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Optimal Cost Design of Flow Lines with Reconfigurable Machines for Batch Production

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Abstract:

Modular reconfigurable machines offer the possibility to efficiently produce a family of different parts. This paper formalizes a cost optimization problem for flow lines equipped with reconfigurable machines which carry turrets, machining modules and single spindles. The proposed models take into account constraints related to: i) design of machining modules, turrets, and machines, ii) part locations, and iii) precedence relations among operations. The goal is to minimize equipment cost while reaching a given output and satisfying all the constraints. A Mixed Integer Program (MIP) is developed for the considered optimization problem. The approach is validated through an industrial case study and extensive numerical experiments.

Keywords: Reconfigurable manufacturing systems, Reconfigurable machine-tools, Manufacturing lines, Batch processing, Line design and balancing, Cost minimization, Optimization.

1. Introduction

Batch production is often used in low-volume diversified manufacturing context which is gaining more and more importance due to the major trend of product customization (Liao et al., 2017, Buer et al., 2018, Xu et al., 2018, Yin et al., 2018, Dolgui et al., 2019). A batch is a set of different parts that are manufactured together, it means that all the parts of the same batch follow the same manufacturing path, but each part receives its own operations. In reconfigurable manufacturing systems, operating modules of machines can be adapted for each batch. However, it is preferably to not use equipment reconfigurations for the parts of the same batch. Therefore, the batches are constituted in the way that the processing of the parts within the same batch will not require any equipment reconfiguration. The reconfiguration is only realized if needed between two different batches. In this paper, we consider a design problem for a reconfigurable manufacturing system for batch production. The design objective is to decide which modules will be used for each batch of products at which machine as well as which will be the general layout of the manufacturing system.

The considered manufacturing system may consist of several reconfigurable machines composed of different machining modules (spindle heads). A subset of machining modules is used for manufacturing each batch of parts. The decision on the modules to be used in the system and the number of machines has

to be made taking into account the parts to be produced and their quantity (size of each batch). Machining modules of a machine can be activated simultaneously or sequentially when applied to the same part. Simultaneous activation is realized for modules machined different accessible sides of the part. Sequential application is realized by the use of turrets which can carry several machining modules. In such a configuration, each part can be machined by at most 4 parallel modules corresponding to accessible sides of the machined part (two examples of such machines are shown in Fig. 1 and Fig. 2).

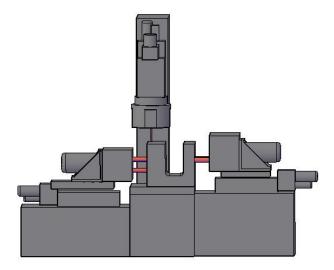


Fig. 1. A machine with one vertical turret and 2 horizontal modules

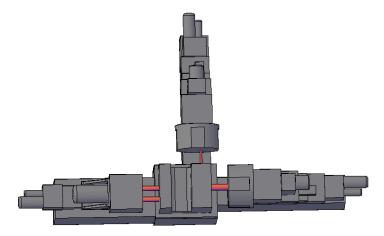


Fig. 2. A machine with two horizontal turrets and one horizontal module

The design problem considered in this paper concerns the choice of number of reconfigurable machines as well as the equipment to be installed on in terms of turrets and machining modules in order to machine a given number of batches. Since the location and machining axe and direction has to be decided for each machining module, the design solution defines also the orientations of each part on each machine, as well as cutting modes for each machining module.

The rest of the paper is organized as follows. Section 2 analyses the state of the art in the field. Section 3 provides a formal problem statement and Section 4 presents a mathematical model for it. A case study is described in Section 5. Finally, Section 6 concludes the study and suggests research perspectives.

2. Literature review

The research on reconfigurable manufacturing systems (RMS) introduced by Koren et al. 20 years ago (Koren et al. 1999) actually attracts a growing number of academic and industrial studies (Bortolini et al., 2018). This is due to the fact that RMS offer a novel effective manufacturing paradigm for nowadays production challenges characterized by an increasing variety of customized, high-quality products. A large number of publications dedicated to the research on RMS can be found in the literature. They provide multiple insights on the research perspectives in the field: on theoretical and practical challenges in the design of RMS (Rosio and Safsten, 2013), on measures of reconfigurability (Farid, 2014), on the use of artificial intelligence for the design of RMS (Renzi et al., 2014), on the adaptation of RMS for industrial assembly (Huettemann et al., 2015), on reconfigurable manufacturing on multiple levels (Andresen et al., 2015), on prerequisites and barriers for the development of RMS (Andresen et al., 2016), on the current design methods for RMS (Andersen et al., 2017). In particular, three recent surveys give a comprehensive view of challenges and research trends (Singh et al., 2017, Bortolini et al., 2018, Brahimi et al., 2019) where the interested reader will find a historical perspective of RMS development, detailed discussions on the topics attracting the major streams of research and future trends. Due to the availability of these complete bibliographical studies in the literature, we focus our literature analysis on the work that is the closest to our research question.

As for traditional manufacturing systems, the research on RMS concerns both the design and operation levels. Since in this study, a design problem is considered, the literature review is focused on the design level where a configuration of the system has to be defined and an equipment selection has to be made. These decisions however are usually closely related to the decisions on operation assignments to manufacturing modules and machines (Koren and Shpitalni, 2010).

One of the first studies on the design of RMS with the objective to optimize the investment cost can be found in Youssef and ElMaraghy (2006). This study has been consequently extended by considering a reconfiguration smoothness metric (Youssef and ElMaraghy, 2007) and machine availability (Youssef and ElMaraghy, 2008). Dou et al. (2007, 2010) studied reconfigurable flow lines with the objective to optimize the total investment cost for single- and multi-part flow lines, respectively. This research was further extended to a configuration generation problem where both the total cost and the total tardiness were minimized (Dou et al., 2016). Ashraf and Hasan (2018) considered the problem of optimal configuration selection of reconfigurable machine tools (RMT) for a reconfigurable serial flow line. These studies are close to our work, however, our design problem needs a more sophisticated mathematical model since the configuration of each machining module and the corresponding cutting

tools have to be defined according to the results of the optimization and not taken from a catalogue. As a consequence, more refined mathematical models are developed in the present study. In our previous work (Battaïa et al., 2017), we studied a close problem for rotary machining systems and here we extend the obtained mathematical model to a more general machining system layout.

The estimation of the total cost of RMS over planning horizon has been investigated in several studies. The optimization of the cost for scalability of RMS systems, in other words, their ability to be adapted to the scale of production volume was investigated by Son et al. (2001), Spicer et al. (2005), Wang and Koren (2012), and Koren et al., (2016). Saxena and Jain (2012) explored the structure of the cost of RMS including equipment investment, reconfiguration cost, operating cost, maintenance cost and residual marginal value over time. Xiaobo et al. (2000 a,b, 2001 a,b) used stochastic optimization to maximize the average expected profit for a designed RMS. Production planning of RMS with stochastic demands was investigated by Abbasi and Houshmand (2009, 2011). The problem of RMS configuration design is also addressed in Moghaddam et al. (2017), where the demand of a single product varies throughout its production life cycle, and the system configuration can change to satisfy the required demand with a minimum cost. The problem of selecting the optimal machining parameters for all operations of a batch processing line that minimize the total batch production cost while ensuring the required line throughput was studied by Dolgui et al. (2019).

In terms of multi-objective optimization, the cost is often regarded jointly with the total completion time, e.g. Bensmaine et al. (2013) and Haddou-Benderbal et al. (2017). Goyal et al. (2012, 2013) introduced a multi-objective model for the optimal machine assignment for a single part flow line with the objective to find a trade-off between the cost and the responsiveness of the RMS. The bi-objective model by Dou et al. (2016) was developed both for design and scheduling levels in order to minimize the investment and reconfiguration costs as well as total tardiness. The choice of single or multi-objective optimization depends on the criteria used in particular decision context. In our study, the industrial designers are principally interested by the criterion of the total investment cost while the production time is considered as a hard constraint.

The description of the optimization problem addressed in this paper is given in the next section. The aim of the study is not on the technology level but to facilitate the decision making process related to the preliminary design (combinatorial design) of RMS. In our study, we consider the stage of the design of RMS for a planning horizon with a known set of parts, therefore, the designed system includes all equipment required to process all known parts, it means that the reconfiguration "cost" is included in the investment cost, if the same equipment can be used for different parts it will be preferred in the design solution. The problem of reconfiguration for "unknown" parts is not considered in the present study. This optimization problem for batch production design is novel and not been addressed in the literature yet.

Moreover, the model we develop does not only help to solve this specific problem, but also offers new ideas for modelling and solving other similar optimization issues.

3. Problem statement

The considered design problem consists in defining the number of reconfigurable machines m and their configurations in order to produce d_0 types of parts. The parts are grouped in \aleph batches with required output O_{υ} (where υ is the number of batch , $\upsilon=1,2,...,\aleph$). Batches and parts in each batch are processed sequentially. Parts of υ^{th} batch are loaded in sequence $\pi_{\upsilon}=(\pi_{\upsilon 1}, \pi_{\upsilon 2}, ..., \pi_{\upsilon \mu_{\upsilon}})$ where $\pi_{\upsilon i}\in\{1,2,...,d_0\}$, $j=1,2,...,\mu_{\upsilon}$ where μ_{υ} is the number of parts processed in a sequence in batch $\upsilon=O_{\upsilon}$ Using sequences π_{υ} we can define, in one-to-one manner, function $\pi_{\upsilon}(i,k)$, $i=1,...,O_{\upsilon}\mu_{\upsilon}+m-1$, of part number on k^{th} machine after i movements of the conveyer, $\pi_{\upsilon}(i,k) \in \{0,1,2,...,d_0\}$ ($\pi_{\upsilon}(i,k)=0$ means that no part is processed on k^{th} machine).

Figures 3 to 5 in Section 5 show an example of batch of three different parts.

Let us recall the introduced indexes for further presentation of parameters and decision variables:

- Index for reconfigurable machines k = 1,...,m, where m is the number of machines.
- Index for parts: $d \in D = \{1, 2, ..., d_0\}$, where d_0 is the number of types of parts.
- Index for batches: $v=1,2,...,\aleph$, where \aleph is the number of batches to be produced.
- Index for the positions of parts in sequence of each batch: π_{υ} =($\pi_{\upsilon 1}$, $\pi_{\upsilon 2}$, ..., $\pi_{\upsilon \mu_{\upsilon}}$) where $\pi_{\upsilon j}$ ∈ {1,2,..., d_{υ} }, j=1,2,..., μ_{υ} and μ_{υ} is the number of parts processed in a sequence in batch υ = O_{υ} .

Let \mathbb{N}^d be the set of machining operations needed for machining of elements of d^{th} part, $d \in D = \{1, 2, ..., d_0\}$ located on its n_d sides and N_s^d , $s = 1, 2, ..., n_d$, is a subset of operations for machining of elements of s^{th} side of the part d. The part d can be located at a machine in different orientations $\mathbf{H}(d)$ but elements of no more than one side can be processed by a machining module. Machining modules can be of 4 types: type j=1 (from above), j=2 (from the left), j=3 (from behind), j=4 (from the right) and no more than 3 types of machining modules can be installed at each machine. $\mathbf{H}(d)$ can be represented by a matrix of dimensions $r_d \times n_d$ where $h_{rs}(d)$ is equal j, j=0,1,2,3,4, if the operations of N_s^d can be processed by machining modules of type j (j=0 means that processing is not possible).

Let **N** be the set of all operations required for completing all considered parts, $\mathbf{N} = \bigcup_{d=1}^{d_0} \mathbf{N}^d$. Any operation

 $p \in \mathbb{N}$ is characterized by the following parameters: the length $\lambda(p)$ of the working stroke for operation p, the range $[\gamma_1(p), \gamma_2(p)]$ of feasible values of feed rate defining cutting speed, and the set J(p) of types of possible machining modules for execution of operation p.

The considered optimization problem is to determine:

- the number *m* of machines;
- the orientation H_{dk} of part $d \in D$ at machine k, k = 1,...,m;
- the number b_{kj} of machining modules of type j, j=1,...,4, that are installed at machine k, k=1,...,m;
- the set of machining operations N_{dkjl} from \mathbf{N}^d , assigned to machining module l of type j at machine k, $d \in D$, k = 1,...,m, j=1,...,4, $l=1,...,b_{kj}$;
- the feed rate $\Gamma_{dkjl} \in [\max{\{\gamma_1(u)|u \in N_{dkjl}\}}, \min{\{\gamma_2(u)|u \in N_{dkjl}\}}]$ for execution of operations from N_{dkjl} , $d \in D$, k=1,...,m, j=1,...,4, $l=1,...,b_{kj}$.

Let $P_{dkj} = (P_{dkjl} = (N_{dkjl}, \Gamma_{dkjl}) | l = 1,...,b_{kj}), P_{dk} = (P_{dkj} | j = 1,...,4), P_{d} = ((P_{dk}, H_{dk}) | k = 1, ...,m), P = (P_{d} | d \in D).$ P is a design decision.

The execution time $t^b(P_{dkil})$ operations from with feed rate $t^b(P_{dkjl}) = L(N_{dkjl})/\Gamma_{dkjl} + \tau^a$, $\Gamma_{dkjl} \in [max\{\gamma_1(p)|p \in N_{dkjl}\}, min\{\gamma_2(p)|p \in N_{dkjl}\}]$ is equal to where $L(N_{dkjl}) = max\{\lambda(p)| p \in N_{dkjl}\}$, and τ^a is an additional time for advance and disengagement of tools.

We assume that if the turret of type j is installed at machine k then the execution time of operations from N_{dkjl} is equal to $t^h(P_{dkj}) = \tau^g b_{kj} + \sum_{l=1}^{b_{kj}} t^b(P_{dkjl}), j=1,...,4$, where τ^g is an additional time for one rotation of a turret. If the spindle head is installed, then $t^h(P_{dkj}) = t^b(P_{dkjl}), j=1,...,4$. If all N_{dkjl} are empty then $t^h(P_{dkj}) = 0$. If $b_{kj} = 1$, then $t^h(P_{dkj}) = t^b(P_{dkjl})$.

The execution time $t^p(P_{dk})$ is defined as $t^p(P_{dk}) = \tau^r + max\{t^h(P_{dkj})|j=1,...,4\}$, where τ^r is an additional time for part transportation and its location at a machine. Then the time $T_v(P)$ of execution of all corresponding operations for batch v is defined as $T_v(P) = \sum_{i=1}^{O_v \mu_v + m - 1} max\{t^p(P_{\pi_v(i,k)k})|k=1,...,m\}$ and

the time T(P) for machining all the batches is equal to $\sum_{v=1}^{N} T_v(P)$.

We assume that the given productivity is provided, if the total time T(P) does not exceed the available time T_0 .

Let C_{sm} , C_{tb} , C_{mm} , C_{sb} , and C_{st} be the relative costs for one machine, one turret, one machining module of a turret, one spindle box, and one workhead with a single spindle, respectively. Then the cost $C(b_{kj})$ for performing set of operations N_{kj} by associated b_{kj} machining modules can be assessed as follows:

$$C(b_{kj}) = \begin{cases} 0 \text{ if } b_{kj} = 0, \\ C_{st} \text{ if } b_{kj} = 1 \text{ and } |N_{kj1}| = 1, \\ C_{sb} \text{ if } b_{kj} = 1 \text{ and } |N_{kj1}| > 1, \\ C_{tb} + C_{mm}b_{kj} \text{ if } b_{kj} > 1. \end{cases}$$

The machine cost Q(P) is calculated as the total cost of all equipment used:

$$Q(P) = C_{sm}m + \sum_{k=1}^{m} \sum_{j=1}^{4} C(b_{kj}) + \sum_{k=1}^{m-1} \sum_{d=1}^{d_0} C(H_{dk}, H_{dk+1}),$$

where $C(H_{dk}, H_{dk+1})$ is the cost incurred by changing orientation of part d between stations k and k+1 and $C(H_{dk}, H_{dk+1})=0$ if $H_{dk}=H_{dk+1}$ and $C(H_{dk}, H_{dk+1})=C_{ro}$, otherwise (C_{ro} is the relative cost for reorientation).

The design decision P should satisfy the following constraints:

- precedence constraints which define time dependencies among operations;
- inclusion constraints which require to execute some pairs of operations from N at the same machine, by the same turret, by the same machining module or even by the same spindle;
- exclusion constraints which require to not assign some pairs of operations from N to the same machine, to the same turret, or to the same machining module;
- constraints on the maximal number m_0 of machines and on the maximal number b_0 of machining modules in a turret;
- constraints on feasible orientations of the part for processing each operation;
- the impossibility to perform operations from N_s^d at one machine by machining modules of different types;
- productivity constraints to provide the required output (a time limit for total completion time).

Precedence constraints are modelled by a directed graph $G^{OR}=(\mathbf{N},D^{OR})$: if an arc $(p,q)\in D^{OR}$ then operation p has to be executed before operation q. It should be noted that if such operations p and q belong to different sides of the part, then they cannot be executed at the same machine without violating the precedence constraint.

Inclusion constraints are represented by undirected graphs G^{SP} =(\mathbf{N} , E^{SP}), G^{ST} =(\mathbf{N} , E^{ST}), G^{SM} =(\mathbf{N} , E^{SM}) and G^{SS} =(\mathbf{N} , E^{SS}): if there is an edge (p,q) \in E^{SS} (or (p,q) \in E^{SM} , or (p,q) \in E^{ST} , or (p,q) \in E^{SP}), then operations p and q must be executed by the same spindle (the same machining module, turret, at the same machine, respectively).

Exclusion constraints are represented by undirected graphs $G^{DM}=(\mathbf{N}, E^{DM})$, $G^{DT}=(\mathbf{N}, E^{DT})$, and $G^{DP}=(\mathbf{N}, E^{DP})$. If there is an edge $(p,q) \in E^{DM}$ (or $(p,q) \in E^{DT}$), or $(p,q) \in E^{DP}$), then operations p and q cannot be executed by the same machining module (turret, at the same machine, respectively).

4. MIP approach

In order to solve the considered problem, we developed a Mixed Integer Programming (MIP) model where the following notations are used:

- X_{pkjl} decision variable which is equal to 1 if operation p is assigned to the lth machining module of spindle head or turret of type j at station $k, p \in \mathbb{N}, k=1,...,m_0, j=1,...,4, l=1,...,b_0$
- h_{rdk} decision variable which is equal to 1 if orientation r is chosen for part d at station k, $r=1,...,r_d$, $d \in D$, $k=1,...,m_0$
- Y_{kj}^{ds} auxiliary variable which is equal to 1 if at least one machining module of type j is installed at station k for execution of operations from N_s^d , $d \in D$, $s=1,...,n_d$, $k=1,...,m_0$, j=1,...,4
- Y_{kj}^d auxiliary variable which is equal to 1 if at least one machining module of type j is installed at station k for execution of operations from N^d , $d \in D$, $k=1,...,m_0$, j=1,...,4
- Y_{kjl} auxiliary variable which is equal to 1 if the lth machining module of type j is installed at station $k, k=1,...,m_0, j=1,...,4, l=1,...,b_0$
- Y_{kjs} auxiliary variable which is equal to 1 if the workhead with a single spindle of type j is installed at station k, $k=1,...,m_0$, j=1,...,4
- Y_{kjb} auxiliary variable which is equal to 1 if the spindle box of type j is installed at station k, $k=1,...,m_0, j=1,...,4$
- Z_k auxiliary variable which is equal to 1 if station k is used, $k=1,...,m_0$
- ρ_{dkk+1} auxiliary variable which is equal to 1 if part d is reoriented between stations k and k+1, $d \in D$, $k=1,...,m_0-1$
- F_{dkjl} execution time for part d by l-th machining module of spindle head or turret type j at the k-th machine
- F_{dk} the execution time for part d at k-th machine
- F^{vi} total remaining processing time when processing of part π_{vi} of v-th batch is finished
- $F_{\rm s}^{\upsilon i}$ total time of first *i* machines after the *i*-th turn of the conveyer for processing υ -th batch
- $F_f^{\upsilon i}$ total time of last i machines after the $O_{\upsilon}\mu_{\upsilon}+m_0-i$ -th turn of the conveyer for processing υ -th batch
- η_υ number of cyclically repeating part sequences when all machines are occupied $η_{\upsilon} = (O_{\upsilon} \mu_{\upsilon} m_0 + 1)/\mu_{\upsilon}$

 θ_{v} additional number of part sequences when all machines are occupied $\theta_{v}=mod(O_{v}\mu_{v}-m_{0}+1,\mu_{v})$

minimal time necessary for the execution of operations p and q by the same machining module, $t_{pq} = \max(\lambda(p), \lambda(q)) / \min(\gamma_2(p), \gamma_2(q)) + \tau^a$

M an upper bound on the maximal number of operations to be potentially assigned to one machining module $M = \sum_{d=1}^{d_0} \max\{N_s^d \mid s=1,...,n_d\}$

The objective function can be expressed as follows:

Minimize
$$C_{sm} \sum_{k=1}^{m_0} Z_k + \sum_{k=1}^{m_0} \sum_{j=1}^{4} (C_{sb} Y_{kjb} + (C_{tb} + 2C_{mm}) Y_{kj2} + C_{st} Y_{kjs}) + C_{mm} \sum_{k=1}^{m_0} \sum_{j=1}^{4} \sum_{l=3}^{b_0} Y_{kjl} + C_{ro} \sum_{k=1}^{m_0-1} \sum_{d \in D} \rho_{dkk+1}$$
 (1)

The following equations determine decision variables X_{pkjl} as well as auxiliary variables Y_{kj}^{ds} , Y_{kjl} , and Z_k .

$$\sum_{k=1}^{m_0} \sum_{j=1}^{4} \sum_{l=1}^{n_0} X_{pkjl} = 1; p \in \mathbb{N}$$
 (2)

$$X_{pkjl}=0; j=1,...,4; j \notin J(p)$$
 (3)

$$\sum_{k'=1}^{k-1} \sum_{j=1}^{4} \sum_{l=1}^{b_0} X_{pk'j'l'} + \sum_{l'=1}^{l-1} X_{pkjl'} \ge X_{qkjl}; (p,q) \in D^{OR}; k=1,...,m_0; j=1,...,4; l=1,...,b_0$$
(4)

$$X_{pkj} = X_{qkj}; \{p,q\} \in E^{SM} \cup E^{SS}; k=1,...,m_0; j=1,...,4; l=1,...,b_0$$
 (5)

$$\sum_{l=1}^{b_0} X_{pkjl} = \sum_{l=1}^{b_0} X_{qkjl} ; \{p,q\} \in E^{ST}; k=1,...,m_0; j=1,...,4$$
(6)

$$\sum_{j=1}^{4} \sum_{l=1}^{b_0} X_{pkjl} = \sum_{j=1}^{4} \sum_{l=1}^{b_0} X_{qkjl} ; \{p,q\} \in E^{SP}; k=1,...,m_0$$
(7)

$$X_{pkjl}+X_{qkjl}\leq 1; \{p,q\}\in E^{DM}; k=1,...,m_0; j=1,...,4; l=1,...,b_0$$
 (8)

$$\sum_{l=1}^{b_0} X_{pkjl} + \sum_{l=1}^{b_0} X_{qkjl} + Y_{kj2} \le 2; \{p,q\} \in E^{DT}; k=1,...,m_0; j=1,...,4$$
(9)

$$\sum_{j=1}^{4} \sum_{l=1}^{b_0} X_{pkjl} + \sum_{j=1}^{4} \sum_{l=1}^{b_0} X_{qkjl} \le 1; \{p,q\} \in E^{DP}; k=1,...,m_0$$
(10)

$$\sum_{l=1}^{b_0} \sum_{p \in N_s^d} X_{pkjl} \le |N_s^d| Y_{kj}^{ds}; Y_{kj}^{ds} \le \sum_{l=1}^{b_0} \sum_{p \in N_s^d} X_{pkjl}; d \in D; s = 1, ..., n_d; k = 1, ..., m_0, j = 1, ..., 4$$
(11)

$$\sum_{j=1}^{4} Y_{kj}^{ds} \le 1, \ d \in D; \ s=1,...,n_d; \ k=1,...,m_0$$
(12)

$$\sum_{s=1}^{n_d} Y_{kj}^{ds} \le 1, \ d \in D; \ k=1,...,m_0; \ j=1,...,4$$
(13)

$$Y_{kjl} \le \sum_{p \in \mathbb{N}} X_{pkjl} ; \sum_{p \in \mathbb{N}} X_{pkjl} \le |\mathbb{N}| Y_{kjl}; k=1,...,m_0; j=1,...,4; l=1,...,b_0$$
 (14)

$$Y_{kjl} \ge Y_{kjl+1}; k=1,...,m_0; j=1,...,4; l=1,...,b_0-1$$
 (15)

$$\sum_{j=1}^{4} Y_{kj1} \le 3; \quad \sum_{j=1}^{4} Y_{kj1} \le 3Z_k; \ Z_k \le \sum_{j=1}^{4} Y_{kj1}; \ k=1,...,m_0$$
 (16)

$$\sum_{p \in \mathbb{N}} X_{pkj1} + Y_{kj2} + MY_{kjs} \le M + 1, \ k=1,...,m_0; j=1,...,4$$
(17)

$$Y_{kib} + Y_{ki2} + Y_{kis} = Y_{ki1}, \ k=1,...,m_0; j=1,...,4$$
 (18)

$$Z_k \leq Z_{k-1}; k=2,...,m_0$$
 (19)

Constraints (2) – (3) guarantee an assignment of each operation to exactly one machining module of compatible type. Precedence constraints, inclusion constraints for spindles, machining modules, turrets and machines as well as exclusion constraints for machining modules, turrets and machines are provided by (4) – (10) respectively. Constraints (11) – (12) prevent from an assignment of operations located at the same side of each part to machining modules of different types of the same machine. Constraints (13) – (19) define the configuration of machines, workheads with single spindle, spindle boxes, turrets, and corresponding machining modules. A workhead with single spindle is installed (Y_{kjs} =1) if and only if $\sum_{p \in \mathbb{N}} X_{pkj1} = 1$, Y_{kj1} =1, Y_{kjb} =0 and Y_{kj2} =0. If Y_{kj2} =1 (a turret is installed), then Y_{kjb} =0 and Y_{kjs} =0.

Decision variables h_{rdk} and auxiliary variables ρ_{dkk+1} are defined by the following constraints:

$$h_{rdk} + Y_{kj}^{ds} \le 1; r=1,...,r_d; d \in D; k=1,...,m_0; s=1,...,n_d; j=1,...,4; j \ne h_{rs}(d)$$
 (20)

$$\sum_{r=1}^{r_d} h_{rdk} = 1 \; ; \; d \in D; \; k=1,...,m_0$$
 (21)

$$r_{d}\rho_{dkk+1} \ge \sum_{r=1}^{r_{d}} rh_{rdk} - \sum_{r=1}^{r_{d}} rh_{rdk+1}; \ r_{d}\rho_{dkk+1} \ge \sum_{r=1}^{r_{d}} rh_{rdk+1} - \sum_{r=1}^{r_{d}} rh_{rdk}; \ d \in D; \ k=1,...,m_{0}-1$$
 (22)

Constraints (20) – (21) defines the orientation of each part at each machine. Variables ρ_{dkk+1} are equal to 0 if $h_{rdk} = h_{rdk+1}$ for all $r=1,...,r_d$ due to (22) and minimization of (1).

The required productivity is provided if the following constraints are satisfied:

$$F_{dkjl} \ge t_{pp} X_{pkjl}; p \in \mathbb{N}^d; d \in D; k=1,...,m_0; j=1,...4; l=1,...,b_0$$
 (23)

$$F_{dkjl} \ge t_{pq}(X_{pkjl} + X_{qkjl} - 1); p, q \in \mathbb{N}^d; d \in D; k=1,...,m_0; j=1,...,4; l=1,...,b_0$$
(24)

$$F_{dk} \ge \sum_{l=1}^{b_0} F_{dkjl} + \tau^g (2Y_{kj2} + \sum_{l=3}^{b_0} Y_{kjl}) + b_0 \tau^g (Y_{kj}^d - 1); \ d \in D; \ k = 1, \dots, m_0; \ j = 1, \dots, 4$$
 (25)

$$F^{\cup i} \ge F_k^{\pi_{\upsilon}(m_0 + i - 1, k)} + \tau^r; \ \upsilon = 1, \dots, \aleph; \ i = 1, \dots, \mu_{\upsilon}; \ k = 1, \dots, m_0$$
(26)

$$F_s^{\upsilon i} \ge F_k^{\pi_{\upsilon}(i,k)} + \tau^r; \ \upsilon = 1, \dots, \aleph; \ i = 1, \dots, m_0 - 1; \ k = 1, \dots, i$$
 (27)

$$F_f^{\upsilon i} \ge F_k^{\pi_{\upsilon}(O_{\upsilon}\mu_{\upsilon} + m_0 - i, k)} + \tau^r; \, \upsilon = 1, \dots, \aleph; \, i = 2, \dots, m_0; \, k = i, \dots, m_0$$
(28)

$$\sum_{\nu=1}^{\aleph} (\eta_{\nu} \sum_{i=1}^{\mu_{\nu}} F^{\nu i} + \sum_{i=1}^{\theta_{\nu}} F^{\nu i} + \sum_{i=1}^{m_{0}-1} F_{s}^{\nu i} + \sum_{i=2}^{m_{0}} F_{f}^{\nu i}) \le T_{0}$$
(29)

Constraints (21) – (22) calculate the execution time of the l-th machining module of type j at the k-th machine for processing part d. If a turret of type j with b_{kj} machining modules is installed at the k-th machine then: $F_{dk} \ge \sum_{l=1}^{b_{kj}} F_{kjl}^d + b_0 \tau^g$ due to (25) if at least one operation from N^d is executed by the turret and $F_{dk} = 0$, otherwise. If a spindle head of type j is installed at the k-th machine then $F_{dk} \ge F_{kj1}^d$.

Variables domains are defined as follows:

$$X_{pkjl} \in \{0,1\}; p \in \mathbb{N}; k=1,...,m_0; j=1,...,4; l=1,...,b_0$$
 (30)

$$Y_{kj}^{ds} \in \{0,1\}; k=1,...,m_0; d \in D; s=1,...,n_d; j=1,...,4$$
 (31)

$$Y_{kj}^d \in \{0,1\}; k=1,...,m_0; d \in D; j=1,...,4$$
 (32)

$$Y_{kjl}^d \in \{0,1\}; k=1,...,m_0; j=1,...,4; l=1,...,b_0; d \in D$$
 (33)

$$Y_{kjl} \in \{0,1\}; k=1,...,m_0; j=1,...,4; l=1,...,b_0$$
 (34)

$$Z_k \in \{0,1\}; k=1,...,m_0$$
 (35)

$$h_{rdk} \in \{0,1\}; r=1,...,r_d; d\in D; k=1,...,m_0$$
 (36)

$$\rho_{dkk+1} \in \{0,1\}; d \in D; k=1,...,m_0-1$$
(37)

$$F_{kil}^{d} \in [0, \bar{t}_{k}^{d} - \tau^{r}]; k=1, \dots, m_{0}; j=1, \dots, 4; l=1, \dots, b_{0}; d \in D$$
(38)

$$F_k^d \in [0, \bar{t}_k^d - \tau^r]; k=1, ..., m_0; d \in D$$
(39)

$$F^{\cup i} \in [t^{\cup i}, \bar{t}^{\cup i}]; \ \upsilon = 1, ..., \aleph; \ i = 1, ..., \mu_{\upsilon}$$
 (40)

$$F_{S}^{\upsilon i} \in [\max\{\underline{t}^{\pi_{\upsilon}(i,k)} \mid k = 1,...,i\}, \max\{\bar{t}_{k}^{\pi_{\upsilon}(i,k)} \mid k = 1,...,i\}]; \upsilon = 1,...,\aleph; i = 1,...,m_{0}-1$$

$$(41)$$

$$F_{f}^{\upsilon i} \in [\max\{\underline{t}^{\pi_{\upsilon}(O_{\upsilon}\mu_{\upsilon}-i,k)} \mid k=i,...,m_{0}\}, \max\{\overline{t}_{k}^{\pi_{\upsilon}(O_{\upsilon}\mu_{\upsilon}-i,k)} \mid k=i,...,m_{0}\}]; \upsilon=1,...,\aleph; i=2,...,m_{0}$$
(42)

where

$$\underline{t}^d = \min\{\lambda(p)/\gamma_2(p) + \tau^a + \tau^r | p \in \mathbf{N}^d\},\,$$

$$\underline{t}^{\cup i} = \max\{\underline{t}^{\pi_{\cup}(i,k)} \mid k = 1,...,m_0\},\,$$

$$\bar{t}^{\upsilon i} = (T_0 - O_{\upsilon'} \sum_{\upsilon'=1,\upsilon'\neq\upsilon}^{\aleph} \sum_{i'=1}^{\mu_{\upsilon'}} \underline{t}^{\upsilon'i'}) / O_{\upsilon} - \sum_{i'=1,i'\neq i}^{\mu_{\upsilon}} \underline{t}^{\upsilon i'}$$

and

$$\bar{t}_k^d = \max\{\bar{t}^{\cup i} \mid \upsilon = 1,..., \aleph, i = 1,..., O_{\upsilon}\mu_{\upsilon} + m_0 - 1, \pi_{\upsilon}(i,k) = d\}$$

5. A case study

In order to validate the developed model, it was used to design a flow line for machining 3 different parts (Fig. 3 to Fig. 5) in 2 batches with O_1 =34 and O_2 =68. The parameters of machining operations are given in Table 1. The operations to be executed for the first part are located on 4 different sides and those for the second and third parts are located on only 3 sides. The loading sequences of parts are {1,2} and {3}. Other parameters are: $\tau^a = \tau^g = \tau^r = 0.1$ min, C_{sm} =20, C_{sh} =4, C_{tb} =5, C_{mm} =2, C_{ro} =0.5. The available time T_0 is 384 min. The possible orientations of the parts are defined as follows:

$$\mathbf{H}(1) = \begin{pmatrix} -1 & 1 & 4 & -1 \\ 4 & 1 & -1 & -1 \\ 4 & -1 & -1 & 1 \\ -1 & -1 & 4 & 1 \\ 1 & 2 & -1 & -1 \end{pmatrix}, \ \mathbf{H}(2) = \begin{pmatrix} -1 & 1 & 4 \\ 4 & 1 & -1 \\ 1 & 2 & -1 \end{pmatrix}, \ \mathbf{H}(3) = \begin{pmatrix} -1 & 1 & 2 \\ -1 & 1 & 4 \\ 2 & -1 & -1 \\ 4 & -1 & -1 \end{pmatrix}.$$

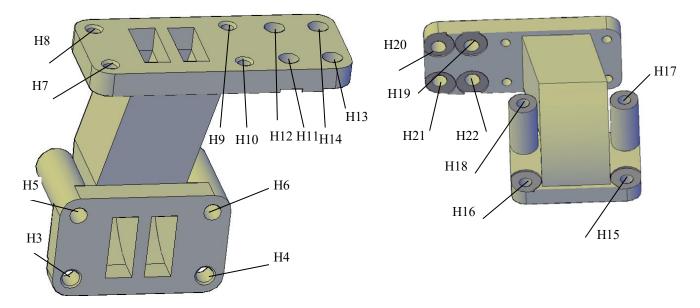


Fig.3. The first part to be machined

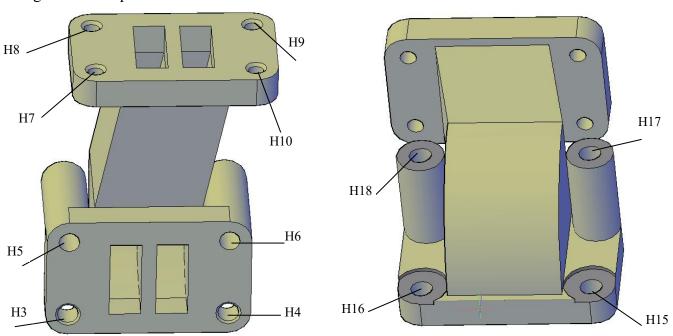


Fig.4. The second part to be machined

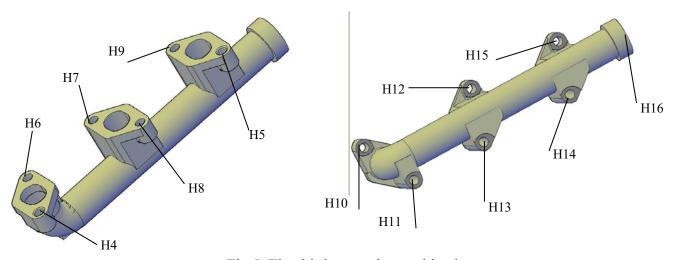


Fig.5. The third part to be machined

Table 1. Operations and their parameters

p	Hole	Part	Side	$\lambda(p)$ mm	mm/	$\gamma_2(p)$, mm/	J(p)	p	Hole	Part	Side	$\lambda(p)$, mm	mm/	$\gamma_2(p)$, mm/	J(p)
					min	min							min	min	
1	H3	1	1	34	37.7	63.4	{1,4}	36	H6	2	1	75	29.7	105.7	{1,4}
2	Н3	1	1	22	27.8	249.5	{1,4}	37	H7	2	2	24	24.6	83.6	{1,2}
3	H4	1	1	34	37.7	63.4	{1,4}	38	H7	2	2	9	28.3	106.3	{1,2}
4	H4	1	1	22	27.8	249.5	{1,4}	39	H8	2	2	24	24.6	83.6	{1,2}
5	H5	1	1	79	22.8	81.3	{1,4}	40	H8	2	2	9	28.3	106.3	{1,2}
6	H5	1	1	75	29.7	105.7	{1,4}	41	H9	2	2	24	24.6	83.6	{1,2}
7	H6	1	1	79	22.8	81.3	{1,4}	42	H9	2	2	9	28.3	106.3	{1,2}
8	H6	1	1	75	29.7	105.7	{1,4}	43	H10	2	2	24	24.6	83.6	{1,2}
9	H7	1	2	24	24.6	83.6	{1,2}	44	H10	2	2	9	28.3	106.3	{1,2}
10	H7	1	2	9	28.3	106.3	{1,2}	45	H15	2	3	2	18.8	62.7	{4 }
11	H8	1	2	24	24.6	83.6	{1,2}	46	H16	2	3	2	18.8	62.7	{4 }
12	H8	1	2	9	28.3	106.3	{1,2}	47	H17	2	3	2	18.8	62.7	{4 }
13	Н9	1	2	24	24.6	83.6	{1,2}	48	H18	2	3	2	18.8	62.7	{4}
14	Н9	1	2	9	28.3	106.3	{1,2}	49	H4	3	1	53	39.2	62.9	{2,4}
15	H10	1	2	24	24.6	83.6	{1,2}	50	H4	3	1	34	27.2	248	{2,4}
16	H10	1	2	9	28.3	106.3	{1,2}	51	H5	3	1	53	39.2	62.9	{2,4}
17	H11	1	2	25	22	82.2	{1,2}	52	H5	3	1	34	27.2	248	{2,4}
18	H12	1	2	25	22	82.2	{1,2}	53	H6	3	1	100	22.8	81.3	{2,4}
19	H13	1	2	25	22	82.2	{1,2}	54	H6	3	1	98	29.7	105.7	{2,4}
20	H14	1	2	25	22	82.2	{1,2}	55	H7	3	1	100	22.8	81.3	{2,4}
21	H15	1	3	2	18.8	62.7	{4}	56	H7	3	1	98	29.7	105.7	{2,4}
22	H16	1	3	2	18.8	62.7	{4 }	57	H8	3	1	45	22.8	81.3	{2,4}
23	H17	1	3	2	18.8	62.7	{4 }	58	H8	3	1	43	29.7	105.7	{2,4}
24	H18	1	3	2	18.8	62.7	{4 }	59	H9	3	1	100	22.8	81.3	{2,4}
25	H19	1	4	2	18.8	62.7	{1}	60	H9	3	1	98	29.7	105.7	{2,4}
26	H20	1	4	2	18.8	62.7	{1}	61	H10	3	2	3	15.5	51.6	{1}
27	H21	1	4	2	18.8	62.7	{1}	62	H11	3	2	3	15.5	51.6	{1}
28	H22	1	4	2	18.8	62.7	{1}	63	H12	3	2	3	15.5	51.6	{1}
29	H3	2	1	34	37.7	63.4	{1,4}	64	H13	3	2	3	15.5	51.6	{1}
30	H3	2	1	22	27.8	249.5	{1,4}	65	H14	3	2	3	15.5	51.6	{1}
31	H4	2	1	34	37.7	63.4	{1,4}	66	H15	3	2	3	15.5	51.6	{1}
32	H4	2	1	22	27.8	249.5	{1,4}	67	H16	3	2	30	43.7	74.1	{2,4}
33	H5	2	1	79	22.8	81.3	{1,4}	68	H16	3	5	24	31.9	197.1	{2,4}
34	H5	2	1	75	29.7	105.7	{1,4}	69	H16	3	5	24	26.9	161.6	{2,4}
35	H6	2	1	79	22.8	81.3	{1,4}	70	H16	3	5	18	26.7	160.2	{2,4}

Precedence constraints, exclusion constraints for machining modules, turrets and machines are presented in Tables 2, 3, 4, and 5, respectively. Inclusion constraints for machines, turrets and machining modules are given in Tables 6, 7 and 8. Operations to be executed by the same spindle are presented in Table 9.

Table 2. Precedence constraints

Operation	Predecessors	Operation	Predecessors
2	1 3 29 31	40	9 11 13 15 37 39 41 43
4	1 3 29 31	42	9 11 13 15 37 39 41 43
6	5 7 33 35	44	9 11 13 15 37 39 41 43
8	5 7 33 35	50	49 51
10	9 11 13 15 37 39 41 43	52	49 51
12	9 11 13 15 37 39 41 43	54	53 55 59
14	9 11 13 15 37 39 41 43	56	53 55 59
16	9 11 13 15 37 39 41 43	58	57
30	1 3 29 31	60	53 55 59
32	1 3 29 31	68	67
34	5 7 33 35	69	68
36	5 7 33 35	70	69
38	9 11 13 15 37 39 41 43		

Table 3. Incompatibility of operations in machining modules

Operation	Incompatible operations
9/10/11/12/13/14/15/16	12345678
17	1 2 3 4 5 6 7 8 13 14
18	1 2 3 4 5 6 7 8 17
19	1 2 3 4 5 6 7 8 18
20	1 2 3 4 5 6 7 8 15 16 17 19
21/22	1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20
23	1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21
24	1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 22
25	1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24
26	1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25
27	1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 26
28	1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 26 27
37/38/39/40/41/42/43/44	29 30 31 32 33 34 35 36
45/46	29 30 31 32 33 34 35 36 37 38 39 40 41 42 43 44
47	29 30 31 32 33 34 35 36 37 38 39 40 41 42 43 44 45
48	29 30 31 32 33 34 35 36 37 38 39 40 41 42 43 44 46
61/62/63/64/66	49 50 51 52 53 54 55 56 57 58 59 60
67/68/69/70	49 50 51 52 53 54 55 56 57 58 59 60 61 62 63 64 65 66

Table 4. Incompatibility of operations in turrets

Operation	Incompatible operations
9/10/11/12/13/14/15/16/17/18/19/20	1 2 3 4 5 6 7 8
21/22/23/24	1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20
25/26/27/28	1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24
37/38/39/40/41/42/43/44	29 30 31 32 33 34 35 36
45/46/47/48	29 30 31 32 33 34 35 36 37 38 39 40 41 42 43 44
61/62/63/64/65/66	49 50 51 52 53 54 55 56 57 58 59 60
67/68/69/70	49 50 51 52 53 54 55 56 57 58 59 60 61 62 63 64 65 66

Table 5. Incompatibility of operations in machines

Operation	Incompatible operations
1/2/3/4/5/6/7/8	21 22 23 24
17/18/19/20	25 26 27 28
49/50/51/52/53/54/55/56/57/58/59/60	61 62 63 64 65 66

Table 6. Operations to be assigned to the same machine

Operation	Operations to be executed at the same machine	Operation	Operations to be executed at the same machine
49	51 53 55 57 59	50	52 54 56 58 60

Table 7. Operations to be assigned to the same turret

Operation	Operations to be executed by the same	Operation	Operations to be executed by the same		
	turret	Operation	turret		
25	26 27 28	67	68 69 70		

Table 8. Operations to be assigned to the same machining module

Operation	Operations to be executed by the same machining module	Operation	Operations to be executed by the same machining module
1	3	33	35
5	7	34	36
6	8	37	39 41 43
9	11 13 15	49	51
17	19	53	55 59
18	20	54	56 60
29	31		

Table 9. Operations to be executed by the same spindle

Operation	Operations to be executed by the same tool	Operation	Operations to be executed by the same tool
1	29	11	39
2	30	12	40
3	31	13	41
4	32	14	42
5	33	15	43
6	34	16	44
7	35	21	45
8	36	22	46
9	37	23	47
10	38	24	48

Solver CPLEX 12.2 was used to solve the corresponding problems (1) - (42) for b_0 =4 and m_0 =2, 3, 4, 5 on PC Intel Pentium (2.40 Ghz, 4 Gb RAM). The obtained results are presented in Table 10.

Table 10. Results of optimization

m_0	Number of variables	Number of constraints	Solution time, s
2	1639	49700	8.18
3	2458	74553	20.13
4	3277	99406	40.51
5	4096	124259	56.82

In all obtained solutions the flow line consists of 2 machines and 2 turrets should be installed at each machine but the assignment of operations to machining modules is quite different. The obtained optimal solution for m_0 =5 and its characteristics are presented in Tables 11 and 12. The total cost Q(P) is equal to $2C_{sm}+0C_{sh}+4C_{tb}+12C_{mm}+3C_{ro}5=85.5$. In this case p_1 =32, p_2 =64, $\eta_1 = \eta_2 = 0$. The total time T(P) is equal to 32(2.46+2.46)+1.03+2.46+2.46+2.46+2.46+64(3.02)+3.02+3.02+3.02+3.02+1.96=375.63 min.

Table 11. An optimal solution

Set N_{dkjl}	Operations of N_{dkjl}	$L(N_{dkjl})$	γdkjl	$t^b(P_{dkjl})$
$\overline{N_{1111}}$	25	2	62.7	0.13
N_{1112}	27	2	62.7	0.13
N_{1113}	28	2	62.7	0.13
N_{1114}	26	2	62.7	0.13
N_{1141}	21 24	2	62.7	0.13
N_{2141}	45 48	2	62.7	0.13
N_{3141}	49 51 53 55 59 57	100	62.9	1.69
N_{1142}	22 23	2	62.7	0.13
N_{2142}	46 47	2	62.7	0.13
N_{3142}	50 52 54 56 60 58	98	105.7	1.03
N_{1211}	1 3 5 7	79	63.4	1.35
N_{2211}	29 31 33 35	79	63.4	1.35
N_{1212}	2 4 6 8	75	105.7	0.81
N_{2212}	30 32 34 36	75	105.7	0.81
N_{3212}	61 62 63 64 65 66	3	51.6	0.16
N_{3221}	67	40	74.1	0.77
N_{1222}	9 11 13 15	24	83.6	0.39
N_{2222}	37 39 41 43	24	83.6	0.39
N_{3222}	68	24	197.1	0.22
N_{1223}	12 14 18 20	25	82.2	0.4
N_{2223}	40 42	9	106.3	0.18
N_{3223}	69	24	161.6	0.25
N_{1224}	10 16 17 19	25	82.2	0.4
N_{2224}	38 44	9	106.3	0.18
N_{3224}	70			

Table 12. Characteristics of the solution

Machine k	$t^p(P_{1k})$	$t^p(P_{2k})$	$t^p(P_{3k})$	H_{1k}	H_{2k}	H_{3k}
1	1.03	0.56	3.02	-1,-1,4,1	-1,1,4	4,-1,-1
2	2.46	2.46	1.96	1,2,-1,-1	1,2,-1	-1,1,2
3	0	0	0	-	-	-
4	0	0	0	-	-	-
5	0	0	0	-	-	_

The proposed approach was also tested on 20 industrial examples with $|\aleph|=2$. The proprieties of these instances are given in Table 9 where OS^{OR} is the order strength of the precedence constraints (the ratio between the number of arcs in the closure of G^{OR} and the number of arcs in the complete digraph with the same number of vertices. Densities of other constraints (D^{DM} , D^{DT} , D^{DP} , D^{SS} , D^{SM} , D^{ST} , and D^{SP}) are defined in the same way.

Table 9. The proprieties of the industrial examples

Problem	$ \mathbf{N} $	OS^{OR}	D^{DB}	D^{DT}	D^{DP}	D^{SS}	D^{SB}	D^{ST}	D^{SP}	D	$\sum_{v=1}^{8} \mu_v$
1	66	0.050	0.502	0.228	0.030	0.002	0.025	0.003	0.026	2	4
2	58	0.015	0.356	0.000	0.039	0.000	0.033	0.004	0.031	2	4
3	58	0.060	0.521	0.257	0.015	0.002	0.030	0.000	0.016	2	4
4	50	0.020	0.358	0.000	0.020	0.000	0.041	0.000	0.017	2	4
5	92	0.037	0.712	0.205	0.017	0.013	0.022	0.000	0.039	3	4
6	82	0.013	0.640	0.000	0.022	0.014	0.028	0.000	0.048	3	4
7	100	0.034	0.515	0.224	0.018	0.006	0.018	0.001	0.022	3	4
8	88	0.010	0.376	0.000	0.023	0.004	0.023	0.002	0.027	3	4
9	92	0.038	0.550	0.242	0.011	0.007	0.020	0.000	0.019	3	4
10	80	0.013	0.408	0.000	0.015	0.005	0.026	0.000	0.023	3	4
11	90	0.037	0.517	0.247	0.040	0.006	0.018	0.001	0.015	3	4
12	78	0.011	0.360	0.000	0.053	0.004	0.024	0.002	0.018	3	4
13	80	0.041	0.524	0.233	0.043	0.006	0.022	0.002	0.010	3	4
14	70	0.012	0.380	0.000	0.056	0.005	0.028	0.002	0.014	3	4
15	124	0.027	0.493	0.239	0.024	0.007	0.014	0.001	0.015	4	4
16	108	0.008	0.335	0.000	0.032	0.005	0.018	0.001	0.019	4	4
17	114	0.029	0.482	0.229	0.029	0.007	0.016	0.001	0.018	4	4
18	100	0.009	0.329	0.000	0.037	0.006	0.020	0.001	0.022	4	4
19	148	0.023	0.493	0.226	0.019	0.008	0.012	0.001	0.019	5	6
20	130	0.007	0.346	0.000	0.025	0.007	0.016	0.001	0.024	5	6

The experiments were carried out for two variants of each problem that differed in matrices $\mathbf{H}(d)$ and sets J(p). The results are presented in Tables 10 and 11 where NO is the number of rows of a matrix \mathbf{H} of possible orientations of all parts, TC is the calculation time in seconds, NV and NC are the numbers of variables and constraints in MIP formulations, respectively. \mathbf{H} is built on the base of matrices $\mathbf{H}(d)$ and takes also into account constraints D^{SS} and D^{SM} .

Table 10. The results for the first variant

Problem	NO	TC	Q(P)	m	NV	NC
1	20	10.9	158.5	4	3513	122608
2	16	16.4	119.5	3	3188	72585
3	10	14.4	143.5	4	3183	100419
4	8	13.6	103	3	2858	56796
5	125	10.6	136	4	3112	293550
6	48	14.2	96.5	3	2772	206452
7	100	16.9	159.5	4	4307	266213
8	64	69.2	120.5	3	3977	153665
9	50	24.8	144.5	4	3977	240794
10	32	24.3	104	3	3647	137846
11	40	20.2	159	4	4212	220392
12	32	44.2	119.5	3	3887	119849
13	40	40.1	148.5	4	3732	178124
14	24	8.3	91	2	3402	103066
15	200	29.2	160	4	5006	385227
16	128	64.6	120.5	3	4676	200959
17	200	25.2	160	4	4606	322529
18	96	43.1	120	3	4271	173516
19	1000	47.8	161	4	5422	536299
20	384	76.6	121	3	5082	288161

Table 11. The results for the second variant

Problem	NO	TC	Q(P)	m	NV	NC
1	45	48.8	147.5	4	3538	122093
2	45	71.7	87	2	3218	72155
3	27	95	138	4	3208	99944
4	27	110.8	79.5	2	2888	56406
5	405	128.6	133.5	4	3152	293090
6	405	45.8	75.5	2	2832	206092
7	405	95.6	148	4	4352	265598
8	405	350.2	87.5	2	4032	153140
9	243	202.3	138.5	4	4022	240219
10	243	534	80	2	3702	137361
11	135	78.6	148	4	4242	219782
12	135	75.9	87	2	3922	119324
13	75	157.7	148	4	3742	177714
14	75	33.1	91	2	3422	102746
15	1215	167.1	148.5	4	5056	384517
16	1215	398	87.5	2	4736	200339
17	675	62.6	152.5	4	4636	321939
18	675	104.5	91.5	2	4316	173021
19	6075	645.4	153	4	5472	535609
20	6075	1152.2	92	2	5152	287571

The obtained results show that the tested industrial problems could be solved to optimality in reasonable time. However, the complexity of the problem and the time required to solve it strongly depends on the density of inclusion and exclusion constraints which in fine define the number of possible solutions.

Conclusions

In this paper, we developed a new mathematical model for the cost optimization problem in the design of machining flow lines composed of reconfigurable machines and used for batch production. This new model includes the definition of precedence, exclusion and inclusion constraints among operations and different types of equipment, such as machining modules, turrets, spindle boxes and workheads with single spindles. The configuration of the line is decided by optimizing the assignment of operations to the pieces of equipment while satisfying technical and technological constraints as well as the required final completion time.

The developed model is general and can be applied to different reconfigurable manufacturing systems (RMS) used in machining industry. The proposed model has been successfully validated on industrial case studies. The obtained results show that it can be used in practice for designing RMS for variable batches of parts required in low volume. In particular, the solution time for finding optimal solutions for industrial problems is acceptable. Therefore, the use of the developed model may help designers to evaluate quickly different design configurations and the investment costs required. Thus, the total cost and design time can be substantially reduced. The future research will include the integration of environmental factors into the considered problem. Despite of the fact that the investment cost is still

important for decision makers in industry, solutions can be also compared on the basis of their environmental indicators such as energy consumption, waste generation, options available for the end of life treatment. However, an important effort will be needed to properly model these environmental indicators.

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