



Supply planning optimization for linear production system with stochastic lead-times and quality control

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Abstract— This work consider the problem of supply planning optimization of imperfect production systems with stochastic lead times and quality control. A model for supply planning of the production system and three quality control policies are analyzed. Experimental results highlights the economic advantage of integrating quality control planning at the early phase of supply planning optimization of production systems.

Keywords—Imperfect production systems; Uncertainty; Linear chain production process; Quality control.

I. Introduction and related publications

The growing need for more flexible and responsive production systems leads to several new approaches aiming to endow the production systems of the future with a certain intelligence, agility and resilience. In this context, one of the most important challenges to deal with is uncertainty and its impact on the performances of the system. In fact, uncertainty causes to manufacturers several difficulties in planning production, in regulating inventory, and in meeting demand [1].

In the following, the serial supply chain structures with stochastic lead times are reviewed. We present the case of deterministic demand for one and multilevel serial supply chains. Then, we focus on some relevant works related to production planning integrating quality control.

The simulation study presented by Whybark and Williams [2] was among the first to suggest that safety lead times may perform better than safety stocks in a multi-level linear production-inventory system when the production and replenishment times are stochastic.

A few years later, Weeks [3] proposed a one-stage model where it is assumed a deterministic demand, a stochastic processing time and tardiness and holding costs. This problem has been proved to be equivalent to the standard “Newsboy” formulation.

Another analytical approach is proposed by Yano [4] to model serial production systems in order to optimize the planned lead times. In this model, which is based on a Lot-for-Lot (L4L) policy, it is assumed that procurement and processing times are stochastic and demand is deterministic. The objective function is composed of the holding and tardiness costs. A procedure is developed to generate solutions for two-stage systems. The same author extended this model to three-stage systems in [5]. In this

model, a replenishment cost is introduced each time a given component is available after a planned intermediary due date. The performance indicator to be minimized is the sum of the following costs: i) the expected holding and tardiness costs at each stage, and ii) the expected tardiness cost for the finished product. The author underlined the difficulty to model a system composed of more than three stages. It is only after many year that Elhafi [6] overcome this barrier by proposing an analytical model based on recursivity. The convexity of the objective function is proved and a dynamic programming based heuristic is proposed to find approximate solutions of good quality in an acceptable computer processing time.

In inventory control process, Kim et al. [7] consider one product, a known fixed demand and an Erlang-distributed lead times. The performance indicator is the expected total cost, which is composed of ordering, holding and tardiness costs. Based on the analytical formulation, the authors propose an approximate solution and compare it to optimal ones for the case where the prior information on the lead time distribution is available, and another case where no information exists. The results show the effectiveness and the robustness of this approach and how costs could be reduced under uncertain lead times. Two years later, Kim et al. [8] studied the same problem and analysed the bullwhip effect which is encountered when both demand and lead times are stochastic. An analytical method is used to prove that prior information on lead times variability could be very helpful to control the bullwhip effect in a multi-stage supply chain.

Apparently, the literature contains few publications considering linear chain supply planning with discrete lead times. The most of existing studies investigates the lead time uncertainty for one item or assembly systems. Readers can refer to [9], [10] and [11] where more complete literature reviews are provided concerning supply planning dealing with uncertainty on lead times. For other sources of uncertainty (demand, capacity, cost, etc.) readers can refer to the surveys of Aloulou et al. [12] and Díaz-Madroñero et al. [13].

In the last few decades, there has been a growing interest in integrating quality control in supply and production planning. In fact, the quality should be rigorously checked before delivering the finished product. In this way, the company ensures customer satisfaction and avoid excessive product returns. Nevertheless, it will be necessary to find a good compromise between the inspection costs and the cost of non-conformity. In this regard, Colledani and Tolio [14] investigate the interaction between

quality control system and production system and show how they impact each other's performances. They underline the necessity of jointly considering quality and logistics requirements while designing a production system.

The work of Rosenblatt and Lee [15] studies the relationship and the economic impact of the interaction between production planning, quality of products and the deterioration of the production process. The considered production process has two possible states in which it generates different rates of defectives. The objective is to determine the economic production quantities of this kind of production process while minimizing the total annual cost. The optimal production run time is determined for two deterioration models and multi-state deterioration.

Recently, Bettayeb et al. [16] proposed an integrated model for single item lot sizing and quality control planning. The objective is to minimize the total cost while ensuring a given level of outgoing quality. This cost is composed of holding, setup, production and inspection costs.

In the field of supply planning and quality control few papers analyses the uncertainty of lead time and quality control. In this work, we propose a model of supply planning optimization of imperfect production systems with stochastic lead times and quality control. The rest of the paper is organized as follows. The problem description and the general assumptions are described in the second section. The third one details the analytical models corresponding to three different quality control policies and the optimization approach. Section 4 presents the experimental results, compares the economical performances of each model and analysis the effects of some parameters. The paper ends by giving the conclusion and the perspectives of this work.

II. Problem description and assumptions

The problem under consideration in this paper concerns the optimization of supply planning of an imperfect linear production process where the lead times of the production steps are stochastic. The objective is to evaluate the effect of quality issues on the total cost of such a system and analyze the opportunity of integrating quality control activities in the optimization model of the system.

Before detailing the problem and the general assumptions, the following are the notations used in next sections. All costs are expressed in monetary unit and durations are expressed in time units.

Parameters

C_p	Production cost
c_i	Unit inspection cost
c_{nc}	Unit non-conformity cost
C_Q	Quality cost: the sum of inspection and non-conformity costs
SR	Sampling rate, i.e. proportion of products inspected
c_h	Unit inventory holding cost per time unit
c_t	Unit tardiness penalty per time unit
T	Due date

q_0	Proportion of non-conforming finished products generated by the whole production process.
m	Number of production operations (flowshop)
L_j	Lead time of operation j , $\forall j \in \{1, 2, \dots, m\}$ L_j is a discrete random variable which varies between l_j and u_j
t_i	Unit inspection duration
L	Total lead time, i.e. $L = \sum_{j=1}^m L_j$

Decision variables

X	Order release date
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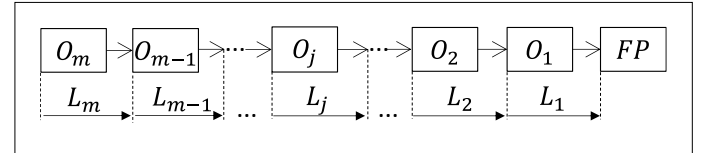
Functions

$E[\cdot]$	Expected value of $[\cdot]$
$F_m(\cdot)$	Cumulative distribution function of L

The considered production process is composed of m process operations (Fig. 1), which are consecutively executed to obtain the finished product FP . Each process step j has a random lead time duration L_j . The product starts to be processed at time X and the finished product is available at $T_{PF} = X + L$, after accumulating the m random lead times $L = L_1 + L_2 + \dots + L_m$. Note that T_{PF} is a random variable because of the randomness of the total lead time L .

Apart from the processing cost (omitted here), and depending on the due date T and the effective release date of the finished product, the generated production cost either equals the holding cost, if the product is finished before the due date, equals the tardiness penalty if it is available after the due date or equals zero if it is finished just in time.

Fig. 1. Linear production process



III. Analytical models and optimization approach

This section contains three models each of them corresponds to a certain policy regarding the quality control. In the first case, only the non-conformity penalty is assumed and no quality control is performed. In the second case, the quality control is performed at the end of production process but only the lead times are taken into account in the supply and production planning. In the last case, the quality control and the lead time variability are both taken into account while optimizing the supply and production planning.

A. Reference Policy (π_0): No quality control

It corresponds to the case where no quality control is performed during the production process. However, for each non-conforming product delivered to the customer, the producer pays a non-conformity penalty. The total cost is composed of inventory holding cost, tardiness cost and non-conformity cost:

$$TC^0(X, L) = C_P^0 + C_Q^0 \quad (1)$$

Where C_P^0 is equal to the sum of C_H the inventory holding cost and C_T the tardiness cost:

- If the finished products are available before T , it will be stored. The corresponding inventory holding cost C_H is equal to:

$$C_H = c_h \times (T - \min(T; X + L)) \quad (2)$$

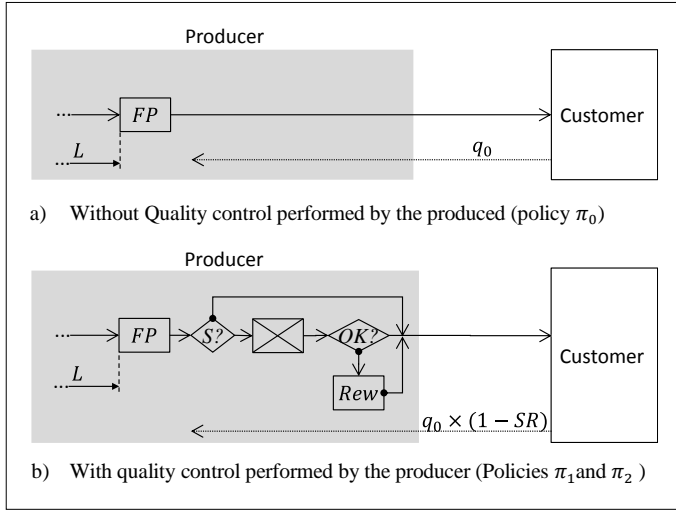
- There will be a stockout of the finished products if they are delivered after the due date T . Then tardiness cost C_T is equal to:

$$C_T = c_t \times (\max(T; X + L) - T) \quad (3)$$

And the non-conformity penalty C_Q^0 is equal to $q_0 \times c_{nc}$.

The total cost $TC^0(X, L)$ is a discrete random variable because of the randomness of the lead times L_j . Each of the latter varies between l_j and u_j , so the total cost has a finite number of possible values.

Fig. 2. Quality control policies



Proprety 1. The expression of the expected value of the total cost, denoted by $E[TC^0(X, L)]$, is given bellow:

$$E[TC^0(X, L)] = c_h \times \left(T - \sum_{0 \leq s \leq T-1} (1 - F_m(s - X)) \right) + c_t \times \sum_{s \geq T} (1 - F_m(s - X)) + q^0 \times c_{nc} \quad (4)$$

Proof. From expressions (1), (2) and (3), we have:

$$\begin{aligned} E[TC^0(X, L)] &= E[C_P^0(X, L)] + q_0 \times c_{nc} \\ &= c_h \times (T - E[\min(T; X + L)]) \\ &\quad + c_t \times (E[\max(T; X + L)] - T) + q_0 \times c_{nc} \end{aligned}$$

Let Γ a positive discrete random variable (with a finite range). Its expected value can be expressed as follows:

$$E[\Gamma] = \sum_{s \geq 0} (1 - \Pr[\Gamma \leq s]) = \sum_{s \geq 0} (1 - F_\Gamma(s)) \quad (5)$$

Knowing that $X + L$ does not depend on T , and using expression (5), we get:

$$\begin{aligned} E[\max(T; X + L)] &= \sum_{s \geq 0} (1 - \Pr[T \leq s] \times \Pr[X + L \leq s]) \\ &= T + \sum_{s \geq T} (1 - \Pr[X + L \leq s]) \\ &= T + \sum_{s \geq T} (1 - F_m(s - X)) \end{aligned} \quad (6)$$

Where F_m is the cumulative distribution function of the total lead time L , which defined as bellow:

- For $\alpha = 0$:

$$F_m(s - X) = \Pr[X + L_m \leq s]$$

- For $1 \leq \alpha \leq m - 1$; $m' = m - \alpha$:

$$F_{m'}(s - X) = \sum_{\substack{u_\alpha + v_\alpha = s \\ u_\alpha + v_\alpha \in \mathbb{Z}}} \Pr[L_{m'} = u_\alpha] \times F_{m'+1}(v_\alpha - X)$$

In the same way, we can easily deduce that:

$$E[\min(T; X + L)] = \sum_{0 \leq s \leq T-1} (1 - F_m(s - X)) \quad (7)$$

The expected total cost (4) is the objective function. It is a non-linear function which should be minimized. An exact method based on Newsboy formula is used to solve this problem in polynomial time.

Proposition 1. The method based on Newsboy formula presented in [17] gives the optimal order release date of the reference case X_0^* :

$$F_m(T - X_0^* - 1) \leq \frac{c_t}{c_h + c_t} \leq F_m(T - X_0^*) \quad (8)$$

Where $F_m(\cdot)$ is the cumulative distribution function of the total lead time L .

B. Policy 1 (π_1): Separated afterwards quality control planning

In this case, we suppose that the quality inspection is done just after the production process. A proportion of finished products is randomly sampled and inspected. The non-conforming products detected by inspection are repaired with no-extra cost.

The total cost is composed of production and quality costs, denoted by C_P^1 and C_Q^1 , respectively.

$$TC^1(X, L, SR) = C_P^1 + C_Q^1 \quad (9)$$

Note that, in this case, C_p^1 is the real production cost. It is equal to $C_p^{0*} - \Delta_h + \Delta_t$, where:

- C_p^{0*} is the optimal planned production cost associated to optimal order release date X_0^* ,
- Δ_h is the holding cost reduction corresponding to the time the product is being inspected :

$$\Delta_h = c_h \times \max(\min(I; T - X - L); 0) \quad (10)$$

Where $I = i_u \times SR$ corresponds to the total duration of inspection that corresponds to the quality control strategy in place.

- Δ_t is the tardiness cost increase due to the product inspection :

$$\Delta_t = c_t \times (I - \max(\min(I; T - X - L); 0)) \quad (11)$$

The quality cost C_Q^1 is composed of the cost of inspection and the penalty paid by the producer to the customer for non-conforming units:

$$C_Q^1 = q_0 \times (1 - SR) \times c_{nc} + SR \times c_i$$

As in the previous case, the total cost is also a random discrete random variable with a finite range. The expected cost is derived by the following property.

Proprety 2. The expression of the expected value of the total cost, denoted $E[TC^1(X_0^*, L, SR)]$, is given bellow:

$$E[TC^1(X_0^*, L, SR)] = E[TC^0(X_0^*, L)] - \left[(c_h + c_t) \times \sum_{s \leq I-1} F_m(T - s - 1 - X_0^*) - c_t \times I - SR \times (c_{nc} \times q_0 - c_i) \right]$$

Proof. From expression (4) and (7), $E[TC^0(X_0^*, L)]$ can be easily calculated. From expressions (10) and (11), we can easily deduce that:

$$E[\Delta_h - \Delta_t] = \sum_{s \leq I-1} F_m(T - s - 1 - X_0^*) \blacksquare$$

C. Policy 2 (π_2): Integrated production and quality control planning

In this case, we suppose that the quality control is planned in advance. It is made after the production process. A proportion of finished products is randomly sampled and inspected. The non-conforming products detected by inspection are repaired with no-extra cost.

The total cost is composed of production and quality costs, denoted by C_p^2 and C_Q^2 , respectively.

$$TC^2(X, L, SR) = C_p^2 + C_Q^2 \quad (12)$$

Where:

$$C_p^2 = [c_h \times (T - \min(T; X + L + I)) + c_t \times (\max(T; X + L + I) - T)]$$

$$C_Q^2 = q_0 \times (1 - SR) \times c_{nc} + SR \times c_i$$

The total cost $TC^2(X, L, SR)$ is a discrete random variable because of the randomness of lead times L_j . Each of the latter varies between l_j and u_j . Therefore, this total cost has a finite number of possible values. Thus, the mathematical expectation of cost can be determined.

Proprety 1. The expression of the expected value of the total cost, noted $E[TC^0(X, L)]$, is given bellow:

$$E[TC^2(X, L, SR)] = c_h \times \left(T - \sum_{0 \leq s \leq T-1} (1 - F_m(s - X - I)) \right) + c_t \times \sum_{s \geq T} (1 - F_m(s - X - I)) + q_0 \times (1 - SR) \times c_{nc} + SR \times c_i \quad (13)$$

Proof. From expressions (6), (7) and (10), we can deduce:

$$E[\max(T; X + L + I)] = T + \sum_{s \geq T} (1 - F_m(s - X - I))$$

And

$$E[\min(T; X + L)] = \sum_{0 \leq s \leq T-1} (1 - F_m(s - X - I)) \blacksquare$$

Proposition 2. The method based on Newsboy formula presented in [17] gives the optimal order release date of the reference case X_2^* :

$$F_m(T - X_2^* - 1 - I) \leq \frac{c_t}{c_h + c_t} \leq F_m(T - X_2^* - I) \quad (14)$$

Where $F_m(\cdot)$ is the cumulative distribution function of the total lead time L .

IV. Numerical example and discussion

The proposed cases described in Section 3 have been coded in C++. The experiments are carried on computer with 2.32 GHz Intel core i7 and 8 GB of RAM memory.

A. Experiments setting

The experiments are based on an example of a linear production system with three production steps, i.e. $m = 3$. The due date T is equal to 20 and the lead time of each process step is a discrete random variable uniformly distributed over the range of integers from 1 to 5, i.e. $\Pr(L_j = l) = 0.2 \forall j = 1, 2, 3 \forall l = 1, 2, \dots, 5$. It is assumed that the system is operating in a steady state, generating a constant proportion of non-conforming finished products $q_0 = 0.05$. The unit holding cost c_h is equal to the monetary unit and the other cost parameters experimented are given in Table 1.

Table 1. Experimented values for each parameter

Parameter	Range
L_j	[1, 5]
$\rho_t = c_t/c_h$	{0.1, 1, 10, 100, 1000}
$\rho_i = c_i/c_h$	{0.1, 1, 10, 100, 1000}
$\rho_{nc} = c_{nc}/c_h$	{0.1, 1, 10, 100, 1000}

SR	{0, 0.2, 0.4, ..., 1}
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B. Results and discussions

The preliminary results presented in this section aim to analyze the effect of integrating quality control in the early stage of supply and production planning. To do so, the following notion are introduced:

- The gap between policy π_1 and policy π_3

$$GAP1 (\%) = 100 \times \frac{ETC^1 - ETC^3}{ETC^1}$$

- The gap between policy π_2 and policy π_3

$$GAP2 (\%) = 100 \times \frac{ETC^2 - ETC^3}{ETC^2}$$

Fig. 3. Impact of the inspection cost on the Gap between ETC^3 and ETC^1

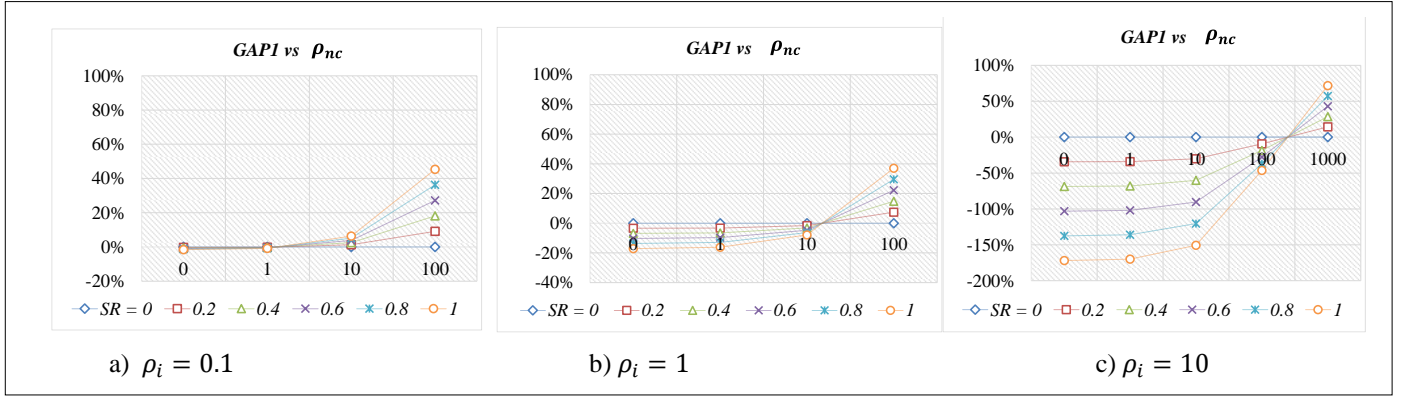


Fig. 4. Gap between ETC^2 and ETC^1 for some combinations of cost parameters

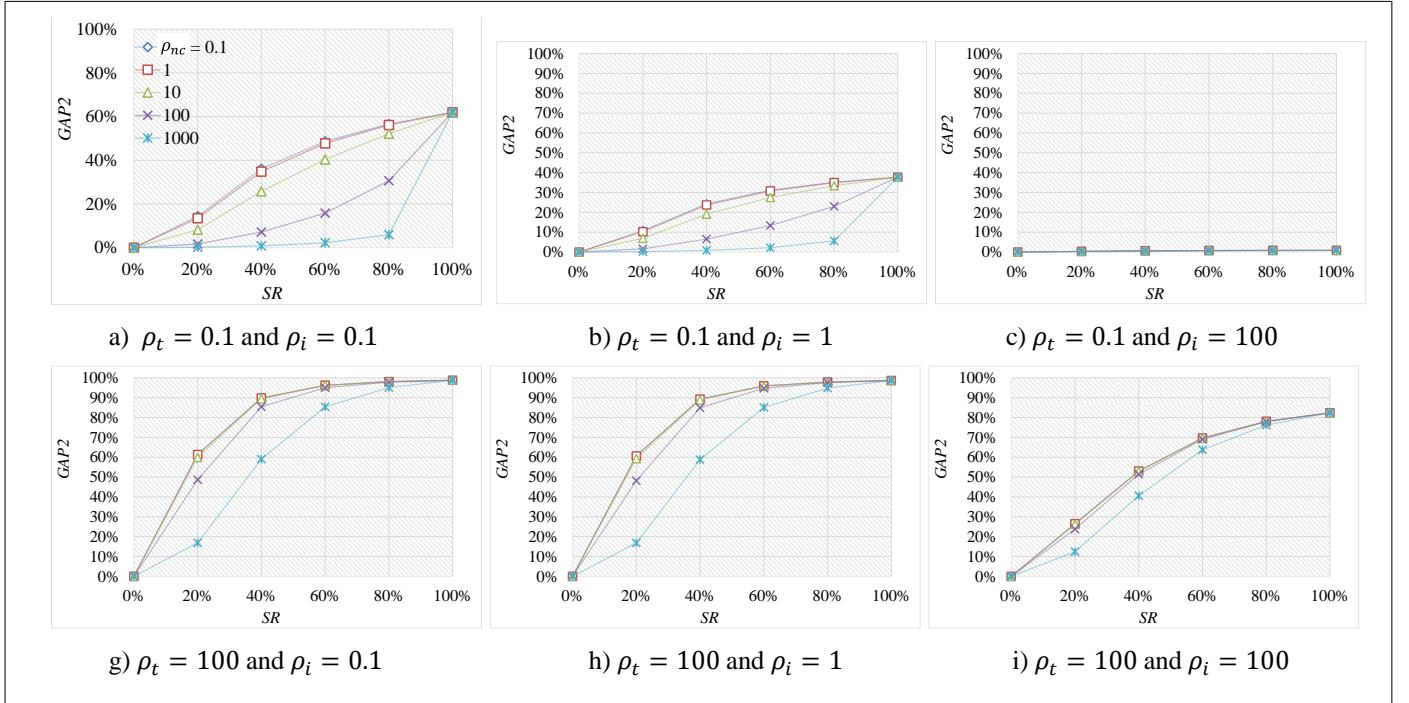


Figure 4 compares the performances of Policy 2 and Policy 3 by analyzing the gap between them. It proves that integrating quality control in supply and production planning reduces significantly the expected total cost whatever is the number of inspected products. We note that when inspection cost is much higher than the tardiness cost (Figure 4 c)), the policies are equivalent.

V. Conclusion

In this work, we are interested in supply and production planning under uncertainty of lead times and quality control. A model for supply planning of the production system and three quality control policies are analyzed. Experimental results highlights the economic advantage of integrating quality control planning at the early phase of supply planning optimization of production systems.

This exploratory study confirms the opportunity to extend this model by integrating maintenance activities and considering other kinds of production systems such as assembly systems where the finished products are assembled from several components.

Further, the second objective is to extend this model and different proposed technics to multi-period planning, in particular, calculate planned lead times when such a company deals with quality control, uncertainties of production and supply lead times.

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