consists of the following elements:

- (a) sales revenue (pQ_4)
- (b) purchasing cost (cQ_4)
- (c) return products cost ($p\rho Q_4$)
- (d) customer relationship management cost ($((1 - \rho)M_4 Q_4)$)
- (e) ordering cost (s)
- (f) holding cost, which is calculated by:

$$h \int_0^{t_4} D_4(t, M_4) t dt = h a_4 (N + M_4) \left[\frac{1}{(\ln b_4)^2} - \frac{t_4 b_4^{-t_4}}{\ln b_4} - \frac{b_4^{-t_4}}{(\ln b_4)^2} \right].$$

Therefore, a retailer's total profit for the decline stage is:

$Z_4(t_4, M_4)$ = sales revenue – purchasing cost – cost of return products—cost of customer relationship management – ordering cost – holding cost, which is determined by:

$$= [p - c - p\rho - (1 - \rho)M_4] \frac{a_4(N + M_4)}{\ln b_4} (1 - b_4^{-t_4}) - s - h a_4 (N + M_4) \left[\frac{1}{(\ln b_4)^2} - \frac{t_4 b_4^{-t_4}}{\ln b_4} - \frac{b_4^{-t_4}}{(\ln b_4)^2} \right]. \quad (4)$$

Furthermore, a retailer's total profit for a product's life cycle is:

$$Z(A_1, A_2, A_3, t_1, t_2, t_3, t_4, M_2, M_3, M_4) = Z_1(A_1, t_1) + Z_2(A_2, t_2, M_2) + Z_3(A_3, t_3, M_3) + Z_4(t_4, M_4). \quad (5)$$

3. Theoretical Results

In this section, the solution procedure is presented and an algorithm is established to determine the optimal solution to the aforementioned cases. The goal was to maximize retailer total profit. First, we found the optimal solution that could maximize the retailer total profit for the four distinct stages of the product life cycle. We, then proceeded to use the software program Mathematica 13.0 to obtain the optimal solu-

tion, $A_1^*, A_2^*, A_3^*, t_1^*, t_2^*, t_3^*, t_4^*, M_2^*, M_3^*, M_4^*$, and the maximum retailer total profit for a product's life cycle, $Z^*(A_1^*, A_2^*, A_3^*, t_1^*, t_2^*, t_3^*, t_4^*, M_2^*, M_3^*, M_4^*)$. Then, for fixed $A_i, i = 1, 2, 3$, we obtained the optimal solution that could maximize retailer total profit for the four distinct stages of the product life cycle, as follows:

Stage 1. Introduction

To maximize the total profit, for fixed $A_1 = A_1^*$, taking the first-order and second-order derivatives of $Z_1(t_1|A_1)$ in (1) with respect to t_1 , we obtain

$$\frac{\partial Z_1}{\partial t_1} = (p - c - p\rho - ht_1)(a_1 + b_1 N t_1)(1 + A_1)^\lambda, \quad (6)$$

and

$$\frac{\partial^2 Z_1}{\partial t_1^2} = [(p - c - p\rho)b_1 N - h(a_1 + 2b_1 N t_1)](1 + A_1)^\lambda. \quad (7)$$

If $\frac{\partial^2 Z_1}{\partial t_1^2} < 0$, then $Z_1(t_1|A_1)$ is a concave function of t_1 and $Z_1(t_1|A_1)$ has its maximum value at

$$t_1 = \frac{p - c - p\rho}{h}. \quad (8)$$

Stage 2. Growth

To maximize the total profit, for fixed $A_2 = A_2^*$, taking the first-order derivative of $Z_2(t_2, M_2|A_2)$ in (2) with respect to t_2 and M_2 , respectively, we obtain

$$\frac{\partial Z_2}{\partial t_2} = [p - c - p\rho - M_2(1 - \rho) - ht_2] \left[a_2 + N^{b_2(t_2+M_2)} \right] (1 + A_2)^\lambda \quad (9)$$

and

$$\begin{aligned} \frac{\partial Z_2}{\partial M_2} = & [p - c - p\rho - M_2(1 - \rho)] N^{b_2 M_2} (N^{b_2 t_2} - 1) (1 + A_2)^\lambda \\ & - h \left[t_2 N^{b_2(t_2+M_2)} - \frac{N^{b_2(t_2+M_2)}}{b_2 \ln N} + \frac{N^{b_2 M_2}}{b_2 \ln N} \right] (1 + A_2)^\lambda \\ & - (1 - \rho) \left[a_2 t_2 + \frac{N^{b_2(t_2+M_2)} - N^{b_2 M_2}}{b_2 \ln N} \right] (1 + A_2)^\lambda. \end{aligned} \quad (10)$$

The optimal solution of (t_2, M_2) must simultaneously satisfy $\frac{\partial Z_2}{\partial t_2} = 0$ and $\frac{\partial Z_2}{\partial M_2} = 0$. In addition,

$$\begin{aligned} \frac{\partial^2 Z_2}{\partial t_2^2} = & [p - c - p\rho - M_2(1 - \rho)] N^{b_2(t_2+M_2)} b_2 \ln N (1 + A_2)^\lambda \\ & - h \left[a_2 + N^{b_2(t_2+M_2)} + t_2 N^{b_2(t_2+M_2)} b_2 \ln N \right] (1 + A_2)^\lambda, \end{aligned} \quad (11)$$

$$\begin{aligned} \frac{\partial^2 Z_2}{\partial M_2^2} = & [p - c - p\rho - M_2(1 - \rho)] b_2 \ln N \left[N^{b_2(t_2+M_2)} - N^{b_2 M_2} \right] (1 + A_2)^\lambda \\ & - 2(1 - \rho) \left[N^{b_2(t_2+M_2)} - N^{b_2 M_2} \right] (1 + A_2)^\lambda \\ & - h \left[t_2 N^{b_2(t_2+M_2)} b_2 \ln N - N^{b_2(t_2+M_2)} + N^{b_2 M_2} \right] (1 + A_2)^\lambda, \end{aligned} \quad (12)$$

and

$$\begin{aligned} \frac{\partial^2 Z_2}{\partial M_2 \partial t_2} = & [p - c - p\rho - M_2(1 - \rho) - ht_2] N^{b_2(t_2+M_2)} b_2 \ln N (1 + A_2)^\lambda \\ & - (1 - \rho) \left[a_2 + N^{b_2(t_2+M_2)} \right] (1 + A_2)^\lambda. \end{aligned} \quad (13)$$

Due to the complexity of the problem, we could not find a simple result in which the Hessian matrix was a positive value. However, using Equations (11)–(13), we could check this condition by using the software Mathematica 13.0 in numerical examples.

Stage 3. Maturity

To maximize the total profit, for fixed $A_3 = A_3^*$, taking the first-order derivative of $Z_3(t_3, M_3|A_3)$ in (3) with respect to t_3 and M_3 , respectively, we obtain

$$\frac{\partial Z_3}{\partial t_3} = [p - c - p\rho - M_3(1 - \rho) - ht_3] a_3 (N + M_3)^{b_3} (1 + A_3)^\lambda \quad (14)$$

and

$$\frac{\partial Z_3}{\partial M_3} = [p - c - p\rho - M_3(1 - \rho)]a_3b_3(N + M_3)^{b_3-1}(1 + A_3)^\lambda t_3 - \frac{h}{2}a_3b_3(N + M_3)^{b_3-1}(1 + A_3)^\lambda t_3^2 - (1 - \rho)a_3(N + M_3)^{b_3}(1 + A_3)^\lambda t_3. \quad (15)$$

The optimal solution of (t_3, M_3) must simultaneously satisfy $\frac{\partial Z_3}{\partial t_3} = 0$ and $\frac{\partial Z_3}{\partial M_3} = 0$. We obtain

$$t_3 = \frac{p - c - p\rho - M_3(1 - \rho)}{h} \quad (16)$$

and

$$M_3 = \frac{(p - c - p\rho - \frac{h}{2}t_3)b_3 - (1 - \rho)N}{(1 - \rho)(1 + b_3)}. \quad (17)$$

In addition,

$$\frac{\partial^2 Z_3}{\partial t_3^2} = -ha_3(N + M_3)^{b_3}(1 + A_3)^\lambda < 0, \quad (18)$$

$$\begin{aligned} \frac{\partial^2 Z_3}{\partial M_3^2} = & [p - c - p\rho - M_3(1 - \rho)]a_3b_3(b_3 - 1)(N + M_3)^{b_3-2}(1 + A_3)^\lambda t_3 \\ & - 2(1 - \rho)a_3b_3(N + M_3)^{b_3-1}(1 + A_3)^\lambda t_3 \\ & - \frac{h}{2}a_3b_3(b_3 - 1)(N + M_3)^{b_3-2}(1 + A_3)^\lambda t_3^2, \end{aligned} \quad (19)$$

and

$$\begin{aligned} \frac{\partial^2 Z_3}{\partial t_3 \partial M_3} = & [p - c - p\rho - M_3(1 - \rho)]a_3b_3(N + M_3)^{b_3-1}(1 + A_3)^\lambda \\ & - (1 - \rho)a_3(N + M_3)^{b_3}(1 + A_3)^\lambda - ha_3b_3(N + M_3)^{b_3-1}(1 + A_3)^\lambda t_3 \\ = & \{[p - c - p\rho - M_3(1 - \rho) - ht_3]b_3 - (1 - \rho)(N + M_3)\} \\ & \times a_3(1 + A_3)^\lambda (N + M_3)^{b_3-1}. \end{aligned} \quad (20)$$

Due to the complexity of the problem, we were unable to find a simple result in which the Hessian matrix was a positive value. However, using Equations (18)–(20), we could check this condition by using the software Mathematica 13.0 in numerical examples.

Stage 4. Decline

To maximize the total profit, taking the first-order derivative of $Z_4(t_4, M_4)$ in (4) with respect to t_4 and M_4 , respectively, we obtain

$$\frac{\partial Z_4}{\partial t_4} = [p - c - p\rho - M_4(1 - \rho) - ht_4]a_4(N + M_4)b_4^{-t_4} \quad (21)$$

and

$$\begin{aligned} \frac{\partial Z_4}{\partial M_4} = & [p - c - p\rho - (1 - \rho)(2M_4 + N)]\frac{a_4}{\ln b_4}\left(1 - b_4^{-t_4}\right) \\ & - ha_4\left[\frac{1}{(\ln b_4)^2} - \frac{t_4 b_4^{-t_4}}{\ln b_4} - \frac{b_4^{-t_4}}{(\ln b_4)^2}\right]. \end{aligned} \quad (22)$$

The optimal solution of (t_4, M_4) must simultaneously satisfy $\frac{\partial Z_4}{\partial t_4} = 0$ and $\frac{\partial Z_4}{\partial M_4} = 0$. We obtain

$$t_4 = \frac{p - c - p\rho - M_4(1 - \rho)}{h} \quad (23)$$

and

$$M_4 = \frac{(1 - \rho)(p - N) - c}{2(1 - \rho)} - \frac{h}{2(1 - \rho)}\left[\frac{1}{\ln b_4} - \frac{t_4 b_4^{-t_4}}{1 - b_4^{-t_4}}\right]. \quad (24)$$

In addition,

$$\frac{\partial^2 Z_4}{\partial M_4^2} = -2(1 - \rho)a_4\frac{1 - b_4^{-t_4}}{\ln b_4} < 0, \quad (25)$$

$$\frac{\partial^2 Z_4}{\partial t_4^2} = -\{[p - c - p\rho - M_4(1 - \rho) - ht_4] \ln b_4 + h\} a_4 (N + M_4) b_4^{-t_4} < 0, \quad (26)$$

and

$$\frac{\partial^2 Z_4}{\partial t_4 \partial M_4} = a_4 b_4^{-t_4} [p - c - p\rho - (1 - \rho)(2M_4 + N) - ht_4]. \quad (27)$$

Due to the complex nature of the problem, we could not find a simple result in which the Hessian matrix was a positive value. However, using Equations (25)–(27), we could check this condition by using the software Mathematica 13.0 in numerical examples.

Algorithm 1. Solution Procedure

Step 1. Input the values of parameters

Step 2. Introduction stage

- 2.1 Set $i = 0$ and $A_{1,i} = 0$.
- 2.2 Using Equation (8), $t_{1,i}$ is obtained, and substituting $t_{1,i}$ into Equation (7), if $\frac{\partial^2 Z_1}{\partial t_1^2} < 0$, then, by substituting $t_{1,i}$ into Equation (1), $Z_{1,i}$ is obtained. Otherwise, $Z_{1,i} = 0$.
- 2.3 Let $i = i + 1$ and $A_{1,i+1} = i + 1$. Using Equation (8), $t_{1,i+1}$ is obtained, and substituting $t_{1,i+1}$ into Equation (7), if $\frac{\partial^2 Z_1}{\partial t_1^2} < 0$, then, by substituting $t_{1,i+1}$ into Equation (1), $Z_{1,i+1}$ is obtained. Otherwise, $Z_{1,i+1} = 0$.
- 2.4 If $Z_{1,i+1} > Z_{1,i}$, we can proceed to 2.3. Otherwise, $Z_{1,i} = Z_1^*$, $t_{1,i} = t_1^*$, and $A_{1,i} = A_1^*$. Proceed to Step 6.

Step 3. Growth stage

- 3.1 Set $i = 0$ and $A_{2,i} = 0$.
- 3.2 Using Equations (9) and (10), and by simultaneously solving $\frac{\partial Z_2}{\partial t_2} = 0$ and $\frac{\partial Z_2}{\partial M_2} = 0$, we obtain $t_{2,i}$ and $M_{2,i}$. Substituting $t_{2,i}$ and $M_{2,i}$ into Equations (11)–(13), if the Hessian matrix has a positive value, then, by substituting $t_{2,i}$ and $M_{2,i}$ into (2), $Z_{2,i}$ is obtained. Otherwise, $Z_{2,i} = 0$.
- 3.3 Let $i = i + 1$ and $A_{2,i+1} = i + 1$. Using Equations (9) and (10), and by simultaneously solving $\frac{\partial Z_2}{\partial t_2} = 0$ and $\frac{\partial Z_2}{\partial M_2} = 0$, we obtain $t_{2,i+1}$ and $M_{2,i+1}$. Substituting $t_{2,i+1}$ and $M_{2,i+1}$ into Equations (11)–(13), if the Hessian matrix has a positive value, then, by substituting $t_{2,i+1}$ and $M_{2,i+1}$ into (2), $Z_{2,i+1}$ is obtained. Otherwise, $Z_{2,i+1} = 0$.
- 3.4 If $Z_{2,i+1} > Z_{2,i}$, we can proceed to 3.3. Otherwise, $Z_{2,i} = Z_2^*$, $t_{2,i} = t_2^*$, $M_{2,i} = M_2^*$, and $A_{2,i} = A_2^*$. Proceed to Step 6.

Step 4. Maturity stage

- 4.1 Set $i = 0$ and $A_{3,i} = 0$.
- 4.2 Using Equations (16) and (17), we obtain $t_{3,i}$ and $M_{3,i}$. Substituting $t_{3,i}$ and $M_{3,i}$ into Equations (18)–(20), if the Hessian matrix has a positive value, then, by substituting $t_{3,i}$ and $M_{3,i}$ into (3), $Z_{3,i}$ is obtained. Otherwise, $Z_{3,i} = 0$.
- 4.3 Let $i = i + 1$ and $A_{3,i} = i + 1$. Using Equations (16) and (17), we obtain $t_{3,i+1}$ and $M_{3,i+1}$. Substituting $t_{3,i+1}$ and $M_{3,i+1}$ into Equations (18)–(20), if the Hessian matrix has a positive value, then, by substituting $t_{3,i+1}$ and $M_{3,i+1}$ into (3), $Z_{3,i+1}$ is obtained. Otherwise, $Z_{3,i+1} = 0$.
- 4.4 If $Z_{3,i+1} > Z_{3,i}$, we can proceed to 4.3. Otherwise, $Z_{3,i} = Z_3^*$, $t_{3,i} = t_3^*$, $M_{3,i} = M_3^*$, and $A_{3,i} = A_3^*$. Proceed to Step 6.

Step 5. Decline stage

- 5.1 Using Equations (23) and (24), we obtain t_4 and M_4 .
- 5.2 Substituting t_4 and M_4 into Equations (25)–(27), if the Hessian matrix has a positive value, then $t_4 = t_4^*$ and $M_4 = M_4^*$. Otherwise, $Z_4 = 0$.
- 5.3 By substituting t_4^* and M_4^* into (4), Z_4^* is obtained. Proceed to Step 6.

Step 6. $Z^* = Z(A_1^*, A_2^*, A_3^*, t_1^*, t_2^*, t_3^*, t_4^*, M_2^*, M_3^*, M_4^*) = Z_1^*(A_1^*, t_1^*) + Z_2^*(A_2^*, t_2^*, M_2^*) + Z_3^*(A_3^*, t_3^*, M_3^*) + Z_4^*(t_4^*, M_4^*)$.

4. Numerical Examples

Example 1. A retailer offers customers an appreciation period. We established proper models for four different stages of a product's life cycle and incorporated the frequency of advertisements and the cost of customer relationship management into the demand function. The given parameter values were as follows:

$p = \$27/\text{unit}$, $c = \$7/\text{unit}$, $h = \$1.2/\text{unit/day}$, $\rho = 0.05$, $N = 7 \text{ days}$, $\lambda = 0.6$, $c_a = \$300$, $s = \$200$, $a_1 = 3$, $a_2 = 2$, $a_3 = 2$, $a_4 = 2$, $b_1 = 0.1$, $b_2 = 0.13$, $b_3 = 0.85$, and $b_4 = 1.05$.

By applying the solution procedure in Algorithm 1, we determined that the optimal solution was $(A_1^*, A_2^*, A_3^*, t_1^*, t_2^*, t_3^*, t_4^*, M_2^*, M_3^*, M_4^*) = 4, 7, 15, 15.54, 13.93, 14.80, 13.56, 2.03, 0.94, 2.50$, $Z(A_1^*, A_2^*, A_3^*, t_1^*, t_2^*, t_3^*, t_4^*, M_2^*, M_3^*, M_4^*) = 7717.06$, and the optimal order quantity of each stage of a product's life cycle was $(Q_1^*, Q_2^*, Q_3^*, Q_4^*) = (344.5, 855.2, 909.1, 188.5) \text{ units}$.

Example 2. We discussed the effects of changing the values of the parameters (the selling price p , the purchase cost c , the holding cost h , the return rate ρ , the appreciation period N , and the cost of each advertisement c_a) on the optimal solution. The remaining parameter values were identical to those in Example 1. The sensitivity analysis was performed by changing $p \in \{25, 26, 27, 28, 29\}$, $c \in \{5, 6, 7, 8, 9\}$, $h \in \{1.0, 1.1, 1.2, 1.3, 1.4\}$, $\rho \in \{0.01, 0.02, 0.05, 0.1, 0.15\}$, $N \in \{5, 6, 7, 8, 9\}$, and $c_a \in \{200, 250, 300, 350, 400\}$ (one at a time) while keeping the remaining parameters unchanged. We followed the algorithms, and the computational results are shown in Tables 1 and 2.

Table 2. Effects of the parameters on A_i^* , t_i^* , and M_i^* .

Parameter		A_1^*	A_2^*	A_3^*	t_1^*	t_2^*	t_3^*	t_4^*	M_2^*	M_3^*	M_4^*
p	25	2	2	8	13.96	13.96	13.68	12.57	0.00	0.35	1.75
	26	3	4	11	14.75	14.75	14.24	13.07	0.00	0.64	2.12
	27	4	7	15	15.54	13.93	14.80	13.56	2.03	0.94	2.50
	28	6	13	20	16.33	12.30	15.35	14.05	5.10	1.24	2.88
	29	9	23	26	17.13	11.65	15.91	14.54	6.92	1.54	3.26
c	5	9	25	27	17.21	11.60	15.96	14.59	7.09	1.57	3.30
	6	6	13	21	16.38	12.25	15.38	14.08	5.21	1.26	2.90
	7	4	7	15	15.54	13.93	14.80	13.56	2.03	0.94	2.50
	8	3	4	11	14.71	14.71	14.21	13.04	0.00	0.63	2.10
	9	2	2	8	13.88	13.88	13.63	12.52	0.00	0.31	1.71
h	1.0	9	31	25	18.65	18.65	17.75	16.17	0.00	0.94	2.62
	1.1	6	14	19	16.95	16.95	16.14	14.75	0.00	0.94	2.55
	1.2	4	7	15	15.54	13.93	14.80	13.56	2.03	0.94	2.50
	1.3	3	4	12	14.35	10.92	13.66	12.55	4.69	0.94	2.46
	1.4	2	2	10	13.32	9.33	12.68	11.68	5.88	0.94	2.42
ρ	0.01	6	13	21	16.44	16.34	15.59	14.26	0.12	1.03	2.64
	0.02	6	11	19	16.22	15.41	15.39	14.09	0.99	1.01	2.61
	0.05	4	7	15	15.54	13.93	14.80	13.56	2.03	0.94	2.50
	0.10	2	3	10	14.42	12.46	13.80	12.68	2.61	0.82	2.32
	0.15	1	1	6	13.29	11.69	12.81	11.79	2.26	0.68	2.12
N	5	2	3	8	15.54	15.54	13.68	12.57	0.00	2.35	3.75
	6	3	5	11	15.54	15.54	14.24	13.07	0.00	1.64	3.12
	7	4	7	15	15.54	13.93	14.80	13.56	2.03	0.94	2.50
	8	5	11	20	15.54	11.25	15.35	14.05	5.42	0.24	1.88
	9	6	17	26	15.54	10.11	15.54	14.54	6.86	0.00	1.26
c_a	200	13	21	44	15.54	13.93	14.80	13.56	2.03	0.94	2.50
	250	7	12	25	15.54	13.93	14.80	13.56	2.03	0.94	2.50
	300	4	7	15	15.54	13.93	14.80	13.56	2.03	0.94	2.50
	350	2	5	10	15.54	13.93	14.80	13.56	2.03	0.94	2.50
	400	2	3	7	15.54	13.93	14.80	13.56	2.03	0.94	2.50

From Tables 2 and 3, we obtained the following results:

- (1) A higher selling price value p resulted in a higher optimal frequency of advertisements for stage i , A_i^* ; a higher optimal customer relationship management cost value for stage i , M_i^* ; a higher retailer optimal total profit value for stage i , Z_i^* ; a higher value for retailer total profit for a product life cycle Z^* ; and an optimal order quantity of each stage for a product life cycle Q_i^* . This suggests that when a selling price is higher, a retailer should increase the frequency of advertisements and customer relationship management cost.
- (2) A higher purchase cost value c resulted in a lower optimal frequency of advertisements for stage i , A_i^* ; a lower optimal customer relationship management cost value for stage i , M_i^* ; a lower retailer optimal total profit value for stage i , Z_i^* ; a lower value for retailer total profit for a product life cycle Z^* ; and an optimal order quantity of each stage for a product life cycle Q_i^* . This suggests that when a purchase cost is higher, a retailer should consider decreasing the frequency of advertisements and customer relationship management cost.
- (3) In addition, a higher holding cost value h resulted in a lower optimal frequency of advertisements for stage i , A_i^* ; a lower optimal length of stage i for a product life cycle t_i^* ; a lower retailer optimal total profit value for stage i , Z_i^* ; a lower value for retailer total profit for a product life cycle Z^* ; and an optimal order quantity of each stage for a product life cycle Q_i^* . This suggests a retailer should consider reducing their order quantity to avoid incurring higher holding costs. Additionally, they may also need to shorten the length of stage i of the product life cycle.
- (4) Moreover, a higher return rate value for ρ resulted in a lower optimal frequency of advertisements for stage i , A_i^* ; a lower optimal length of stage i for a product life cycle t_i^* ; a lower retailer optimal total profit value for stage i , Z_i^* ; a lower value for retailer total profit for a product life cycle Z^* ; and an optimal order quantity for each stage of a product life cycle Q_i^* . This suggests that retailers should consider reducing their order quantity to avoid incurring higher costs associated with returned products. Additionally, they may also need to shorten the length of stage i of a product's life cycle to mitigate the impact of returns on their overall profits.
- (5) A higher appreciation period value N resulted in a higher optimal frequency of advertisements for stage i , A_i^* ; a higher retailer optimal total profit value for stage i , Z_i^* ; a higher value for retailer total profit for a product life cycle Z^* ; and an optimal order quantity for each stage of a product life cycle Q_i^* . This suggests that when an appreciation period is longer, a retailer may need to consider increasing the frequency of advertisements and the order quantity of each stage of a product's life cycle to maximize their profits.
- (6) A higher cost value of each advertisement c_a resulted in a lower optimal frequency of advertisements for stage i , A_i^* ; a lower retailer optimal total profit value for stage i , Z_i^* ; a lower value for retailer total profit for a product life cycle Z^* ; and an optimal order quantity for each stage of a product life cycle Q_i^* . As a result, one straightforward management approach for retailers could be to reduce the frequency of advertisements for stage i , A_i^* , to avoid incurring higher advertisement costs. Additionally, a retailer may explore alternative advertising strategies to reduce the cost per advertisement.

Table 3. Effects of the parameters on Z^* , Z_i^* , and Q_i^* .

Parameter		Z^*	Z_1^*	Z_2^*	Z_3^*	Z_4^*	Q_1^*	Q_2^*	Q_3^*	Q_4^*
p	25	4481.00	614.01	726.02	1974.28	1166.69	212.8	307.4	557.1	164.4
	26	5857.46	831.83	1108.14	2588.95	1328.53	276.6	500.4	712.7	176.3
	27	7717.06	1122.35	1722.82	3370.07	1501.82	344.5	855.2	909.1	188.5
	28	10,408.37	1504.11	2864.78	4352.66	1686.81	457.6	1619.1	1145.7	201.0
	29	14,424.88	2000.63	4961.96	5578.49	1883.80	613.2	2920.2	1422.8	213.8
c	5	14,957.40	2062.19	5266.02	5723.95	1905.24	618.1	3149.2	1464.0	215.1
	6	10,580.61	1527.04	2946.32	4410.37	1696.88	459.5	1642.6	1182.3	201.7
	7	7717.06	1122.35	1722.82	3370.07	1501.82	344.5	855.2	909.1	188.5
	8	5774.39	818.03	1083.24	2553.39	1319.73	275.3	495.6	710.0	175.7
	9	4357.50	592.84	695.57	1918.79	1150.31	210.7	301.6	552.7	163.2
h	1.0	15,773.50	2189.95	6560.35	5258.98	1764.23	707.4	3806.5	1459.8	215.0
	1.1	10,429.25	1534.74	3105.83	4164.96	1623.72	486.9	1615.2	1133.8	200.9
	1.2	7717.06	1122.35	1722.82	3370.07	1501.82	344.5	855.2	909.1	188.5
	1.3	6139.31	850.83	1116.69	2776.72	1395.07	264.4	561.7	740.9	177.5
	1.4	5082.40	666.88	790.16	2324.52	1300.84	197.3	360.7	622.3	167.7
ρ	0.01	10,563.00	1563.92	2857.64	4450.71	1690.73	462.6	1378.9	1170.7	198.1
	0.02	9755.01	1440.42	2514.89	4156.98	1642.73	452.2	1226.0	1089.0	195.7
	0.05	7717.06	1122.35	1722.82	3370.07	1501.82	344.5	855.2	909.1	188.5
	0.10	5277.48	734.18	930.86	2334.85	1277.59	224.2	450.5	668.4	176.2
	0.15	3645.92	488.30	518.40	1580.10	1067.12	154.2	229.1	465.9	163.5
N	5	4832.35	766.22	925.16	1974.28	1166.69	206.9	344.1	557.1	164.4
	6	6121.17	933.79	1269.90	2588.95	1328.53	273.6	548.2	712.7	176.3
	7	7717.06	1122.35	1722.82	3370.07	1501.82	344.5	855.2	909.1	188.5
	8	9880.66	1333.93	2507.25	4352.66	1686.81	419.7	1517.4	1145.7	201.0
	9	12,720.67	1569.24	3712.59	5555.04	1883.80	499.2	2500.9	1453.6	213.7
c_a	200	12,369.85	1878.43	2981.26	6008.35	1501.82	639.0	1569.2	1690.7	188.5
	250	9428.28	1394.09	2183.23	4349.15	1501.82	456.7	1144.5	1216.5	188.5
	300	7717.06	1122.35	1722.82	3370.07	1501.82	344.5	855.2	909.1	188.5
	350	6638.66	956.50	1435.07	2745.27	1501.82	253.6	719.7	726.1	188.5
	400	5936.66	856.50	1254.07	2324.26	1501.82	253.6	564.2	599.8	188.5

5. Conclusions

Advertising, return policies, and customer relationship management (CRM) are all crucial components of running a successful business. Advertising increases brand awareness, builds brand credibility, and drives sales and revenue. Meanwhile, a return policy helps build trust with customers by assuring them that they can purchase with confidence, and it can improve customer satisfaction by addressing their needs. A return policy can also encourage sales by giving customers the confidence to try a product without fear of being stuck with something they do not want. Additionally, CRMs can help businesses deliver personalized, responsive, and proactive customer service, which can increase customer satisfaction and loyalty. Incorporating the frequency of advertisements and the cost of CRMs into the demand function under the product life cycle (PLC) can help retailers understand how these factors impact consumer behavior. Retailers can also offer return guarantees during the appreciation period to further boost customer confidence in their products. If customers are dissatisfied and return products, retailers can refund the money and discard the returned products. Overall, by effectively utilizing advertising, return policies, and CRMs, businesses can improve customer satisfaction, drive sales, and build loyal customer bases.

In this paper, we discussed the optimal ordering and marketing policies for four distinct stages of the product life cycle and the optimal ordering and marketing policies for a product's life cycle. The four distinct stages of the product life cycle include introduction, growth, maturity, and decline. We provided an easy and useful algorithm to identify the optimal ordering and advertisement policies. Finally, numerical examples are provided to illustrate the solution procedure. The results of the sensitivity analysis showed that

a retailer needs to increase the frequency of advertisements and increase the customer relationship management cost when a product's selling price is higher. However, a retailer should decrease the frequency of advertisements and decrease the customer relationship management cost when a product's purchase cost, holding cost, or return rate is higher. In addition, a retailer needs to reduce their order quantity to avoid higher holding costs (or higher costs of return products) and shorten the length of stage i of the product life cycle when the holding cost (or return rate) is higher. A retailer needs to increase the frequency of advertisements and the order quantity of each stage of a product's life cycle when the appreciation period is longer. However, retailers need to reduce the frequency of advertisements to avoid higher advertisement costs when the cost of each advertisement is higher. A retailer can also find alternative means of advertising to reduce such costs. For future research, it would be worthwhile to consider a real market situation in which retailers incorporate some hidden items, such as interest earned, the interest charged, and transportation costs. Additionally, it would be beneficial to explore how varying advertisement costs and advertising policies can impact each stage of a product's life cycle.

Author Contributions: Conceptualization, M.-C.C. and C.-T.C.; Methodology, M.-C.C., C.-T.C. and T.-P.H.; Software, T.-P.H.; Formal analysis, M.-C.C. and C.-T.C.; Writing—original draft, M.-C.C. and C.-T.C.; Writing—review & editing, M.-C.C., C.-T.C. and T.-P.H. All authors have read and agreed to the published version of the manuscript.

Funding: This research was supported by the Ministry of Science and Technology of the Republic of China under grants MOST 107-2410-H-570-001 and MOST108-2410-H-032-054.

Data Availability Statement: Not applicable.

Acknowledgments: The authors are grateful to the anonymous referees for their encouragement and constructive comments.

Conflicts of Interest: The authors declare no conflict of interest.

References

1. Donaldson, W.A. Inventory replenishment policy for a linear trend in demand: An analytical solution. *Oper. Res. Q.* **1977**, *28*, 663–670. [\[CrossRef\]](#)
2. Henery, R.J. Inventory replenishment policy for increasing demand. *J. Oper. Res. Soc.* **1979**, *30*, 611–617. [\[CrossRef\]](#)
3. Teng, J.T.; Yang, H.L. Deterministic economic order quantity models with partial backlogging when demand and cost are fluctuating with time. *J. Oper. Res. Soc.* **2004**, *55*, 495–503. [\[CrossRef\]](#)
4. Hill, R.M. Inventory models for increasing demand followed by level demand. *J. Oper. Res. Soc.* **1995**, *46*, 1250–1259. [\[CrossRef\]](#)
5. Ahmed, M.A.; Ai-Khamis, T.A.; Benkherouf, L. Inventory models with ramp type demand rate, partial backlogging and general deterioration rate. *Appl. Math. Comput.* **2013**, *219*, 4288–4307. [\[CrossRef\]](#)
6. Cheng, M.; Wang, G. A note on the inventory model for deteriorating items with trapezoidal type demand rate. *Comput. Ind. Eng.* **2009**, *56*, 1296–1300. [\[CrossRef\]](#)
7. Skouri, K.; Konstantaras, I. Order level inventory models for deteriorating seasonable/fashionable products with time dependent demand and shortages. *Math. Probl. Eng.* **2009**, *2009*, 679736. [\[CrossRef\]](#)
8. Lin, J.; Hung, K.C.; Julian, P. Technical note on inventory model with trapezoidal type demand. *Appl. Math. Model.* **2014**, *38*, 4941–4948. [\[CrossRef\]](#)
9. Wu, J.; Chang, C.T.; Teng, J.T.; Lai, K.K. Optimal order quantity and selling price over a product life cycle with deterioration rate linked to expiration date. *Int. J. Prod. Econ.* **2017**, *193*, 343–351. [\[CrossRef\]](#)
10. Pinkerton, I. Getting religion about returns. *Deal. Consum. Electron. Marketpl.* **1997**, *39*, 19–20.
11. Trager, I. Not so many happy returns. *Interact. Week* **2000**, *7*, 44–45.
12. Petersen, J.A.; Kumar, V. Are product returns a necessary evil? Antecedents and Consequences. *J. Mark.* **2009**, *73*, 35–51. [\[CrossRef\]](#)
13. Yalabik, B.; Petruzzzi, N.C.; Chhajed, D. An integrated product returns model with logistics and marketing coordination. *Eur. J. Oper. Res.* **2005**, *161*, 162–182. [\[CrossRef\]](#)
14. Chen, J.; Bell, P.C. Implementing market segmentation using full-refund and no-refund customer returns policies in a dual-channel supply chain structure. *Int. J. Prod. Econ.* **2012**, *136*, 56–66. [\[CrossRef\]](#)
15. Akçay, Y.; Boyaci, T.; Zhang, D. Selling with Money-Back Guarantees: The Impact on Prices, Quantities, and Retail Profitability. *Prod. Oper. Manag.* **2013**, *22*, 777–791. [\[CrossRef\]](#)

16. Hu, W.; Li, Y.; Govindan, K. The impact of consumer returns policies on consignment contracts with inventory control. *Eur. J. Oper. Res.* **2014**, *233*, 398–407. [\[CrossRef\]](#)
17. Priyan, S.; Uthayakumar, R. Two-echelon multi-product multi-constraint product returns inventory model with permissible delay in payments and variable lead time. *J. Manuf. Syst.* **2015**, *36*, 244–262. [\[CrossRef\]](#)
18. Ülkü, M.A.; Gürler, Ü. The impact of abusing return policies: A newsvendor model with opportunistic consumers. *Int. J. Prod. Econ.* **2018**, *203*, 124–133. [\[CrossRef\]](#)
19. Li, D.; Chen, J.; Liao, Y. Optimal decisions on prices, order quantities, and returns policies in a supply chain with two-period selling. *Eur. J. Oper. Res.* **2021**, *290*, 1063–1082. [\[CrossRef\]](#)
20. Kumari, M.; De, P.K. An EOQ model for deteriorating items analyzing retailer's optimal strategy under trade credit and return policy with nonlinear demand and resalable returns. *Int. J. Optim.* **2022**, *12*, 47–55. [\[CrossRef\]](#)
21. Cheng, M.C.; Ouyang, L.Y. Advance sales system with price-dependent demand and an appreciation period under trade credit. *Int. J. Inf. Manag. Sci.* **2014**, *25*, 251–262.
22. Cheng, M.-C.; Hsieh, T.-P.; Lee, H.-M.; Ouyang, L.-Y. Optimal ordering policies for deteriorating items with a return period and price-dependent demand under two-phase advance sales. *Int. J. Oper. Res.* **2020**, *20*, 585–604. [\[CrossRef\]](#)
23. Kotler, P.; Armstrong, G. *Principles of Marketing*; Pearson Education Limited: London, UK, 2012.
24. Urban, T.L. Deterministic inventory models incorporating marketing decisions. *Comput. Ind. Eng.* **1992**, *22*, 85–93. [\[CrossRef\]](#)
25. Lee, W.J.; Kim, D. Optimal and heuristic decision strategies for integrated product and marketing planning. *Decis. Sci.* **1993**, *24*, 1203–1214. [\[CrossRef\]](#)
26. Sadjadi, S.J.; Oroujee, M.; Aryanezhad, M.B. Optimal production and marketing planning. *Comput. Optim. Appl.* **2005**, *30*, 195–203. [\[CrossRef\]](#)
27. Wang, S.D.; Zhou, Y.W.; Wang, J.P. Supply chain coordination with two production modes and random demand depending on advertising expenditure and selling price. *Int. J. Syst. Sci.* **2010**, *41*, 1257–1272. [\[CrossRef\]](#)
28. Yadav, D.; Singh, S.R.; Kumari, R. Effect of demand boosting policy on optimal inventory policy for imperfect lot size with backorder in fuzzy environment. *Control Cybern.* **2012**, *41*, 191–212.
29. Rabbani, M.; Zia, N.P.; Rafiei, H. Coordinated replenishment and marketing policies for non-instantaneous stock deterioration problem. *Comput. Ind. Eng.* **2015**, *88*, 49–62. [\[CrossRef\]](#)
30. Manna, A.K.; Dey, J.K.; Mondal, S.K. Imperfect production inventory model with production rate dependent defective rate and advertisement dependent demand. *Comput. Ind. Eng.* **2017**, *104*, 9–22. [\[CrossRef\]](#)
31. Shaikh, A.A.; Cárdenas Barrón, L.E.; Bhunia, A.K.; Tiwari, S. An inventory model of a three parameter Weibull distributed deteriorating item with variable demand dependent on price and frequency of advertisement under trade credit. *RAIRO Oper. Res.* **2019**, *53*, 903–916. [\[CrossRef\]](#)
32. Khan, M.A.-A.; Shaikh, A.A.; Konstantaras, I.; Bhunia, A.K.; Cárdenas-Barrón, L.E. Inventory models for perishable items with advanced payment, linearly time-dependent holding cost and demand dependent on advertisement and selling price. *Int. J. Prod. Econ.* **2020**, *230*, 107804. [\[CrossRef\]](#)
33. San-José, L.A.; Sicilia, J.; Abdul-Jalbar, B. Optimal policy for an inventory system with demand dependent on price, time and frequency of advertisement. *Comput. Oper. Res.* **2021**, *128*, 105169. [\[CrossRef\]](#)
34. Mandal, A.; Pal, B. Optimizing profit for pricing and advertisement sensitive demand under unreliable production system. *Int. J. Syst. Sci. Oper.* **2021**, *8*, 99–118. [\[CrossRef\]](#)
35. Giri, B.C.; Dash, A. Optimal batch shipment policy for an imperfect production system under price-, advertisement- and green-sensitive demand. *J. Manag. Anal.* **2022**, *9*, 86–119. [\[CrossRef\]](#)

Disclaimer/Publisher's Note: The statements, opinions and data contained in all publications are solely those of the individual author(s) and contributor(s) and not of MDPI and/or the editor(s). MDPI and/or the editor(s) disclaim responsibility for any injury to people or property resulting from any ideas, methods, instructions or products referred to in the content.