

Document downloaded from:

<http://hdl.handle.net/10251/99055>

This paper must be cited as:

Araujo, FF.; Costa, AM.; Miralles Insa, CJ. (2012). Two extensions for the ALWABP: Parallel stations and collaborative approach. *International Journal of Production Economics*. 140(1):483-495. doi:10.1016/j.ijpe.2012.06.032



The final publication is available at

<https://doi.org/10.1016/j.ijpe.2012.06.032>

Copyright Elsevier

Additional Information

# **Two extensions for the assembly line worker assignment and balancing problem: parallel stations and collaborative approach.**

**Felipe F. B. Araújo**

*Instituto de Ciências Matemáticas e de Computação  
Universidade de São Paulo  
felipe@icmc.usp.br*

**Alysson M. Costa**

*Instituto de Ciências Matemáticas e de Computação  
Universidade de São Paulo  
alysson@icmc.usp.br*

**Cristóbal Miralles**

*ROGLE - Dpto. Organización de Empresas  
Universitat Politècnica de València  
cmiralles@omp.upv.es*

## **Abstract**

*In this article, we introduce two new variants of the assembly line worker assignment and balancing problem that allow parallelization of and collaboration between heterogeneous workers. These new line balancing approaches introduce an additional level of flexibility in the assignment and planning process, which may be particularly useful in practical situations where the aim is to progressively integrate slow or limited workers in conventional assembly lines. We present linear models and heuristic procedures for these two new problems. Computational results show the efficiency of the proposed approaches and the efficacy of the studied layouts in different situations.*

**Keywords:** assembly line balancing, disabled workers.

## **1. Introduction**

According to the International Labour Organization (ILO), people with disabilities represent an estimated 10 per cent of the world's population, or some 650 million people worldwide. Approximately 470 million are of working age. Many have demonstrated that with the right opportunities along with adaptations and support, if needed, they can make a major contribution at all levels of the economy and society. Yet, they are often excluded and marginalized, and are particularly vulnerable in times of crisis (ILO, 2009).

Like everyone else, people with disabilities have a right to a full life in every sense, and one of those fundamental rights is the right to work. Employment is, in itself, the key to integration and autonomy. In fact, the allocation of roles that employment gives becomes the most effective form of rehabilitation (Wolfensberger, 1983). Work provides structure, friends outside of one's domestic circle, something to look forward to each day in short and long terms, and valuable socialization as well as the need to develop new skills and to enhance existing skills. Consequently, working in the competitive workforce has been a goal of the movement towards social inclusion (Stephens et al., 2005).

But the disabled do not form a homogeneous group. They may be affected by a congenital disability, or one acquired during infancy or adolescence, or equally later, during higher education or active life. The disability may not practically influence the person's capacity to work and participate in society, or, on the contrary, may have a serious repercussion, meaning a need for considerable help and support, with multiple variations between these two extremes. The social context is also important and thus, while in some societies the disabled they are considered valuable members of the active population, in others they are not considered suitable for employment; being apparent that in all countries the unemployment rates of the disabled are much higher than the average. Thus, in the United States, only 3 in 10 disabled people aged 16 to 64 work part- or full-time. In the European Union in 2003, 40 per cent of disabled people of working age were employed compared to 64.2 per cent of persons without a disability; whereas in Paraguay, 18.5 per cent of people with disabilities participate in the labour force compared to 59.8 per cent of their non-disabled counterparts (ILO, 2009).

To improve this situation, the legal frameworks and regulations have been significantly modified over the last three decades and many governments have implemented policies aimed at promoting the right of the disabled to be integrated as fully as possible into society. One of the references has been the Declaration of the fundamental workplace principles and rights of the International Labour Organization (ILO, 2007), that proposes active policies to combat the discrimination against these workers in three ways: (1) Via the formulation of policies and regulations against discrimination in the workplace; (2) Via an increase in the opportunities for people liable of being discriminated to find a job; (3) Via an improvement of the hiring procedures in public and private sectors (Miralles et al., 2010).

In this framework, one of the strategies most commonly adopted in many countries to facilitate the integration of disabled workers into the labor market has been the creation of sheltered work centers for Disabled (henceforth SWDs). This model of socio-labor integration tries to move away from the traditional stereotype that considers disabled people as unable to develop continuous professional work. Just as in any other firm, a SWD competes in real markets and must be flexible and efficient enough to adapt to market fluctuations and changes, the only difference being that the SWD is a Not-For-Profit organization and most of its workers (normally around 70%) must be disabled. Thus, the potential benefits that may be obtained from increased efficiency are usually invested into the growth of the SWD. This results in more jobs for the Disabled and the gradual integration of people with higher levels of disability, which are in fact the primary aims of every SWD.

In countries such as Spain, this labour integration formula has been really successful in decreasing the former high unemployment rates, and one of the most usual strategies used by SWDs to facilitate this integration has been the adoption of assembly lines. In this sense Miralles et al. (2007) have shown how the integration of disabled workers in the productive systems can be done without losing, even gaining, productive efficiency through the use of assembly lines. In particular, the traditional division of work into single tasks seems to be a useful tool for making certain worker disabilities invisible; even becoming a good method for therapeutic rehabilitation of certain disabilities if appropriate task assignments and job rotation mechanisms are applied (Costa and Miralles, 2009).

### **1.1. Real integration formulas**

But, despite the great legislative efforts made by multiple national and international institutions, total social-employment integration of people with disabilities still seems far away. This fact confirms the perception that the solution has to come not only by special work formulas like SWDs, but also by overcoming the prejudices about the capabilities of the disabled, and by the genuine commitment of ordinary companies to include integration programs in their strategies and models. In this context the Corporate Social Responsibility policies of companies should

include those people with disabilities as a priority within the stakeholder group of employees; empowering voluntarily their integration in the production systems, and seeing this as an opportunity, and not as an imposition.

The incorporation of disabled people to many productive activities generates added value to a company, as well as to society as a whole, which is particularly timely given the current pensions crisis and ageing population. Moreover, to integrate disabled workers into normal companies would doubly contribute to the normalization: on one hand those that are integrated as work colleagues into the staff of ordinary companies, with the consequent benefit for them; and on the other hand, the fact that the other workers without a disability begin to perceive as normal an heterogeneous working environment where all workers, with or without some form of disability, undertake the most appropriate tasks according to their abilities and capacities. With this, it is not intended to play down the great contribution that SWDs have made towards diminishing the high levels of unemployment in this difficult sector, but it is also certain that, if the efforts to provide work to these people are only focused on these centers, the separation between ordinary workers and those with a disability is made more critical, relegating the latter to work centers away from the rest.

This is especially true for the case of developmental disabilities, where employment not only provides economic benefits to the individuals and the community, but it also contributes to the development of various adaptive skills such as physical abilities, cognitive abilities, and social skills. In this sense, the longitudinal study of Stephens et al. (2005) also confirms that beneficial skills appear to be learned within integrative work settings and lost within segregated work settings.

## **1.2. Contribution of this work**

As introduced earlier, some previous studies have highlighted the importance of assembly lines as a means to integrate as many disabled workers as possible into the workforce of SWDs (Miralles et al. (2007, 2008), Costa and Miralles (2009); Chaves et al. (2009); Blum and Miralles (2011)). These references provide different approaches to face the so-called Assembly Line Worker Assignment and Balancing Problem (ALWABP); which aims to model the heterogeneous scenario of SWDs assembly lines, where different task times are considered depending on the worker assigned.

Although these pioneer references have had a great importance in giving visibility to this social problem throughout our academic area, they may be too centered in the specific scenario of SWDs; where most workers are supposed to be disabled. In the classical ALWABP framework each task has a worker-dependent processing time, which allows taking into account the limitations of each worker depending on his/her disability. Therefore the input data are normally expressed by a time matrix where for every task, several possible times are available depending on the worker; and where task times in the matrix normally have a very high variability.

As it has been introduced, the desirable fact in this scenario would be to integrate the disabled not only in SWDs but also in ordinary companies. Therefore we should face the ALWABP with more open hypothesis of diversity, where for example these two situations are likely to occur:

- Ordinary companies should be able to integrate efficiently one only disabled worker into an assembly line; normally being this worker much slower than the rest. In these cases the parallelization of workstations is a very interesting extension for ALWABP, which can easily integrate disabled workers without losing much efficiency.
- In other cases the disabled workers to be integrated may have incompatible assignments that permit few possible solutions to ALWABP. In these cases, an

efficient way to widen the chances of possible assignments is the “Collaborative ALWABP” approach; where different workers can complement their capabilities collaborating in the same product within certain workstation.

The aim of this paper is to present new models and approaches for these two extensions of ALWABP which, even attending specific requirements that would normally arise in ordinary companies (with few disabled workers), can perfectly be adopted generally by SWDs. The same way around, the classical approaches of ALWABP are completely valid for an environment with less diversity simply by modifying the input data. In this sense, this paper will use data with more or less diversity independently.

The general aim is keep providing the production managers with practical tools that ease the integration of disabled in different scenarios since, obviously, the source of inspiration of a problem is not its only application field.

### **1.3. Paper outline**

In Section 2, we state a formal codification of the two extensions proposed for ALWABP; analysing their practical implications and reviewing those references of the literature with useful related approaches. Section 3 then presents the corresponding IP models for both the parallel and the collaborative approach of ALWABP; while Section 4 proposes resolution procedures for each extension. Finally, Section 5 summarizes the computational results, and in Section 6 we discuss the main findings and offer some conclusions and further research lines.

## **2. Assembly line Worker Assignment and Balancing Problem**

The so-called Simple Assembly Line Balancing Problem (SALBP) was initially defined by Baybars (1986) through several well-known simplifying hypotheses. This classical single-model problem, which consists of finding the best feasible assignment of tasks to stations so that certain precedence constraints are fulfilled, has been the reference problem in the literature in its two basic versions: when the cycle time  $C$  is given, and the aim is to optimize the number of necessary work stations, the problem version is called SALBP-1. Whereas when there is a given number  $m$  of work stations, and the goal is to minimize the cycle time  $C$  the literature knows this second version as SALBP-2 (Scholl, 2006).

SALBP is based on a set of limiting assumptions which reduce the complex problem of assembly line configuration to the “core” problem of assigning tasks to stations. The balancing of real-world assembly lines will, however, require the observation of a large variety of additional technical or organizational aspects, which will heavily affect the structure of the problem. Among the extensions considered in the literature are parallel work stations and tasks, cost synergies, processing alternatives, zoning restrictions, stochastic and sequence-dependent processing times as well as U-shaped assembly lines (Boysen et al. 2008).

We focus on the Assembly Line Worker Assignment and Balancing Problem (ALWABP), a generalization of SALBP where, in addition to the assignment of tasks to stations, a set of heterogeneous workers that execute the tasks also has to be assigned to stations (Miralles et al. 2008). In this scenario each task has a worker-dependent processing time, which allows taking into account the limitations of each worker depending on his/her disability. Thus, in ALWABP the input data are normally expressed by a precedence network and a time matrix like the one in Table 1, where for every task (columns) several operation times (rows) are possible depending on the worker. Even when the time to execute a task for certain worker is very high, this assignment is considered infeasible and the incompatibility is represented by a dash in the corresponding cell of the input data matrix:

Table 1. Example of ALWABP input data (execution times in seconds).

Tasks:	1	2	3	4	5	6	7	8	9	10
<b>W1</b>	4	3	9	5	9	4	8	7	5	1
<b>W2</b>	-	1	2	3	4	4	-	4	2	1
<b>W3</b>	4	3	9	5	2	2	5	3	2	2
<b>W4</b>	4	1	5	1	2	3	4	2	-	1

In our example, assuming the precedence network of the Figure 1 below, one possible solution would be the assignment coded in colors in the same Figure 1. In this solution workers W4, W2 and W3 are assigned to the first, second and fourth stations respectively; while the bottleneck (slowest station) is the third station T3, where the worker W1 develops the tasks 6 (4s) and 7 (8s). Therefore the cycle time of the assembly line is defined by the cycle time of the bottleneck station, which is 12 seconds.

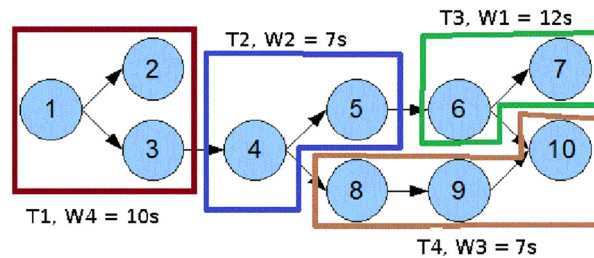


Figure 1: Precedence network and ALWABP solution.

Similarly to what occurs in SALBP, when we wish to minimize the number of stations, the problem is called ALWABP-1 and when the objective is to minimize the cycle time, the problem is called ALWABP-2. In fact this last situation is the most common at SWDs, and was the inspiration for the first mathematical model presented in Miralles et al. (2007), where it is used to model and solve an ALWABP-2 case study of a Spanish SWD.

This applied approach is aligned with a recent major trend of the scientific community aiming to narrow the gap between research and practice that used to exist in this field. A recent attempt to classify all the work made over the last years towards modeling real world assembly lines is the classification scheme provided by Boysen et al. (2007), that structures the vast field of assembly line balancing problems by means of a notation consisting of three elements  $[\alpha|\beta|\gamma]$ , where:

- $\alpha$  concerns the precedence graph characteristics;
- $\beta$  concerns the station and line characteristics;
- and  $\gamma$  concerns the optimization objectives.

Various possible values are defined for each of these three elements, covering all assembly line balancing problems appeared in the literature, and where the absence of values indicates the assumption of SALBP hypothesis (e.g. SALBP-2 is coded as  $[ | |c]$ ).

In this classification Boysen states ALWABP-2 as  $[pa,link,cum|equip|c]$  since we have different workers that suppose processing alternatives ( $\alpha = pa$ ), where tasks are linked to the selected worker within every station ( $\alpha = link$ ), and where the task assignment has incompatibilities and is subject to constraints on the cumulated time assigned ( $\alpha = cum$ ) to the station.

Note that the fundamental difference of ALWABP-2 with respect to other equipment selection problems ( $\beta = equip$ ) is the fact that resources are constrained: each unique worker can only be assigned to exactly one work station. In some cases workers with similar characteristics are considered, but even in these cases there is not an infinite number of resources available, as assumed in most robotic or automated lines, or in general in the so called assembly system design problems (Pinto et al. (1983), Rubinovitz and Bukchin (1993), Pinnoi and Wilhelm (1997, 1998), Bukchin and Rubinovitz (1997), Wilhelm (1999), Dolgui et al. (1999, 2006), Nicosia et al. (2002), Bukchin and Tzur (2000), Nicosia et al. (2002), Rekiek et al. (2002)). Moreover, in contrast to the ALWABP-2 none of these references has the objective of minimizing the cycle time ( $\gamma = c$ ) given certain resources, as ALWABP-2 does. Instead they all aim to minimize the cost ( $\gamma = Co$ ) or the resources implied ( $\gamma = m$ ); objectives that here would seem contradictory, since the objective of integrating disabled workers means to assign them, and not to minimize their number or try to avoid their participation.

Thus, as it has been introduced, two new extensions of ALWABP will be presented in this paper:

- One extension allowing the parallelization of workstations, what can be coded as  $[pa, link, cum|pstat, equip|c]$ .
- And a second extension considering the possible collaboration of various workers mounting the same product within one station  $[pa, link, cum|pwork, equip|c]$ .

In the next two sections these extensions will be described through simple examples, analysing their specific implications and reviewing those references of the literature that use similar approaches. Once the main features of each extension are clarified, a formal definition of their corresponding mathematical models will be presented (Section 3).

### 2.1. Extension 1: ALWABP with parallel stations

In this first extension the constraint of one worker per station is relaxed, so that the same subset of tasks can be assigned in parallel to two (or more) workers within the same station. To facilitate presentation, we will say that each worker is assigned to a station in the same stage of the line. Using the same example of last section, the following assignment could be a solution using this extension:

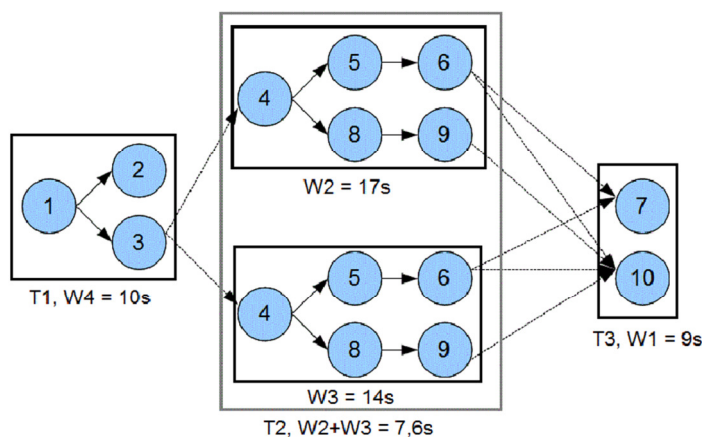


Figure 2: ALWABP with parallel stations.

In this solution we have worker W4 in the first stage developing tasks 1 (4s), 2 (1s) and 3 (5s); and worker W1 in the third stage developing tasks 7 (8s) and 10 (1s). However, the second stage is composed by two stations in parallel with worker W2 and worker W3. They both develop the same tasks (4, 5, 6, 8 and 9), but with their own operation times according to the

time matrix of Table 1. This fact makes it difficult to calculate the overall cycle time of the second stage, which is not as obvious as in Bard (1989), Sarker and Shanthikumar, (1983), or Pinto et al. (1981). Indeed, in order to compute this parameter, it is easier to express the productive parameters in terms of throughput rates (parts produced per unit of time); as will be exposed in the following example:

---

E.g. Let us define a productive system A working at a cycle time of  $C_A=1$  min/part, then we can assume that its throughput rate is  $TR_A=60$  parts/hour. Let us define another productive system B in parallel with A producing the same parts but with a cycle time of  $C_B=2$  min/part, which is  $TR_B=30$  parts/hour. What is very intuitive is that the overall Throughput Rate of both systems working in parallel is the sum:  $TR_{AB} = TR_A + TR_B = 60 + 30 = 90$  parts/hour; and therefore the overall cycle time is given by  $\frac{1}{C_{AB}} = \frac{1}{C_A} + \frac{1}{C_B} = \frac{1}{1} + \frac{1}{2} \rightarrow C_{AB}=0.67$  min/part

---

Then, in our case the cycle time calculated will be:  $\frac{1}{C_{T2}} = \frac{1}{C_{W2}} + \frac{1}{C_{W3}} = \frac{1}{17} + \frac{1}{14} \rightarrow C_{T2}=7.6$  s

Whenever we have two productive systems in parallel we can assume this way of calculating the cycle time. Thus, applying parallelization to ALWABP has implications in the model linearity, that can be solved through the mechanisms that will be exposed in section 3.

Apart from these mathematical implications in the model, the parallelization has also some practical implications that can be summarized as follows:

- As there are two (or more) workers in parallel in the same stage developing the same tasks, they need two different products. Therefore, a sufficient buffer of work in progress products must be set so that no lack of material appears for any of the workers. As soon as the cycle time of the parallelized stage is balanced with respect to the rest of stages, this quantity will be limited by an additional unit in most cases.
- The parallelization of stations suppose the duplication of facilities and tools that may be necessary for every worker in the stage. This practical implication has to be taken into account specially when there are physical constraints or high installation costs.
- In case of workers with very slow operation times, it is much easier to integrate them if we allow parallel stations. This is especially important in ordinary companies, since we can combine slow and fast workers for integrating disabled workers with the minimum loss of efficiency (as will be demonstrated in section 5).
- As it happens with SALBP, when we allow parallelization the cycle time can be lower than the largest operation time. In the case of ALWABP lower than the slowest task to execute (obtained with a max-min operation over all execution times).

Despite the fact that most assembly lines addressed in the literature are quite different from ALWABP, it is important to review the most important references that face somehow parallelization. The research on the inclusion of parallel workstation within assembly lines is limited and of relatively recent origin (see pioneer studies Pinto et al. (1981), Sarker and Shanthikumar, (1983) or Bard (1989). Askin and Zhou (1997) consider mixed-model assembly lines with task dependent equipment costs and arbitrary number of parallel workstations. They propose a heuristic procedure that uses a threshold value for the equipment utilization. Süer (1998) propose a simple resolution method in three phases, while McMullen and Frazier (1997, 1998) propose first a heuristic and later a simulated annealing technique for a problem with parallel stations and stochastic times. In both, they measure the line performance by the total cost and the extent to which cycle time is met. Vilarinho and Simaria (2002) introduce a two-stage heuristic method for mixed-model assembly lines, where their primary goal is to



minimize the number of workstations for a given cycle time, and the secondary goal is to balance the workload between workstations. Bukchin and Rubinovitz (2003) study the equipment selection problem on parallel workstations, investigating the influence of assembly sequence flexibility and cycle time on the balancing improvement due to the station paralleling. Simaria and Vilarinho (2004) propose a model and a genetic algorithm for a mixed model assembly line with parallel stations. Two years later the same authors propose in (Vilarinho and Simaria, 2006) an ant colony optimization algorithm for balancing the mixed-model assembly lines with zoning restrictions and parallel workstations. Boysen and Fliedner (2006) present a versatile graphic algorithm that can be adapted for parallel stations, and Ege (2009) consider parallel workstations and propose a branch and bound algorithm to minimize total equipment and workstation opening costs. Finally, Scholl and Boysen (2009) consider a problem first introduced by Gökçen et al. (2006) and named as parallel assembly line balancing problem (PALBP) where multiple products have to be assigned to parallel lines that have to be balanced.

To our knowledge, no reference has used parallel stations in ALWABP. From section 3 on, this paper will contribute to model and propose resolution procedures for this extension.

## 2.2. Extension 2: Collaborative ALWABP

Another way of enlarging the potential of assembly lines for integrating the disabled in the workforce is the “collaborative ALWABP” approach we propose here. In this variant, different workers can be assigned to the same station, where they collaborate mounting the same product. This doesn’t mean to set parallel workplaces like in the previous extension: in the collaborative ALWABP framework, even if workers share the same stage, they are assigned different tasks.

In general, many products manufactured on assembly lines are large enough to be worked at by several workers simultaneously. As a major example, we can cite the final assembly of cars in the automobile industry, where up to five workplaces are installed within a single station on a paced assembly line (Becker and Scholl, 2009). But the characteristic of the collaborative approach that makes it particularly interesting in this scenario is the allowance of many more feasible combinations than the conventional ALWABP, what is especially useful in those cases where certain incompatibilities may difficult the worker integration.

Let us illustrate this fact using the same input data matrix of Table 1. Let us imagine that we are applying a simple constructive heuristic (like one of those proposed in Miralles et al. 2008 or the ones that will be presented in Section 4) for this ALWABP example. We can start by building up a first partial solution where tasks 1 and 2 are developed by worker W1 in the first station; and tasks 3, 4 and 5 are assigned to worker W2 in the second station:

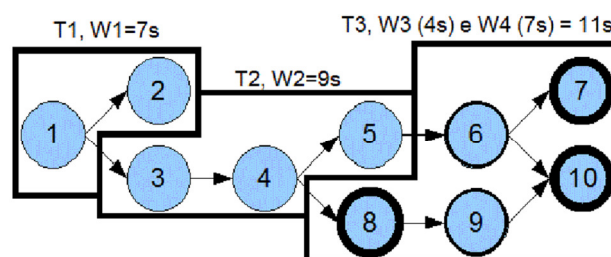


Figure 3: Collaborative ALWABP solution.

At this point, the best remaining assignment is the one shown by the shaded cells in the following table:

Table 2: Remaining input data matrix.

Tasks:	6	7	8	9	10
W3	2	5	3	2	2
W4	3	4	2	-	1

We notice that worker W4 is very efficient with task 7, and also assigning him task 8 together with task 10 would be, initially, a good combination. But, due to precedence constraints, this assignment of task 8 and 10 would only be possible if he also assumed task 9, which is incompatible for him/her.

In cases like this the flexibility introduced by the collaborative approach is particularly useful. The collaborative solution would assume a shared station where worker W4 initially develops task 8, then worker W3 executes task 9 and 6 in the same product, and then worker W4 can finish this product executing tasks 7 and 10. Following this work cycle the precedence constraint is not a problem anymore, and the third station becomes the bottleneck with a cycle time of 11 seconds (even if certain idle times exist for workers W3 and W4 within these 11 seconds). Note that, although this simple example has other feasible combinations that would improve the solution, in more complex cases the collaborative solution might considerably improve the optimal serial solution, as suggested by the computational tests in Section 5.

Although the literature on collaboration inside assembly lines is quite scarce, we can find some interesting references of problems that somehow require the collaboration of various workers due to different reasons. For example, when large workpieces are assembled, it is possible that different operators work at the same product unit simultaneously (Akagi et al., 1983). Several workplaces are also installed at the same station in those cases where each worker gets an own set of tasks at individual mounting positions, ensuring that the workers do not interfere with each other. In case of a two-sided line (2ALBP), each station consists of (up to) two workplaces, one at the right and one at the left side of the line (see Bartholdi, 1993; Kim et al., 2000; or Lee et al., 2001). Moreover, there are often incompatible tasks that must not be assigned to the same station, e.g. due to the danger of soiling the seats of a car if the same worker has to handle the seats and to lubricate movable parts (Rachamadugu, 1991). Finally, Becker and Scholl (2009) specifically address a collaborative approach inside workstations ( $\beta = pwork$ ) in a problem that they call assembly line balancing problem with variable workplaces (VWALBP). The variant VWALBP-2 is coded as  $[inc, fix, type, ch|pwork|c]$ . The problem is defined and modelled as an integer linear program, and a branch-and-bound procedure is proposed which can also be applied as a heuristic.

Once reviewed the references that apply intra-station collaboration, and once exposed the reasons why both collaboration and parallel stations are important extensions of ALWABP, the next sections are devoted to formulating their IP models and presenting resolution procedures.

### 3. Mathematical models

In this section, we describe mathematical models associated to the ALWABP variants presented earlier. We use the following notation:

$N$	set of tasks;
$W$	set of workers;
$S$	set of stages;
$D_j$	set of tasks that precede task $j$ in the precedence graph;
$I_w$	set of tasks that worker $w$ is not able to execute;
$p_{wi}$	time spent by worker $w$ to execute task $i$ ;

The original formulation for the ALWABP was proposed by Miralles et al (2007) and contains a continuous variable indicating the cycle time and two sets of binary variables, as indicated below:

$C$	cycle time of the line,
$x_{swi}$	binary variables equal to one if worker $w$ executes task $i$ in station $s$ and to zero, otherwise, $i \in N, w \in W, s \in S$ .
$y_{sw}$	binary variables equal to one if worker $w$ is designated to station $s$ , $w \in W, s \in S$ .

With these variables, the authors were able to write the following linear model for the ALWABP:

$$\begin{aligned}
 & \text{Min } C && (1) \\
 \text{Subject to} & \sum_{w \in W} \sum_{s \in S} x_{swi} = 1, && \forall i \in N, && (2) \\
 & \sum_{s \in S} y_{sw} = 1, && \forall w \in W, && (3) \\
 & \sum_{w \in W} y_{sw} = 1, && \forall s \in S, && (4) \\
 & \sum_{w \in W} \sum_{s \in S} s x_{swi} \leq \sum_{w \in W} \sum_{s \in S} s x_{swj}, && \forall i, j \in N | i \in D_j, && (5) \\
 & \sum_{w \in W} \sum_{i \in N} p_{wi} x_{swi} \leq C, && \forall s \in S, && (6) \\
 & \sum_{i \in N} x_{swi} \leq |N| y_{sw}, && \forall w \in W, \forall s \in S, && (7) \\
 & x_{swi} = 0, && \forall w \in W, \forall s \in S, \forall i \in I_w. && (8)
 \end{aligned}$$

The objective function (1) minimizes the cycle time. Constraints (2) guarantee the execution of each task in one station by a single worker, while constraints (3) and (4) create a biunivocal relationship between workers and stations. Constraints (5) handle the precedence relationships between tasks, ensuring that a task  $i$  that must precede a task  $j$  is only assigned to the same station to which  $j$  has been assigned or to an earlier station. Constraints (6) define the cycle time as the total execution time of the bottleneck station, while constraints (7) guarantee coherence between the binary variables: a task can only be executed by a worker in a station if that worker is assigned to that station. Finally, constraints (8) handle *a-priori* assignment infeasibilities.

In this article, we propose extensions to formulation (1)-(8) in order to model the two new problem variants described in Section 2. The first new model, presented below, deals with the parallel ALWABP. In this model, the definition of the set  $S$  described earlier is changed, and each element of  $S$  represents no longer a station but a stage (which might contain at most  $K_{max}$  stations). The definitions of variables  $x_{swi}$  and  $y_{sw}$  are modified accordingly. Moreover, in order to describe this model, the following additional variables are defined:

- $t_{is}$  binary variables equal to one if task  $i$  is executed in stage  $s$  and to zero, otherwise,  $i \in N, s \in S$ .
- $z_{sk}$  binary variables indicating the level of parallelism of a stage, equal to one if stage  $s$  contains  $k$  stations in parallel and to zero, otherwise. ( $s \in S, k=1 \dots K_{max}$ ).

With these new variables, the parallel ALWABP model can be written as:

$$\text{Min } C \quad (9)$$

$$\text{Subject to: } \sum_{s \in S} s t_{is} \leq \sum_{s \in S} s t_{js}, \quad \forall i, j \in N | i \in D_j, \quad (10)$$

$$\sum_{w \in W} x_{swi} \geq k(t_{is} + z_{sk} - 1), \quad \forall s \in S, \forall k \in [0, K_{max}], \forall i \in N, \quad (11)$$

$$x_{swi} \leq t_{is}, \quad \forall s \in S, \forall w \in W, \forall i \in N, \quad (12)$$

$$\sum_{k=0}^{K_{max}} z_{sk} = 1, \quad \forall s \in S, \quad (13)$$

$$\sum_{w \in W} y_{sw} = \sum_{k=0}^{K_{max}} k z_{sk}, \quad \forall s \in S, \quad (14)$$

$$\sum_{s \in S} y_{sw} = 1, \quad \forall w \in W, \quad (15)$$

$$\sum_{s \in S} t_{is} = 1, \quad \forall i \in N, \quad (16)$$

$$x_{swi} = y_{sw}, \quad \forall s \in S, \forall w \in W, \forall i \in N, \quad (17)$$

$$x_{swi} = 0, \quad \forall w \in W, \forall s \in S, \forall i \in I_w, \quad (18)$$

$$\sum_{w \in W | y_{sw}=1} \frac{1}{\sum_{i \in N} p_{wi} x_{swi}} \geq \frac{1}{C}, \quad \forall s \in S | z_{s0} = 0. \quad (19)$$

As before, the objective function (9) minimizes the cycle time. Constraints (10) represent the precedence relationships between tasks and are similar to constraints (5). Constraints (11) force that in a stage with  $k$  stations, each task be executed by  $k$  different workers. Constraints (12) state that a worker can execute a task in a stage only if that task is assigned to that stage, while constraints (13) indicate that each stage has only one level of parallelism. Notice that the number of stations in a stage can be zero, that is, a stage can be empty (this only indicates the fact that the number of workers is fixed and, by assigning more than a worker to a stage, the final number of stages must be reduced). Constraints (14) force the assignment of  $k$  workers to a stage with  $k$  stations. Constraints (15) and (16) are concerned with the fact that each worker and each task must be assigned to a single stage, while constraints (17) guarantee that a worker only executes tasks in a stage if that worker is assigned to that stage. Constraints (18) handle the *a-priori* infeasibility assignments. Finally, constraints (19) calculate the cycle time as the inverse of the throughput rate of the assembly line, which is limited by the lowest throughput rate among non-empty stages. The throughput rate of a stage is the sum of the throughput rates of each worker assigned to that stage, which can be computed as the inverse of the task execution times for each worker (see example in Section 2.1). Due to constraints (19), model (9)-(19) becomes non-linear.

In order to linearize restrictions (11) we define  $F$  as the throughput rate of the assembly line, and  $f_{sw}$  as the throughput rate of worker  $w$  in stage  $s$ . If  $w$  is assigned to stage  $s$ , that is, if  $y_{sw}=1$ , then  $f_{sw} = \frac{1}{\sum_{i \in N} \sum_{k=0}^{K_{max}} p_{swi} x_{swi}}$ , which is equivalent to  $\sum_{i \in N} \sum_{k=0}^{K_{max}} p_{swi} f_{sw} x_{swi} = 1$ . Otherwise,  $f_{sw}=0$ .

The linearization is completed with the definition of additional variables  $v_{swi} = f_{sw} x_{swi}$ . Therefore,  $v_{swi} = f_{sw}$  if the task  $i$  is executed by worker  $w$  in stage  $s$  and  $v_{swi} = 0$ , otherwise. Using a large constant  $M$ , a linear model for the ALWABP with parallel stations can be written as follows:

$$\text{Max } F \quad (20)$$

Subject to: (10) – (18),

$$\sum_{w \in W} f_{sw} \geq F - M z_{s0}, \quad \forall s \in S, \quad (21)$$

$$\sum_{i \in N} p_{swi} v_{swi} = y_{sw}, \quad \forall w \in W, \forall s \in S, \quad (22)$$

$$f_{sw} \leq M y_{sw}, \quad \forall w \in W, \forall s \in S, \quad (23)$$

$$v_{swi} \geq f_{sw} - M(1 - x_{swi}), \quad \forall s \in S, \forall w \in W, \forall i \in N, \quad (24)$$

$$v_{swi} \leq M x_{swi}, \quad \forall s \in S, \forall w \in W, \forall i \in N. \quad (25)$$

The new objective function (20) maximizes the assembly line throughput rate. This is equivalent to the former objective of minimizing the line cycle time. Constraints (19) were replaced by their linearized form (21)-(25). Constraints (21) define the production throughput rate as the throughput rate of the slowest non-empty stage. Constraints (22) define the values of the variables  $v_{swi}$  as the throughput rate obtained by a worker if that worker is assigned to the station and to zero, otherwise. Constraints (23) state that, if a worker  $w$  is not assigned to a stage  $s$ , then  $f_{sw}=0$ . Constraints (24) and (25) force variables  $v_{swi}$  to assume appropriate values: If  $x_{swi} = 1$ , then  $v_{swi} = f_{sw}$ , otherwise,  $v_{swi} = 0$ .

Contrary to what happened in the ALWABP with parallel stations case, the collaborative ALWABP model can be obtained with very simple modifications in the original formulation (1)-(8). Indeed, only constraints (4) need to be changed, as indicated below:

$$\sum_{w \in W} y_{sw} \leq K_{max}, \quad s \in S. \quad (26)$$

The new constraint (26) relaxes the original constraints (4) so that more than a worker can be designated to a single stage. In this case, according to equation (7), any of the assigned workers can execute any of the tasks associated to that stage (and the model will naturally choose the fastest worker for each of the tasks). Moreover, equation (3) will force the use of the available number of workers.

The efficiency and validity of the two presented models have been tested through a set of computational tests presented in Section 5. As expected, the difficulty of exact resolution of the models (which was already the case for model (1)-(8)) justified the development of heuristic approaches, which will be presented in the following section.

#### 4. Heuristic solution method

In this section we describe a heuristic procedure that can be used to solve the two new proposed extensions. This procedure relies on the construction of a solution that designates workers and tasks to the stages in a sequential manner and is adapted from the SALBP and ALWABP literatures.

In the following we first present a description of the general algorithmic scheme in which the heuristic is used. Then, we describe the task and worker selection priority rules, stating in both cases the particularities for parallel and collaborative scenarios. These particularities are illustrated through the example exposed in section 4.2. Finally, two further variants of the heuristic developed are described in section 4.3.

##### 4.1. Algorithmic scheme

The core characteristic of the proposed method is the sequential resolution of version-1 assembly line balancing problems (minimization of the number of stations given a maximum allowed cycle time) in order to solve the version-2 problem. Algorithm 1 below presents a pseudo-code of this idea.

<b>Algorithm 1:</b> Station oriented procedure for version-2 assembly line balancing problems
<b>Requires:</b> Lower and upper bounds on cycle time, LC and UC, respectively
<ol style="list-style-type: none"><li>1. <math>\bar{c} = LC</math>;</li><li>2. <b>while</b> <math>\bar{c} &lt; UC</math>, <b>do</b></li><li>3.     Solve version-1 problem;</li><li>4.     <b>if</b> version-1 problem finds a solution with cycle time less or equal to <math>\bar{c}</math></li><li>5.         <b>END</b>;</li><li>6.     <b>Else</b></li><li>7.         increase <math>\bar{c}</math>;</li><li>8.     <b>end if</b></li><li>9. <b>end while</b></li><li>10. <b>Return</b> “no solution found”; <b>END</b>.</li></ol>

The strategy of the algorithm is quite straightforward. For each possible cycle time, a version-1 assembly line balancing problem is solved. If a solution containing at most the desired number of stations is found, then there is a solution with the current cycle time for the version-2 problem and the algorithm ends (lines 5 and 6). Otherwise, the algorithm restarts with a larger cycle time value (line 7).

Scholl and Voß (1996) and Moreira et al. (2011) have proposed algorithms for the SALBP-1 and ALWABP-1, respectively. Both algorithms rely on a constructive resolution method, which sequentially assign tasks (SALBP-1) or task and workers (ALWABP-1) to the stations, respecting the cycle time. This assignment is done according to a criterion that might give priority, for example, to tasks with a large number of successors. In the case of ALWABP, the worker selection criterion might, for example, give priority to workers that assign the larger number of tasks at that point of the assembly line.

We propose a version-2 problem to be used in the cases of the parallel and collaborative ALWABP. We also define appropriate priority criteria for this situation and discuss the main particularities of the method in this context. Algorithm 2 presents a pseudo code of the method. In the algorithm,  $t(S_k)$  represents the load of stage  $k$ .

<b>Algorithm 2:</b> Station oriented procedure for the Parallel and collaborative ALWABP-1	
<b>Requires:</b> $K_{max}$ , desired cycle time, $\bar{c}$ .	
1.	$n = 0, t(S_k) = 0;$
2.	$U_w = W, U = N;$
3.	$W_{best} = \emptyset; T_{W_{best}} = \emptyset; cr(W_{best}, T_{W_{best}}) = 0;$
4.	<b>for</b> each $k = 1 \dots K_{max}$ <b>do</b>
5.	<b>for</b> each $W' \subset W,  W'  = K_{max}$ <b>do</b>
6.	compute set of assignable tasks (respecting the task priority criterion and $\bar{c}$ ), $T_{W'}$ ;
7.	compute worker priority criterion associated with the solution, $cr(W', T_{W'})$ ;
8.	<b>if</b> $cr(W', T_{W'}) > cr(W_{best}, T_{W_{best}})$
9.	$cr(W_{best}, T_{W_{best}}) = cr(W', T_{W'}); W_{best} = W'; T_{W_{best}} = T_{W'}$
10.	<b>end if</b> ;
11.	<b>end for</b> ;
12.	<b>end for</b> ;
13.	assign $(W_{best}, T_{W_{best}})$ to stage $n$ ; $U_w = U_w \setminus W_{best}; U = U \setminus T_{W_{best}};$
14.	<b>if</b> $U_w \neq \emptyset$
	<b>goto</b> step 3;
	<b>else</b>
	<b>if</b> $U \neq \emptyset$
16.	a solution has been found; <b>END</b> ;
17.	<b>Else</b>
18.	no solution has been found for $\bar{c}$ ; <b>END</b> ;
19.	<b>end if</b> ;
20.	<b>end if</b> .

A few comments are in order. Algorithm 2 computes, for each possible subset of workers respecting the maximum parallelization (lines 4 and 5), the set of tasks that would be assigned to the current stage,  $T_{W'}$ , and the associated worker assignment criterion (lines 6 and 7). The set with best criterion is chosen and the workers and tasks are assigned to the stage (line 13). This continues until no more workers are available. At this point, if all tasks have been assigned, a solution with cycle time of at most  $\bar{c}$  has been found (line 16). Otherwise, the algorithm returns a message indicating that it was not able to find a feasible solution for the desired cycle time.

Several details must be defined in order to have a functional version of Algorithm 2, including the priority criteria associated with the selection of tasks (line 6) and workers (line 7), and the computation of the cycle time associated with a given solution (see example in section 4.2).

#### 4.1.1. Task selection (line 6)

In what concerns the task priority criteria, we adapted some of the task priority rules used by Scholl and Voß (1996). Since some of these rules calculate the priority of the tasks based on their execution times, in ALWABP they need to be modified in order to use the appropriate parameters available, i.e., the task execution times depending on the selected worker(s) (Moreira et al. 2011).

In the specific case of the parallel and collaborative ALWABP, a new problem arises since task times are not well defined for sets of more than one worker. Indeed, consider the parallel ALWABP: the increase in the current stage load depends not only on the selected workers and on the current task, but also on the tasks that are already assigned to the stage (due to the non-linear characteristics associated with the cycle time computation). The solution found was to define the execution time of task  $i$  in a dynamic fashion:

- For the parallel ALWABP, the execution time was defined as the increase in the execution time of a stage when task  $i$  was assigned to that stage (what needs some extra computation, since the new cycle time is calculated through the corresponding throughput rates (see example of section 2.1.)).
- In the collaborative case the adjustment is simpler, as the task execution time can be set to the minimum task execution time among the selected workers.

With these adjustments, eight task priority rules were defined:

1. **MinD**: increasing difference between the task execution time and the minimum task execution time.
2. **MinR**: increasing ratio between the task execution time and the minimum task time.
3. **MaxFTime**: decreasing ratio between the number of task followers and the task execution time.
4. **MaxIFTime**: decreasing ratio between the number of task immediate followers and the task cost.
5. **MaxPWS<sup>+</sup>**: decreasing maximum static positional weights.
6. **MaxPWS<sup>-</sup>**: decreasing minimum static positional weights.
7. **MaxPWD<sup>+</sup>**: decreasing maximum dynamic positional weights.
8. **MaxPWD<sup>-</sup>**: decreasing minimum dynamic positional weights.

As an example, consider the first criterion. For every assignable task (i.e., a task whose predecessors have all been assigned), the criterion is computed and the task with the least criterion value is “assigned” to the tentative solution.

Despite being quite simple criteria, some comments will certainly contribute to clarify them: the minimum task time is the lowest execution time for that task among the non-assigned workers. The (immediate) followers of a task are those tasks that come (immediately) after a task in the precedence graph. The positional weight of a task is the sum of execution times for a task and its followers. In the static positional weight we use the task execution times among all workers, while in the dynamic positional weight we use the task execution times among the non-assigned workers. One last comment concerns only the parallel ALWABP: in this case, if one of the workers of the stage is not able to execute the task, the assignment is discarded.

#### 4.1.1. Worker selection (line 7)

For the worker priority rules, we adapted two rules used by Moreira and Costa (2011): the rule **MaxTasks** selects the worker that is able to maximize the number of assigned tasks; while the rule **MinRLB** uses the decreasing values of a relaxed lower bound for the cycle time, computed as the sum of the lowest execution times for all non-assigned tasks divided by the number of non-assigned workers. Since sets with more than one worker will generally be able to execute more tasks, we added some penalties to make this rule fairer, yielding eight criteria, as it is shown below:

1. **MaxTasks1**: decreasing number of assigned tasks.
2. **MaxTasks1.2**: decreasing number of assigned tasks divided by 1.2.
3. **MaxTasks1.4**: decreasing number of assigned tasks divided by 1.4.
4. **MaxTasks1.6**: decreasing number of assigned tasks divided by 1.6.
5. **MaxTasks1.8**: decreasing number of assigned tasks divided by 1.8.
6. **MaxTasks2**: decreasing number of assigned tasks divided by 2.
7. **MaxTasks3**: decreasing number of assigned tasks divided by 3.
8. **MaxTasksK**: decreasing number of assigned tasks divided by the number of assigned workers.
9. **MinRLB**: decreasing values of the relaxed cycle time lower bound.



## 4.2. Example

Suppose the data in Table 1 are used and Algorithm 2 is called from within Algorithm 1 with a cycle time  $\bar{c} = 14$ . Also, let us assume that the first task priority rule (**MinD**) and the first worker priority rule (**MaxTasks1**) are being used and  $K_{max} = 2$ . Since all possible combinations of two available workers are available, the following combinations of  $W'$  will be tested at each stage (line 5):  $W' = \{1\}, \{2\}, \{3\}, \{4\}, \{1,2\}, \{1,3\}, \{1,4\}, \{2,3\}, \{2,4\}$  and  $\{3,4\}$ .

In order to exemplify the functioning of the algorithm, let us calculate the assigned tasks and worker priority criterion for the cases with  $W' = \{1\}$ ,  $W' = \{1,2\}$  and  $W' = \{1,4\}$ . These cases are selected because they present a view of different situations that can occur in the algorithm:

---

**Case 1 ( $W' = \{1\}$ ):** At first, the only assignable task is task 1. This task can be assigned and the stage load is updated to 4. At this point, both tasks 2 and 3 are assignable, and therefore, the task to be assigned is decided using the task priority criterion (MinD, in this case). For task 2,  $MinD = (\text{task execution time by worker 1}) - (\text{minimum task execution time among non-assigned workers}) = 3s - 1s = 2s$ . For task 3,  $MinD = 9s - 2s = 7s$ . Since MinD prioritizes the tasks with increasing value of the criterion, task 2 is selected and the load of the station is updated to 7. In order to respect the precedence constraints, only task 3 is assignable at this point. Nevertheless, assigning task 3 would yield a station load of 16, which exceeds the desired cycle time, ending the assignments for worker 1 with  $T_{W'} = \{1,2\}$  and  $cr(W', T_{W'}) = 2$  (nr of assigned tasks).

**Case 2 ( $W' = \{1,2\}$ ):** The only possibility is to try to assign task 1. However, this would imply in an infeasible assignment for worker 2, and therefore, it is discarded, yielding  $T_{W'} = \emptyset$ , and  $cr(W', T_{W'}) = 0$ .

**Case 3 ( $W' = \{1,4\}$ ):** Task 1 will be the first to be assigned. This implies a partial load  $t(S_k)$  given by  $\frac{1}{t(S_k)} = \frac{1}{4} + \frac{1}{4} = 0.5 \rightarrow t(S_k) = 2$ . The assignment of task 1 makes task 2 and 3 assignable. In order to select which task will be assigned, one must first compute the increase in load given by the inclusion of each of the candidate tasks. For task 2, the new load would be given by  $\frac{1}{t(S_k)} = \frac{1}{4+3} + \frac{1}{4+1} = 0.344 \rightarrow t(S_k) = 2.92s$ , i.e., an increase of 0.92s in the stage load. MinD for task 2 can therefore be computed as the difference between this increase and the minimum time needed by a non-assigned worker to execute the task:  $MinD = 0.92s - 1s = -0.08$ . For task 3, the new load would be given by  $\frac{1}{t(S_k)} = \frac{1}{4+9} + \frac{1}{4+5} = 0.188 \rightarrow t(S_k) = 5.32s$ , a load increase of 3.32 which yields a MinD of  $3.32s - 2s = 1.22$ . Therefore task 2 is selected. At this point, only task 3 is assignable. The assignment of this tasks results in  $\frac{1}{t(S_k)} = \frac{1}{4+3+9} + \frac{1}{4+1+5} = 0.162 \rightarrow t(S_k) = 6.15s$ . Task 4 becomes the only task assignable, yielding  $\frac{1}{t(S_k)} = \frac{1}{4+3+9+5} + \frac{1}{4+1+5+1} = 0.139 \rightarrow t(S_k) = 7.22s$ . Tasks 5 and 8 are now assignable. Following the same computations, the increase in time of these tasks can be computed as 1.85s and 1.66s, respectively, which yields associated MinD criteria of -0.15 and -1.34. Task 8 is, therefore, selected. The current stage load is given by  $t(S_k) = 8.88s$  and the assignable tasks are now tasks 5 and 9. Note that since the load of the stage changed, the increase in load must be recomputed for task 5, yielding 1.79s (and a MinD of -0.21), while task 9 cannot be included because worker 4 cannot execute this task. Task 5 is included and the current load updated to  $t(S_k) = 10.67s$ . Only task 6 is assignable at this point. Its inclusion yields a load of  $t(S_k) = 12.6s$ . No more tasks can be assigned and we have  $T_{W'} = \{1,2,3,4,5,6,8\}$  and  $cr(W', T_{W'}) = 7$ .

In this example, if the criterion **MaxTasks** was in use, the workers of set  $W' = \{1,4\}$  would be assigned to the stage and the algorithm would move to the next stage with the remaining workers and tasks.

---

### 4.3. Heuristic variants

Even though sets of workers are considered in line 5 of Algorithm 2, it might happen that the final solution obtained by the algorithm is still serial. Therefore, we also tested versions of the proposed heuristics including simple modifications in order to promote the finding of solutions containing at least one stage with more than a worker.

Two modifications were introduced:

**Mod 1:** In this version, the algorithm is run iteratively  $|S|-1$  times. At each run, a stage is selected and at least two workers are assigned to this stage. This is done by ignoring sets containing a single worker in line 5 of Algorithm 2, when the selected stage is being considered.

**Mod 2:** In this version, the algorithm is run iteratively  $|W|$  times. At each run, a worker is selected and this worker is only considered for assignment in conjunction with other workers.

In the following section, we test the proposed models and heuristics in a set of instances adapted from the literature; analyzing the results and extracting valuable conclusions.

## 5. Computational results

In this section, we present a set of computational experiments used to evaluate the efficiency of the proposed models and algorithms. We used the four families of instances proposed by Chaves et al. (2007). This benchmark was generated from four instances specifically selected from the well-known classical SALBP collection of Hoffmann (1990) so that problems with low and high order strength (which measures the structural properties of the precedence graph), and with low and high number of tasks were represented.

From each instance the benchmark includes 80 different problems that were generated attending high/low variability of task times, high/low number of tasks per worker, and high/low percentage of incompatibilities; resulting in a robust benchmark of 320 ALWABP-2 instances of varying characteristics.

Table 3 summarizes the main characteristics of the instances, including the number of instances (# instances), the number of tasks ( $|N|$ ), the number of workers ( $|W|$ ), and the order strength of the precedence graph.

Table 3. Instances characteristics.

Family	# instances	$ N $	$ W $	Order Strength
Roszieg	80	25	4:6	71.67
Heskia	80	28	4:7	22.49
Tonge	80	70	10:17	59.42
Wee-mag	80	75	11:19	22.67
MRoszieg1	40	25	4	71.57
MRoszieg2	40	25	4	71.57
MHeskia1	40	28	5	22.49
MHeskia2	40	28	5	22.49

We also create four additional sets of instances, based on the first 40 instances of the Roszieg and Heskia families (which have 4 workers). The first two sets, with 40 instances each, simulate a situation in which two workers have the same disability, and were created by inserting a 5<sup>th</sup> worker with execution times equal to those of an existing worker. We called these sets MRoszieg1 and MHeskia1. The other two sets of instances, also with 40 instances

each, simulate a situation in which one of the workers is very slow in comparison to the others. These sets were created by adding to each instance a 5<sup>th</sup> worker with execution times corresponding to three times the largest execution time for each task. We named these sets of tasks MRoszieg2 and MHeskia2.

In the following we present three different sets of experiments with their corresponding discussions: first, we evaluate the behavior of the exact models and discuss the improvement on certain parameters; then we present the results of the heuristic procedure using the different criteria for task and worker selection; and finally we present some interesting results on the heuristic variants.

In all experiments, the results obtained are compared with those obtained by serial layouts using the same set of workers. This allows for a fair comparison, since we want to verify if better results can be obtained by the parallel and collaborative ALWABP using the same resources that were available for the serial problem.

### 5.1. Serial vs parallel and collaborative approaches

The instances with 4 and 5 workers were solved using the commercial package IBM ILOG CPLEX 12.1 in a Intel Core 2 Duo T5450 processor, with 1.66 GHz and 3 GB of RAM. 4 presents the results obtained when solving the models for groups Heskia (first 40 instances), Roszieg (first 40 instances), MRoszieg1 and MHeskia1. The table shows the number of parallel and collaborative solutions that improved the best serial solution, the average improvement obtained for these cases, the best improvement obtained in each family and the average computational time needed to solve the model.

Table 4. Results for the exact resolution of the parallel and collaborative models. (Instances Roszieg, Heskia, MRoszieg1 and MHeskia1).

	Parallel ALWABP				Collaborative ALWABP			
	Number of improved solutions	Average improvement (%)	Best improvement (%)	Time (s)	Number of improved solutions	Average improvement (%)	Best improvement (%)	Time (s)
Roszieg	2	3.25%	4.90%	62.63	4	24.63%	43.42%	9.18
Heskia	0	0.00%	0.00%	143.45	0	0.00%	0.00%	1.24
MRoszieg1	7	3.44%	5.88%	29.91	4	7.03%	9.52%	8.49
MHeskia1	5	1.56%	3.17%	152.16	0	0.00%	0.00%	5.56

The number of workers in the instances of these families is only four. Even though, improvements could be obtained with the flexibility introduced by the new problems. For the parallel ALWABP, better configurations were found in 14 of the 160 instances (only the first 40 instances of the families Roszieg and Heskia were considered) while 8 improved solutions were found for the collaborative ALWABP. Due to the small number of workers, the interesting fact is not the number of instances that could be improved but the existence of instances that could strongly benefit from the new layouts. Indeed, an instance of the Roszieg family presented an improvement of 43.42% in efficiency (compared to the best serial solution) when the workers were allowed to work collaboratively. The parallel ALWABP, in turn, seemed to be useful for the modified instances (where 12 out of the 14 improvements were found). In these cases, the solution usually assigned the slowest worker in parallel with a faster worker. The figures in the table show that, in some cases, the use of collaborative strategies may significantly increase the throughput of the line.

Tables 5 and 6 show the results obtained by the methods for the second class of modified instances (MRoszieg2 and MHeskia2). The computational time was limited to 3600s for the resolution of each instance in this case.

Table 5. Results for the exact resolution of the parallel ALWABP, with  $K_{max}$  limited to 2 and 3. (Instances MRoszieg2 and MHeskia2).

	Parallel ALWABP							
	$K_{max}=2$				$K_{max}=3$			
	Number of improved solutions	Average improvement (%)	Best improvement (%)	Time (s)	Number of improved solutions	Average improvement (%)	Best improvement (%)	Time (s)
MRoszieg2	11	3.18%	10.53%	1730.26	11	3.18%	10.53%	1838.15
MHeskia2	20	4.45%	13.39%	2894.76	18	4.15%	13.39%	3077.08

Table 6. Results for the exact resolution of the collaborative ALWABP. (Instances MHeskia2 and MRoszieg2).

	Collaborative ALWABP				
	Number of improved solutions	Average improvement (%)	Best improvement (%)	Time (s)	Number of improved solutions (respect to parallel)
MRoszieg2	7	25.53%	50%	64.78	7
MHeskia2	0	0%	0%	24.67	0

These instances were much harder to solve and to reduce the computational times in the case of the parallel ALWABP,  $K_{max}$  was limited to 2 and 3. The results indicate that for these instances, the parallel ALWABP could again be useful, yielding improvements in 31 out of the 80 instances. The fact that the average improvements were higher for the case with  $K_{max} = 2$  is due to the fact that optimal solutions could not always be found in these simulations within the time limit of 3600s. When such situation happened, the best incumbent solution was used.

The collaborative ALWABP improved a smaller number of solutions but, when improved solutions were found the gains were important, up to 50% in one of the cases.

## 5.2. Evaluation of heuristics

For larger instances, exact methods were not able to obtain feasible solutions and we therefore tested the solutions yielded by the constructive heuristic approach. In tables 7 and 8, we present the results obtained by the heuristic method for  $K_{max} = 2$  and 3, respectively. In the tables, we present the gap (with respect to the best known serial solution), the percentage of parallel solutions found, the best improvement found and the computational time needed to solve the method. Each row of the tables presents the results for one of the worker priority criteria (associated with the best task priority criterion).

Table 7. Results for the heuristic resolution of the parallel ALWABP with  $K_{max} = 2$ .  
(Averaged over all instances).

	Gap	% of parallel solutions	Best improvement	Time (s)
MaxTasks1 + MaxPWD-	50.85%	90.21%	8.47%	6.78
MaxTasks1.2 + MinD	38.67%	81.88%	13.96%	0.30
MaxTasks1.4 + MinR	27.92%	70.21%	15.08%	0.27
MaxTasks1.6 + MaxPWD-	23.12%	59.79%	14.29%	0.58
MaxTasks1.8 + MinPWD-	21.10%	41.25%	14.29%	0.57
MaxTasks2 + MaxPWD-	19.93%	18.54%	14.29%	5.75
MaxTasks3 + MaxPWD-	19.24%	4.79%	13.49%	5.81
MaxTasksK + MaxPWD-	19.52%	10.00%	13.49%	5.81
MinRLB + MaxPWD-	14.92%	28.54%	14.29%	6.37

Table 8. Results for the heuristic resolution of the parallel ALWABP with  $K_{max} = 3$ .  
(Averaged over all instances).

	Gap	% of parallel solutions	Best improvement	Time (s)
MaxTasks1 + MaxPWD-	68.02%	90.21%	4.76%	28.78
MaxTasks1.2 + MinD	57.59%	75.04%	10.41%	1.51
MaxTasks1.4 + MinD	41.23%	52.96%	11.70%	1.34
MaxTasks1.6 + MinR	31.88%	39.74%	12.05%	1.25
MaxTasks1.8 + MinR	26.83%	31.59%	12.05%	1.19
MaxTasks2 + MaxPWD-	24.75%	29.17%	0.00%	23.68
MaxTasks3 + MaxPWD-	20.17%	7.29%	0.00%	22.71
MaxTasksK + MaxPWD-	21.02%	10.21%	0.00%	22.68
MinRLB + MaxPWD-	14.91%	28.75%	2.65%	24.60

For most of the worker priority criteria, the most effective task priority criterion was the decreasing minimum dynamic positional weights (**MaxPWD**). Among the worker priority criteria, **MinRLB** presented the best results, yielding a gap of 14.92% and 14.91% with respect to the best known serial solutions when  $K_{max}$  was set to 2 and 3, respectively. A considerable number of parallel solutions was obtained by the method and some important improvements in the line efficiency could be verified for some instances. The algorithm was efficient in computational terms, solving the instances in less than one minute in average.

As we can see below, Tables 9 and 10 present the analogous results for the collaborative ALWABP.

Table 9. Results for the heuristic resolution of the collaborative ALWABP with  $K_{\max} = 2$ . (Averaged over all instances).

	<b>Gap</b>	<b>% of collaborative solutions</b>	<b>Best improvement</b>	<b>Time (s)</b>
MaxTasks1 + MinD	53.80%	96.04%	46.77%	0.29
MaxTasks1.2 + MinD	41.84%	81.88%	50.00%	0.27
MaxTasks1.4 + MinD	29.79%	63.54%	46.77%	0.25
MaxTasks1.6 + MinR	23.18%	47.92%	50.00%	0.25
MaxTasks1.8 + MinR	21.31%	36.88%	50.00%	0.25
MaxTasks2 + MinR	19.96%	23.13%	50.00%	0.25
MaxTasks3 + MinR	20.78%	7.29%	50.00%	0.26
MaxTasksK + MinR	19.83%	16.04%	50.00%	0.25
MinRLB + MinR	16.79%	39.79%	46.77%	0.40

Table 10. Results for the heuristic resolution of the collaborative ALWABP with  $K_{\max} = 3$ . (Averaged over all instances).

	<b>Gap</b>	<b>% of collaborative solutions</b>	<b>Best improvement</b>	<b>Time (s)</b>
MaxTasks1 + MinD	74.12%	97.71%	46.77%	1.42
MaxTasks1.2 + MinD	64.81%	87.29%	50.00%	1.39
MaxTasks1.4 + MinD	49.36%	68.33%	46.77%	1.27
MaxTasks1.6 + MinD	40.19%	55.83%	45.45%	1.24
MaxTasks1.8 + MinR	32.81%	47.08%	45.45%	1.20
MaxTasks2 + MinR	26.09%	32.29%	50.00%	1.16
MaxTasks3 + MinR	21.62%	9.79%	50.00%	1.18
MaxTasksK + MinR	19.87%	16.04%	50.00%	1.13
MinRLB + MinR	16.68%	39.79%	46.77%	1.81

As for the parallel ALWABP, the criterion **MinRLB** presented the best results, yielding solutions with a gap of less than 17% with respect to the best known solutions. Solutions with efficiency improved in up to 50% were found by the method in very small computational times (around 1s).

### 5.3. Evaluation of the heuristic variants

The last computational experiments dealt with the version of the heuristics which looked only for solutions including workers in parallel or working collaboratively. The heuristics were implemented in C++ and executed using an AMD Athlon II Dual Core M320 processor with 3 GB of RAM. Tables 11 and 12 present these results. The heuristic has been run for all combinations of the work priority criteria and the task priority criteria (with the exception of the criteria based on positional weights, which were too time-consuming). The columns labeled ‘gap’ indicate the average gap of the best found solution for each instance (considering all criteria combinations) with respect to the best known serial solution. The columns labeled ‘Time’ indicate the average time (over all instances) needed to run all criterion combinations.

Table 21. Results for the modified heuristics for the parallel ALWABP. (Averaged over all instances).

	Gap	% feasible solutions found	Best improvement	Time(s)
Mod 1 ( $K_{max} = 2$ )	16.62%	100.00%	16.73%	44.82
Mod 2 ( $K_{max} = 2$ )	16.27%	100.00%	20.63%	81.00
Mod 1 ( $K_{max} = 3$ )	18.04%	97.97%	20.63%	100.98
Mod 2 ( $K_{max} = 3$ )	17.35%	100.00%	20.63%	191.70

Table 32. Results for the modified heuristics for the collaborative ALWABP. (Averaged over all instances).

	Gap	% of feasible solutions found	Best improvement	Time(s)
Mod 1 ( $K_{max} = 2$ )	16.28%	100.00%	50.00%	37.8
Mod 2 ( $K_{max} = 2$ )	11.25%	77.47%	50.00%	51.84
Mod 1 ( $K_{max} = 3$ )	17.22%	100.00%	47.73%	88.56
Mod 2 ( $K_{max} = 3$ )	11.84%	78.73%	50.00%	175.5

The results indicate that significant improvements could be obtained for some instances (up to 50% in throughput increase in some cases). Moreover, the algorithms were able to find solutions that ensured parallelism or collaboration with average throughputs that were not far from the best-known serial solutions, in reasonable computational times.

In a practical sense, the overall result of this computational study is that the introduction of slow/limited workers in conventional assembly lines can be done very efficiently through the use of the parallel/collaborative approach. This has important implications that will be exposed in the conclusions.

## 6. Conclusions and further research

We have introduced two variants of the Assembly Line Worker Assignment and Balancing problem that allow parallelization and collaboration between heterogeneous workers. From a theoretical point of view, the new variants define hard optimization problems for which no models or algorithms were available. We have proposed two linear formulations for the new problems and an adaptation of a constructive heuristic developed for simpler assembly line balancing problems. A set of computational tests has been used to evaluate the validity and applicability of the proposed models and algorithms.

The results have indicated that the flexibility introduced by these new extensions may significantly improve the efficiency of the line. Therefore, even if there is no particular reason for favouring parallel or collaborative layouts, they can be useful from the point of view of line efficiency. The results also showed that the proposed approaches can be especially beneficial when workers have very different task execution times, as it could be the case when one tries to integrate disabled workers in conventional assembly lines. In these cases, combining the parallel/collaborative approaches, they can be easily integrated in work teams together with experienced ordinary workers; so that they can feel more confident and progressively improve their performance with their advice. Therefore, our main conclusion is that companies can contribute to an important social aim, the socio-labour integration of people with disabilities, without important losses in productivity. This is even more important if we have in mind the global scenario where Corporate Social Responsibility implies the consideration of different stakeholders simultaneously.

About further research, several work lines may emerge from the studies discussed in this paper. For example, it would be interesting to consider how the different assignments obtained by the new approaches may help in obtaining effective job rotation schedules in assembly lines with heterogeneous workers. From an optimization point of view, the study of more elaborate metaheuristics such as grasp and random-key genetic algorithms may be of interest. These heuristics can clearly make use of the constructive approaches developed in this paper.

### **Acknowledgments**

This research was supported by CAPES, CNPq and FAPESP. This support is gratefully acknowledged.

### **References**

Akagi, F., Osaki, H., Kikuchi, S., 1983. A method for assembly line balancing with more than one worker in each station. *International Journal of Production Research* 21, 755-770.

Askin R.G., Zhou M., 1997. A parallel station heuristic for the mixed model production line balancing problem, *International Journal of Production Research* 35 (1997), pp. 3095-3105.

Bard J.F., 1989. Assembly line balancing with parallel workstations and dead time. *International Journal of Production Research* 27, 1005-1018.

Bartholdi, J., 1993. Balancing two-sided assembly lines: A case study. *International Journal of Production Research* 31, 2447-2461.

Baybars, I., 1986. A survey of exact algorithms for the simple assembly line balancing problem. *Management Science* 32, 909-932.

Becker C., Scholl A., 2009. Balancing assembly lines with variable parallel workplaces: Problem definition and effective solution procedure. *European Journal of Operational Research* 199, 359-374.

Blum C., Miralles C., 2011. On solving the assembly line worker assignment and balancing problem via beam search. *Computers & Operations Research* 38, 328-339.

Boysen N., Fliedner M., Scholl A., 2007. A classification of assembly line balancing problems. *European Journal of Operational Research* 183, 674-693



Boysen, N., Fliedner, M., 2006. A versatile algorithm for assembly line balancing, *European Journal of Operational Research*, 184, 39-56.

Boysen N., Fliedner M., Scholl A., 2008. Assembly line balancing: which model to use when? *International Journal of Production Economics* 111, 509-528.

Bukchin J., Rubinovitz, J., 2003. A weighted approach for assembly line design with station paralleling and equipment selection, *IIE Transactions* 35, pp. 73-85.

Bukchin, J., Dar-El, E.M., Rubinovitz, J., 1997. Team oriented assembly system design: a new approach. *International Journal of Production Economics* 51, 47-57.

Bukchin, J., Tzur, M., 2000. Design of flexible assembly line to minimize equipment cost. *IIE Transactions* 32, 585-598.

Chaves A.A.; Lorena L.A.N., Miralles C. 2007. Clustering search approach for the Assembly Line Worker Assignment and Balancing Problem. In Elwany MH, Eltawil AB, editors. *Proceedings of ICC&IE2007—37th International conference on computers and industrial engineering*, 1469-1478.

Chaves A.A.; Lorena L.A.N., Miralles C. 2009. Hybrid Metaheuristic for the Assembly Line Worker Assignment and Balancing Problem. *Lecture notes in Computer Science* 5818, 1-14

Costa A.M.; Miralles C. 2009. Job rotation in assembly lines employing disabled workers. *International Journal of Production Economics* 120, 625-632.

Dolgui, A., Guschinski, N., Levin, G., 1999. On problem of optimal design of transfer lines with parallel and sequential operations. In: Fuertes, J.M. (Ed.), *Proceedings of the Seventh IEEE International Conference on Emerging Technologies and Factory Automation*, Barcelona, Spain, vol. 1, 1999, pp. 329-334.

Dolgui, A., Guschinsky, N., Levin, G., 2006. A special case of transfer lines balancing by graph approach. *European Journal of Operational Research* 168, 732-746.

Ege, Y., Azizoglu, M., Ozdemirel, N. E., 2009. Assembly line balancing with station paralleling, *Computers & Industrial Engineering*, 57, 1218-1225.

Gökçen H., Agpak K., Benzer R., 2006. Balancing of parallel assembly lines. *International Journal of Production Economics* 103, 600-609.

Hoffmann, T.R., 1990. Assembly line balancing: A set of challenging problems. *International Journal of Production Research* 28, 1807-1815.

ILO – International Labour Office, 2007. Equality at work: Tackling the challenges. *International Labour Conference 96th Session 2007 Report I (B)*

ILO – International Labour Office, 2009. Facts on Disability and Decent Work. <http://www.ilo.org/disability>

Lee, T.O., Kim, Y., Kim, Y.K., 2001. Two-sided assembly line balancing to maximize work relatedness and slackness. *Computers and Industrial Engineering* 40, 273-2927.

McMullen P.R., Frazier G.U., 1998. Using simulated annealing to solve a multiobjective assembly line balancing problem with parallel workstations, *International Journal of Production Research* 36, pp. 2717-2741.

McMullen, P.R., Frazier, G.V., 1997. Heuristic for solving mixed-model line balancing problems with stochastic task durations and parallel stations. *International Journal of Production Economics* 51 (3), 177-190

Miralles C., Marin-Garcia J.A., Ferrus G., Costa A. M. 2010. OR/MS tools for integrating people with disabilities into employment. A study on Valencia's Sheltered Work centres for Disabled. *International Transactions In Operational Research* 17, 457-473.

Moreira, M. C. O., Costa, A. M., Chaves, A. A., Ritt, M. (2011). Simple heuristics for the Assembly Line and Worker Assignment Balancing Problem. Working paper.

Nicosia, G., D. Pacciarelli and A.Pacifici, 2002. Optimally balancing assembly lines with different workstations, *Discrete Applied Mathematics* 118, 99-113.

Pinnoi, A. and W.E. Wilhelm, 1997. A family of hierarchical models for assembly system design, *International Journal of Production Research* 35, 253-280.

Pinnoi, A., Wilhelm, W.E., 1998. Assembly system design: A branch and cut approach, *Management Science* 44,103-118.

Pinto, P.A., D.G. Dannenbring and B.M. Khumawala, 1983. Assembly line balancing with processing alternatives: an application, *Management Science* 29, 817-830.

Pinto, P.A., Dannenbring, D.G., Khumawala, B.M., 1981. Branch and bound and heuristic procedures for assembly line balancing with paralleling of stations. *International Journal of Production Research* 19, 565-576.

Rachamadugu, R., 1991. Assembly line design with incompatible task assignments. *Journal of Operations Management* 10, 469-487.

Rekiek, B., de Lit, P., Delchambre, A. 2002. Hybrid assembly line design and user's preferences, *International Journal of Production Research* 40, 1095-1111.

Rubinovitz, J., Bukchin, J., 1993. RALB — a heuristic algorithm for design and balancing of robotic assembly lines. *Annals of the CIRP* 42, 497-500.

Sarker, B.R., Shanthikumar, J.G., 1983. A generalized approach for serial or parallel line balancing. *International Journal of Production Research* 21, 109-133.

Scholl, A., Voß, S. (1996). Simple assembly line balancing - Heuristic approaches. *Journal of Heuristics*, 2, 217-244.

Scholl A., Boysen N., 2009. Designing parallel assembly lines with split workplaces: Model and optimization procedure. *International Journal of Production Economics* 119, 90-100.

Scholl, A. Becker, C. 2006. State-of-the-art exact and heuristic solution procedures for simple assembly line balancing, *European Journal of Operational Research* 168, 666-693.

Simaria, A. S., Vilarinho, P. M., 2004. A genetic algorithm based approach to the mixed-model assembly line balancing problem of type II, *Computers & Industrial Engineering* 47, 391-407.

Stephens D.A., Collins M.D., Dodder, R.A., 2005. A longitudinal study of employment and skill acquisition among individuals with developmental disabilities. *Research in Developmental Disabilities* 26 (2005) 469-486.

Süer, G. A., 1998. Designing parallel assembly lines, *Computers & industrial engineering*, 1998, 35, 467-470.

Vilarinho P.M., Simaria, A.S., 2002. A two-stage heuristic method for balancing mixed-model assembly lines with parallel workstations, *International Journal of Production Research* 40, pp. 1405-1420.

Vilarinho P.M., Simaria, A.S., 2006. ANTBAL: An ant colony optimization algorithm for balancing mixed-model assembly lines with parallel workstations, *International Journal of Production Research* 44, pp. 291-303

Wilhelm, W.E., 1999. A column-generation approach for the assembly system design problem with tool changes. *International Journal of Flexible Manufacturing Systems* 11, 177-205.

Wolfensberger, W. 1983. Social Role Valorization: a proposed new term for the principle of normalization, *Mental Retardation* 21, pp. 234-239.