



ELSEVIER

Contents lists available at ScienceDirect

Data in Brief

journal homepage: www.elsevier.com/locate/dib

Data Article

Application of a mathematical model for ergonomics in lean manufacturing



Lucia Botti*, Cristina Mora, Alberto Regattieri

Department of Industrial Engineering, University of Bologna, Viale del Risorgimento, 2, 40136 Bologna, Italy

ARTICLE INFO

Article history:

Received 10 May 2017

Received in revised form

12 June 2017

Accepted 28 June 2017

Available online 4 July 2017

ABSTRACT

The data presented in this article are related to the research article “Integrating ergonomics and lean manufacturing principles in a hybrid assembly line” (Botti et al., 2017) [1]. The results refer to the application of the mathematical model for the design of lean processes in hybrid assembly lines, meeting both the lean principles and the ergonomic requirements for safe assembly work. Data show that the success of a lean strategy is possible when ergonomics of workers is a parameter of the assembly process design.

© 2017 The Authors. Published by Elsevier Inc. This is an open access article under the CC BY license (<http://creativecommons.org/licenses/by/4.0/>).

Specifications Table

Subject area	<i>Industrial engineering</i>
More specific subject area	<i>Human factors, industrial and logistic system design</i>
Type of data	<i>Tables, Figures</i>
How data was acquired	<i>Input data acquired on field during the research project</i>

DOI of original article: <http://dx.doi.org/10.1016/j.cie.2017.05.011>

* Corresponding author.

E-mail addresses: lucia.botti5@unibo.it (L. Botti), cristina.mora@unibo.it (C. Mora), alberto.regattieri@unibo.it (A. Regattieri).

<http://dx.doi.org/10.1016/j.dib.2017.06.050>

2352-3409/© 2017 The Authors. Published by Elsevier Inc. This is an open access article under the CC BY license (<http://creativecommons.org/licenses/by/4.0/>).

Data format	<i>Raw and analyzed data</i>
Experimental factors	<i>Data acquired during the daily activity of assembly workers at the reference manufacturing company</i>
Experimental features	<i>Data acquired during the daily activity of assembly workers at the reference manufacturing company</i>
Data source location	<i>Not applicable for confidentiality reasons</i>
Data accessibility	<i>Data in this article and in the related research article [1].</i>

Value of the data

- The input data, i.e. parameter values, may be exported in order to be used by different mathematical models.
 - The output data may be used to define different decision functions for the choice of the optimal solution among the Pareto points.
 - The output data, e.g. variable values, may be exported in order to compare them with other results after the application of input data to different models.
-

1. Data

The model inputs refer to a manual assembly line with 6 manual workstations and 6 manual workers. A single worker is assigned to each manual workstation. The assembly task sequence is the same for each product type. Each task is standardizable and the assembly activities are not complex. Sensitive values of the manual assembly-process parameters are hidden, e.g. cycle times, takt times and batch sizes, for confidentiality reasons. The safety time varies from 1 to 3 h while the mean lateness of manual workstations varies from 2 to 12 s, depending on the product type and the task. The following Table 1 shows the other model parameters and the OCRA parameters for the ergonomic risk assessment through the OCRA method [2,3].

Particularly, the values of the technical actions refer to the most stressed arm, for each worker. The work shift is of 8 h. A lunch break and two breaks of 10 min each are distributed among the 8-h shift. Job rotations are not allowed during the work shift and each worker performs the same single task for the whole 8 h. As a consequence, repetitive manual tasks last for a relevant part of the shift. The OCRA indices in Table 2 define the workers exposure to repetitive movements of the upper limbs.

2. Experimental design, materials and methods

The introduced data define the model inputs for the considered case study. 48 binary variables are introduced subjected to 60 feasibility constraints. The model and the input data are coded in AMPL language and processed adopting Gurobi Optimizer© v.5.5 solver. An Intel® Core™ i7-4770 CPU @ 3.50 GHz and 32.0GB RAM workstation is used. The average solving time is approximately of 0.5 s.

The Normalized Pareto frontier in Fig. 1 shows the trends of the two objective functions in the normalized *WIP-Cost* diagram [4]. Particularly, the points from *W* to *C* are the Pareto points composing the normalized Pareto frontier (Fig. 1). Each Pareto point represents an effective non-dominated trade-off assembly layout configuration.

Table 1
Parameters of the mathematical model.

<i>t</i>	1	2	3	4	5	6
<i>i</i> [machines]	1	1	1	1	1	1
<i>l_{max,t}</i> [workers]	1	1	1	1	1	1
<i>o_t</i> [%]	2	2	1	2	1	2
<i>o'_t</i> [%]	4	5	3	5	3	5
<i>q_t</i> [€/h]	2.80	2.40	1.20	1.20	1.20	2.40
<i>r_t</i> [€/machine and hour]	100	100	100	100	100	100
<i>x_t</i> [€/h and machine]	56	40.88	32.1	47.44	32.1	40.88
<i>y_t</i> [€/h and worker]	15	15	15	15	15	15
<i>OCRA parameters</i>						
<i>n_{TC,t}</i> Product 1	14	8	5	4	3	12
<i>n_{TC,t}</i> Product 2	15	8	5	5	3	12
<i>n_{TC,t}</i> Product 3	14	8	5	4	3	12
<i>n_{TC,t}</i> Product 4	15	10	5	5	3	12
<i>k_f</i> Product from 1 to 4	30	30	30	30	30	30
<i>F_{M,t}</i> Product from 1 to 4	0.65	0.35	1.00	0.85	1.00	0.20
<i>P_{M,t}</i> Product from 1 to 4	0.60	0.60	1.00	0.60	1.00	0.60
<i>Re_{M,t}</i> Product from 1 to 4	1.00	0.70	1.00	0.70	1.00	0.70
<i>A_{M,t}</i> Product from 1 to 4	1.00	0.90	1.00	0.95	1.00	0.80
<i>Rc_M</i>	0.60	0.60	0.60	0.60	0.60	0.60
<i>t_M</i>	1.00	1.00	1.00	1.00	1.00	1.00

Table 2
OCRA index for each worker.

Worker	Task	OCRA index
Worker 1	1	3.4
Worker 2	2	1.3
Worker 3	3	0.7
Worker 4	4	1.5
Worker 5	5	0.6
Worker 6	6	3.7

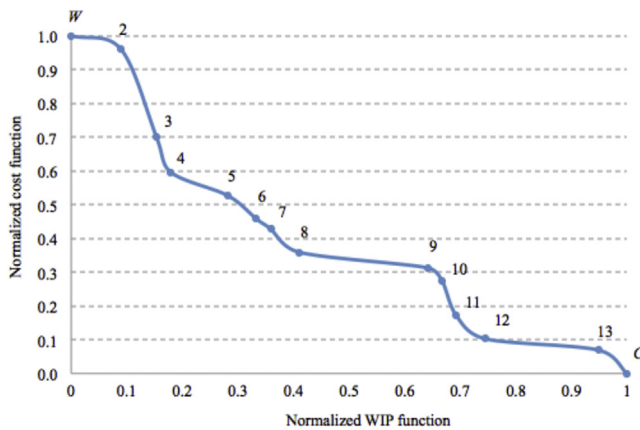


Fig. 1. Normalized Pareto frontier [1].

Table 3
Value of the normalized functions for each Pareto point.

Pareto point	Normalized WIP function	Normalized cost function
W	0.00	1.00
2	0.09	0.96
3	0.15	0.70
4	0.18	0.60
5	0.28	0.53
6	0.33	0.46
7	0.36	0.43
8	0.41	0.36
9	0.64	0.31
10	0.67	0.28
11	0.69	0.17
12	0.74	0.10
13	0.95	0.07
C	1.00	0.00

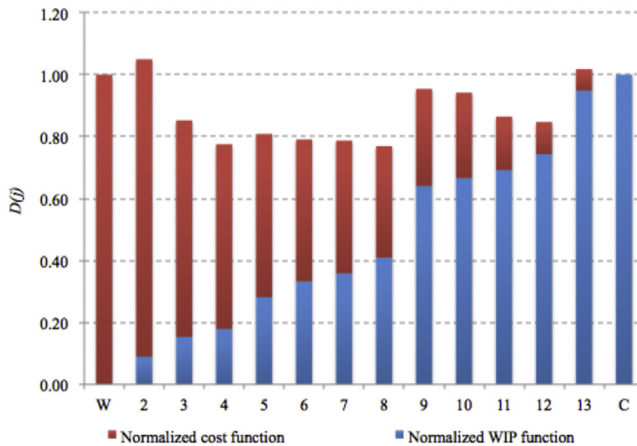


Fig. 2. Values of decision function $D(j)$ for each Pareto point [1].

Table 4
Decision function value for each Pareto point.

j	W	2	3	4	5	6	7	8	9	10	11	12	13	C
$D(j)$	1.00	1.05	0.85	0.78	0.81	0.79	0.79	0.77	0.95	0.94	0.86	0.85	1.02	1.00

The following Table 3 shows the coordinates of each Pareto point.

The following Fig. 2 and Table 3 show the values of decision function $D(j)$ for each Pareto point Table 4.

The solution in point $j=8$ minimises the decision function $D(j)$. The following Fig. 3 shows the assembly layouts for solutions in points W, C and $j=8$.

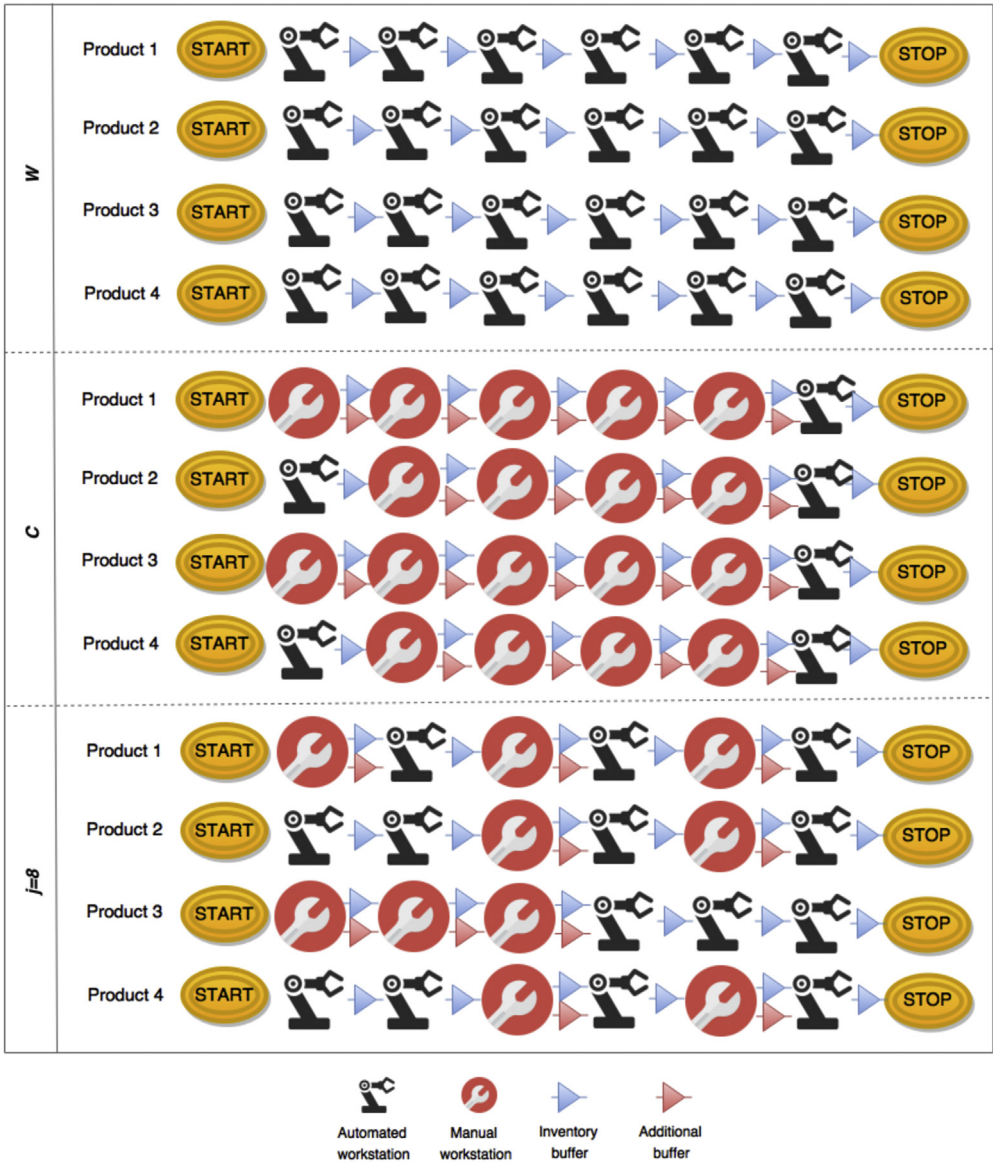


Fig. 3. Assembly layouts for solutions in points W, C and $j=8$ [1].

Acknowledgements

The research is funded by Azienda Unità Sanitaria Locale (AUSL) of Bologna. (Fondi Sanzioni D.Lgs. 758/2011). The authors are grateful for this support.

Transparency document. Supplementary material

Transparency data associated with this article can be found in the online version at <http://dx.doi.org/10.1016/j.dib.2017.06.050>.

References

- [1] L. Botti, C. Mora, A. Regattieri, Integrating ergonomics and lean manufacturing principles in a hybrid assembly line, *Comput. Ind. Eng.* (2017).
- [2] E. Occhipinti, OCRA: a concise index for the assessment of exposure to repetitive movements of the upper limbs, *Ergonomics* 41 (9) (1998) 1290–1311.
- [3] I.S.O. 11228-3, System of Standards for Labor Safety, Ergonomics. Manual Handling. Part 3. Handling of Low Loads at High Frequency, 2007.
- [4] A. Messac, A.I. Yahaya, C.A. Mattson, The normalized normal constraint method for generating the Pareto frontier, *Struct. Multidiscip. Optim.* 25 (2) (2003) 86–98.