## **Co-Evolution of Product Families and Assembly Systems**

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## Abstract

To cope with the intense global competition that is characterized by high product variety and short life cycles, manufacturers need to share manufacturing systems across products and product generations. Co-evolution of product families and assembly systems is proposed as a novel methodology for the joint design and reconfiguration of product families and assembly systems over several product generations. The co-evolution methodology capitalizes on the opportunities for design and assembly systems. As a result, co-evolution can lead to reduced product development costs and increased responsiveness to market changes.

#### Keywords:

Assembly System, Product Family, Co-evolution

#### **1 INTRODUCTION**

In today's market environment, manufacturers are faced with the challenge of cost-effectively supplying high variety within short product development times. Co-evolution of product families and assembly systems is introduced as a new product development methodology for the joint design and reconfiguration of assembly systems within and across product generations. The co-evolution methodology can enable manufacturers to remain competitive as it maximizes the reuse of product modules and reconfigurable systems [1] to ensure that manufacturing systems are effective for as many product generations as possible [2].

Various techniques for the reduction of product development time and costs have been proposed. Concurrent engineering is one such technique. Early concurrent engineering techniques focused on the integration of the detailed design phase with manufacturing [3]. More recent concurrent engineering techniques integrate the earlier conceptual product design phase with manufacturing system design in order to simultaneously determine product family and manufacturing system costs [4-6].

Figure 1 illustrates the main difference between coevolution and traditional concurrent engineering strategies. Co-evolution is a method for the incorporation of product variants and assembly system changes within a family generation, as well as between product generations, through continuously reconfiguring modular products and manufacturing systems. With concurrent engineering techniques, usually each generation of products is associated with a unique assembly system.





Figure 1: Comparison between concurrent engineering and co-evolution of product families and assembly system.

When products change from one generation to another, new assembly systems must be simultaneously designed. Some researchers have previously recognized a need for product evolution or co-evolution [7-11]. However, systematic methods for such co-evolution have not been proposed.

This paper proposes methodologies for the co-evolution of products and assembly systems. The remainder of this paper is organized as follows. The details of the co-evolution methodology are explained in Section 2. This will be followed by an example in Section 3 and conclusions in Section 4.

### 2 METHODOLOGY

There are two main phases of the co-evolution methodology: (1) the joint design of the product family and assembly system in the first generation and (2) the later co-evolution of the product family and system (Figure 1). The first generation design is assumed to be a 'clean sheet' design for both the product family and the assembly system. This initial design phase is important as the decisions made during this phase affect product family changes and assembly system reconfigurations in future generations. The co-evolution phase pursues economical product family changes and assembly system reconfigurations. As shown in Figure 1, the inputs to this phase are the existing product family, the required design changes, and the re-configuration constraints of the product family and assembly system. Using these inputs, new product modules are developed, the product family is modified and the assembly system is reconfigured. Since reconfiguring an assembly system can consume a lot of resources, a method for determining the effectiveness of the co-evolution plan is also proposed.

# 2.1 Phase I: design of the first generation product family and assembly system

The seller's welfare utility model has traditionally been used for the selection of product variants for a product family [12-13]. The selected product family is then used as an input to assembly system design formulations such as mixed product assembly line balancing (MPALB). MPALB is used to determine task assignments and the optimal number of workstations [14]. Since the product family influences the design of the assembly line and the cost of the assembly system influences the product variants selected, the seller's welfare models and MPALB models are combined to jointly determine the product family and assembly line design that leads to maximum profits. The maximum profit ( $\Pi$ ) is achieved by maximizing revenues and minimizing costs as follows:

maximize 
$$\prod_{z_j, x_{km}, s_{ij}, y_m} = \sum_{j \in PF} \rho_j V_j z_j - \varphi M$$
(1)

The decision variables are the product variants to be included in the product family  $(z_j)$ , the product variants selected by customers  $(s_{ij})$ , decision about whether a workstation is needed  $(y_m)$  and the assignment of tasks to workstations  $(x_{km})$  where *PF* is the set of products being considered for the product family and the subscripts *i*, *j*, *k* and *m* are customers, product variants, modules and workstations respectively. The upper limit of *i* is *I*, the number of customers interviewed; the upper limit of *j* is |PF|, the cardinality of set *PF*; the upper limit of both *k* and

*m* is *K*, the maximum number of modules.  $V_j = \frac{\psi}{I} \sum_{i=1}^{I} s_{ij}$  and

 $M = \sum_{m=1}^{K} y_m$ .  $\psi$  is the number of potential customers in the

market. In Eq. 1,  $\rho_j$  is the selling price of one unit of product variant *j* and  $\varphi$  is the combined fixed and variable costs for the assembly system.

The constraints for the problem are as follows:

$$\omega_{ij}z_j - \alpha_i \ge \gamma \left(s_{ij} - 1\right) \qquad \forall i = 1, \cdots, I; \ \forall j \in PF$$
(2)

$$\omega_{ij}z_j - \omega_{ir}z_r \ge \gamma \left(s_{ij} - 1\right) \quad \forall i = 1, \cdots, I; \ \forall r, j \in PF; \ r \neq j$$
(3)

$$\sum_{i \in PF} s_{ij} \le 1 \qquad \forall \ i = 1, \cdots, I$$
(4)

$$s_{ij} \le z_j$$
  $\forall i = 1, \cdots, I; j \in PF$  (5)

$$\sum_{i \in PF} z_j \le J' \tag{6}$$

$$\sum_{m=1}^{K} x_{km} = 1 \qquad \forall k = 1, \cdots, K$$
(7)

$$\sum_{k=1}^{K} \sum_{j \in PF} V_j \tau_{jk} x_{km} \le \lambda y_m \quad \forall \ m = 1, \cdots, K$$
(8)

$$x_{km} \leq \sum_{n=1}^{m} x_{qn} \quad \forall \ m = 1, \cdots, K; \ \forall \ k = 1, \cdots, K; \ \forall \ q \in P(k)$$
(9)

$$y_m \le y_n \quad \forall \ n, m = 1, \cdots, K; \ n \le m$$
(10)

$$z_j \in \{0,1\}; \ x_{km} \in \{0,1\}; \ s_{ij} \in \{0,1\}; \ y_m \in \{0,1\}$$
(11)

In Eqs. 2 and 3,  $\omega_{ii}$  is the utility that customer *i* receives from product variant j,  $\alpha_i$  is the utility that customer i receives from an existing product, and  $\gamma$  is a very large number. Eqs. 2 and 3 ensure that the customer utility for the selected product variant exceeds that of other product variants in the product family and of the existing product. Eq. 4 indicates that each customer selects at most one product variant and Eq. 5 prevents the customer from selecting a product variant not offered in the product family. Eq. 6 ensures that the total number of product variants selected for the product family does not exceed J', the limit set by marketing. Eq. 7 ensures that each assembly task is assigned to a workstation. Eq. 8 prevents the overloading of workstations and tasks from being assigned to unavailable workstations. In this constraint,  $\tau_{ik}$ is the time required to assemble module k of product variant *i* and  $\lambda$  is the planned life of the assembly system. Eq. 9 prevents precedence constraint violations. In this constraint P(k) is the set of predecessors of module k. Eq. 10 ensures that lower numbered stations are filled before higher numbered stations. Eq. 11 contains the feasibility constraints on the decision variables.

The profit maximization formulation outlined in Eqs. 1-11 is a non-linear integer program. These problems are known to be NP hard and therefore difficult to solve by exact optimization methods. Therefore, a genetic algorithm is used to solve this problem.

## 2.2 Phase II: joint reconfiguration of the product family and assembly system

As customer preferences change, new product functions need to be introduced and the assembly system updated. The reconfiguration decisions for the product family and assembly system are based on the new customer requirements and the constraints of the existing assembly system. They include determination of which of the modules with new functionality should be added, how they should be combined with existing product variants to form new product variants, whether any of the existing product variants should be dropped from the product family, and how much additional resources would be required to supply the new product family. Adopting a new product family ( $\Pi^n$ ) would only make sense if it allows the manufacturer to improve profits over the profits obtained from an existing product family ( $\Pi$ ). Therefore, the objective function of the second phase of co-evolution is to maximize profit increase ( $\Delta\Pi$ ) as follows:

maximize 
$$\Delta \Pi = \Pi^{n} - \Pi$$

$$where \Pi^{n} = \sum_{j \in PF} R_{j}V_{j} - \varphi \sum_{m=1}^{M} c_{m}y_{m}$$

$$(12)$$

In this formulation, the previous product family's profit is a parameter. The new decision variables introduced for the co-evolution problem are  $h_{jk^n\ell}$ ,  $c_m$  and  $x_{k^nm}$ .  $h_{jk^n\ell}$  indicates whether option  $\ell$  of module k is in product variant j. This leads to the design of new product variants or the redesign of previous product variants.  $x_{k^n m}$  is the station assignment variable for new modules and  $c_m$  is the number of parallel resources required at each workstation. Since the assembly system already exists,  $x_{km}$  is considered as a parameter and M is the upper bound of *m*.  $k^n$  represents new modules and has an upper limit  $K^n$ ; and  $\ell$  is a subscript for options of new modules options upper limit  $L^{k^n}$ . An intermediate and has variable,  $R_j = \rho_j + \sum_{k^n=1}^{K^n} \sum_{\ell=1}^{k^n} \xi_{k^n \ell} h_{jk^n \ell}$ , depends on the revenue

that could be previously obtained from the product variant  $\rho_j$  and the revenue that can be obtained from the new modules ( $\xi_{\mu_\ell}$ ).

All the constraint equations, except Eq. 8, remain the same with  $x_{km}$  being replaced by  $x_{k^nm}$  where appropriate. Eq. 8 is modified to Eq. 13 to account for changes in the workload assigned to each workstation as the product family evolves. This is required because although the module-workstation assignments remain the same, the volume of previous modules can change. Eq. 13 ensures that the time required to assemble new and old modules at a workstation does not exceed the total available time.

$$\sum_{k^n=1}^{K^n} \sum_{j \in PF} \sum_{\ell=1}^{j \in PF} V_j \left( h_{jk^n \ell} \theta_{k^n \ell} x_{k^n m} + \theta_{jm} \right) \le \lambda c_m y_m \qquad \forall m = 1, \cdots, M$$
(13)

where  $\theta_{k^n m}$  is the time required to assemble one unit option  $\ell$  of module  $k^n$  and  $\beta_{jm}$  is the current unit workload of all the modules previously assigned to workstation *m* for product variant *j*. The feasibility constraints for the new decision variables are as follows:

$$h_{jk^{n}\ell} \in \{0,1\}; \ x_{k^{n}m} \in \{0,1\}; \ c_{m} \ge 0 \tag{14}$$

As in Section 2.1, this formulation is a nonlinear integer program and is solved using a genetic algorithm.

#### 2.3 Effectiveness of the co-evolution plan

A reconfiguration process should be efficient in terms of (1) reuse, (2) time and (3) transition cost. The level of reuse is measured by the reuse ratio which is defined as the ratio of the number of system elements reused to the total number of system elements [2].

$$tr(\delta, \delta+1) = \left(\sum_{e=1}^{E} tr_e(\delta, \delta+1) \cdot \sigma_e\right) / \left(\sum_{e=1}^{E} \sigma_e\right)$$
(15)

 $tr(\delta, \delta+1)$  and  $tr_e(\delta, \delta+1)$  are the transition reuse ratio and the transition reuse ratio for system element *e* in making  $\delta$ and  $\delta+1$  changes to the product family, *E* is the total number of system elements.  $\varpi_e$  is the weight factor for system element *e*. The transition time for updating the assembly system ( $tt(\delta, \delta+1)$ ) is defined as the maximum time required for changing each of the system elements between two versions of the product family ( $tt_e(\delta, \delta+1)$ ).

$$tt(\delta,\delta+1) = \max\left[tt_e(\delta,\delta+1):\forall e\right]$$
(16)

This time should be as short as possible in order to minimize production losses during co-evolution. The efficiency of these transformation activities can be measured in terms of financial units as follows:

$$tc(\delta, \delta+1) = \sum_{e=1}^{E} tc_e(\delta, \delta+1)$$
(17)

where  $tc(\delta, \delta+1)$  is the total transition cost and  $tc_e(\delta, \delta+1)$  is the transition cost for system element *e*.

## 3 ILLUSTRATION

The example outlined in this section demonstrates the implementation of the co-evolution methodology.

## 3.1 Generation 1 product family and assembly system

The first generation (Gen 1) product family is derived from 11 modules (M1-M11). Modules M1-M8 occur in every product variant and do not have options. They are referred to as base modules. The other modules have options. The first option is not to have the module and the remaining options are variations of the module. The base product variant is defined as the product variant without any optional modules.

The precedence diagram for the assembly is shown in Figure 2. This precedence diagram is a modification of the one given in [14]. Each task in the precedence diagram is assumed to represent the assembly of an individual module. This assumption allows the serial number for an assembly task to be consistent with the serial number of the module assembled at the respective assembly task. The assembly times for the base product variant are also shown in Figure 2. Since the first node on the precedence diagram represents the introduction of the first module to the line, its associated task time is set to zero. The assembly times for the second option of M9, M10 and M11 are 2min, 9min and 2min, respectively. The assembly times for the third option of M10 and M11 are 3min and 5min respectively.



Figure 2: Precedence diagram and task times (min) for the Gen 1 base modules.

The selling prices of product variants were randomly generated from a normal distribution with mean \$30 and standard deviation \$10. The customers' utility value for an existing product was set to zero, indicating that there are no currently available products. The utility values for module-options were randomly generated from a normal

distribution with mean \$50 and standard deviation \$30. The selected product family is expected to have a life of two years on the market. The investment cost of each workstation is \$55K and workers are paid \$20/hr. *J*' is 18, the maximum number of potential product variants.

The results indicate that the maximum profit for Gen 1 is \$12.45M. The profit maximizing family has four variants. The options of M9-M11 selected are (222, 232, 223, 233). Each group of three numbers represents a variant and each of the three digits represents the selected module options. This family represents an 80% share of the market. Four workstations are required for assembly at a cycle time of 13.45min. The line layout is shown in Fig. 3 and the task assignments are [(1,2,4), (3,5), (6,7,8), (9,10,11)].



Figure 3: Gen 1 line layout.

## 3.2 Generation 2 product family and assembly system

Two new modules, M12 and M13, are introduced in the second generation (Gen 2). As for M9, these modules have two options. The assembly times for M12 and M13 are 8 min and 4 min respectively. The positions of these modules in the precedence diagram are shown in Figure 4. The distribution of customer utilities, module-option costs, and fixed and variable costs of the assembly system are the same as before.



Assembly tasks for Gen 2 modules with options

### Figure 4: Precedence diagram for Gen 2.

The optimal Gen 2 family has 11 variants and gives a profit increase of \$1.9M. For brevity, these product variants are not shown but can be obtained from the author upon request. Two of the variants had exactly the same modules as in Gen 1. The other product variants consist of combinations of previous product variants with the new modules. No product variants with just the new modules are obtained. Figure 5 shows the changes in the assembly line configuration for the Gen 2. M12 and M13 are assembled at workstations 2 and 4 respectively.



Gen 1 Station Modified Gen 1 Station New Gen 2 Station Figure 5: Gen 2 line layout.

Market changes may require the addition, deletion or change to modules in future generations. The addition/deletion of modules can result in the system layout being the same as for Gen 2 or an increase / decrease in the number of parallel resources. Since the assembly system is reconfigurable, the system can easily change as the product family evolves.

#### 3.3 Effectiveness of the co-evolution process

Assuming all the system elements are equally weighted, the reuse value of the system is 0.98. Upgrade to

workstations requires 1 week and addition of workstations require 2 weeks resulting in a change over time of 2 weeks. The total transition cost is the sum of labor and the investment cost for the two additional workstations. This cost is \$205K. Since the transition cost is much less than the profit increase, the co-evolution plan is feasible.

This example demonstrates how the co-evolution can be used to co-evolve a product family and assembly systems over generations.

### 4 CONCLUSIONS

This paper introduces a set of methodologies for the joint evolution of a product family and assembly system over product generations. Metrics to evaluate the effectiveness of the co-evolution plan were also introduced. The methodologies can potentially reduce product development costs and time by identifying changes to the assembly system as early as possible. An example was used to illustrate the implementation and benefits of the co-evolution approach.

### **5 REFERENCES**

- Koren, Y., Heisel, U., Jovan, F., Moriwaki, T., Pritschow, G., Ulsoy, G., Van Brussel, H., 1999, Reconfigurable Manufacturing Systems, Annals of CIRP, 48/2: 527-540.
- [2] Ko, J., Hu, J., Huang, T., 2005, Reusability Assessment for Manufacturing Systems, Annals of CIRP, 54/1: 113-116.
- [3] Boothroyd, G., Dewhurst, P., 1983, Product Design for Assembly, Wakefield, RI.
- [4] Abdi, M., Labib, A., 2004, Grouping and Selecting Products: The Design Key of Reconfigurable Manufacturing Systems (RMSs), Int. J. Prod. Res., 42/3: 521-546.
- [5] Jiao, J., Zhang, Y., 2005, Product Portfolio Planning with Customer-Engineering Interaction, IIE Transactions, 37/9: 801-814.
- [6] Michalek, J., Ceryan, O., Papalambros, P., Koren, Y., 2005, Manufacturing Investment Allocation in Product Line Decision Making, IDETC/CIE, Long Beach, California, 2A: 189-198.
- [7] Otto, K., Wood, K., 1998, Product Evolution: A Reverse Engineering and Redesign Methodology, Research in Engineering Design, 10: 226-243.
- [8] Kobayashi, H., 2005, Strategic Evolution of Eco-Products: a Product Life Cycle Planning Methodology, Research in Engineering Design, 16: 1-16.
- [9] Tolio, T., 2005, SPECIES Production System Evolution, CIRP working group presentation.
- [10] Kimura, F., Nielsen, J., 2005, A Design for Product Family Under Manufacturing Resource Constraints, Annals of CIRP, 54/1: 139-142.
- [11] Helfat, C., Ruth, R., 2000, Product Sequencing: Co-Evolution of Knowledge, Capabilities and Products, Strategic Management Journal, 21/10: 961-979.
- [12] Green, P., Krieger, A., 1985, Models and Heuristics for Product Line Selection, Marketing Science, 4/1: 1-19.
- [13] Li, H, Azarm, S., 2002, An Approach for Product Line Selection Under Uncertainty and Competition, Journal of Mechanical Design, 124: 385-392.
- [14] Becker, C., Scholl, A., 2006, Survey on Problems and Methods in Generalized Assembly Line Balancing, European Journal of Operational Research, 168: 694-715.