



Received: 22 March 2017
Accepted: 23 May 2017
First Published: 02 June 2017

*Corresponding author: M.K. Adeyeri,
Department of Mechanical and
Production Engineering, The Federal
University of Technology, Akure, Nigeria
E-mail: adeyerimichaeltut@gmail.com

Reviewing editor:
Jun Guo, Wuhan University of
Technology, China

Additional information is available at
the end of the article

PRODUCTION & MANUFACTURING | RESEARCH ARTICLE

Development of dynamic layout model for pondo yam flour processing plant

S.P. Ayodeji¹, M.K. Adeyeri^{1*} and A. Ogunsua¹

Abstract: A documented pondo yam process plant that produces 12.25 kg of pondo yam within 315.04 min was studied so as to address its associated problems. It was found that its production capacity is limited due to process time delay of 77 and 216 min experienced at the parboiling and drying section respectively. This informed the need to design a dynamic facility layout to increase production capacity, introduce flexibility of material flow through the process plant and manage effectively material flow through the process plant. In order to achieve this, a software called *pundosim* was developed using necessary design model equations and data. This will assist in forecasting material flow above the static reference frame of the production facility, and iteratively generate the optimal machine configuration within cost constraint, thereby preventing unending increase of service points at the regions of bottlenecks and ensuring economic viability of the entire processing plant. The successful deployment of the developed *pundosim* using input of values within the range of 12.25–134.75 kg at an increment of 12.25 kg shows a considerable decrease in the total process time at optimal machine configuration compared with

ABOUT THE AUTHORS

Our research team is involved in all forms of need-driven manufacturing activities ranging from agrarian products processing to human mobility support systems. Our research aims to improve manufacturing systems or products and processes via the use of innovative technology or concept. The team incorporates automated and sophisticated computerized design and operational systems in its manufacturing works. Similarly, our core research focus also entails the area of human performance enhancement in both cognitive and physical domains. The team employs classical and experimental ergonomics, cognitive and physiological Engineering, and application of new technologies in achieving defined manufacturing goals.

The group's aims and expertise are hinged on wide spectrum of manufacturing research. The team has key expertise in the following areas: design and fabrication of process plants; intelligent automation and assembly; metal forming; welding technologies; applied ergonomics; modelling and systems design; industrial systems maintenance and sustainability technologies; and production and manufacturing system.

PUBLIC INTEREST STATEMENT

Based on some generic model for dynamic layout facility of pondo yam flour processing plant, a software called *pundosim* was developed to sort out necessary design equations and data of an existing static layout to increase the production volume at optimal production time. The existing model and design equations do address the limitations in the static layout of the pondo yam flour processing plant by increasing the throughput capacity, permitting flexibility of material flow and effective utilization of resources. The developed software does prove useful for processing plant with similar attributes to pondo yam flour processing plant because it takes the machine configuration matrix and the economic implication of production capacity into context. The package developed can assist and guide prospective investors in knowing the amount of capital needed to invest with and what would be the profit outcome at the end of any targeted period of time under a *ceteris paribus* condition.

the initial reference machine configuration. Thus, the developed model does prove useful for process plant which has similar attributes with poudo yam flour processing plant as it takes machine configuration matrix and the economic implication of production capacity into consideration.

Subjects: Industrial Engineering & Manufacturing; Manufacturing Engineering; Engineering Management; Production Engineering

Keywords: poudosim; poudo yam; material handling; model equations; dynamic facility layout

1. Introduction

Facility layout entails the optimum arrangement and allocation of available resources such as space, personnel, storage space etc. to facilitate effectiveness in production process (Subodh & Kuber, 2014). A properly arranged facility improves operating efficiency which has justified the need to formulate facility layout problem with the intent to solve industrial challenges. However, the problems to be solved are diverse in works of literature ranging from material handling problems, space allocation of machines and facilities within plant site etc. With dynamic/static layout problem being the most encountered problem (Amine, Henri, & Sonia, 2007). Dynamic layout is designed to respond to series of changes often dictated by market and product demand; consequently making the layout relevant over periods demarcated in weeks, months or years (Alan & Shang, 2004; Amine et al., 2007). The essence of dynamic facility layout as opposed to static facility layout is to replace the cost of rearrangement with the cost of reconstruction which is inherent in static facility layout. Dynamic facility layout is a layout that permits the rearrangement of production facility to respond to product changes and product mix; dynamic facility layout can be attributed to static facility layout rearrangement to respond market changes demarcated over periods which could be in weeks, months or years (Alan & Shang, 2004; Amine et al., 2007; Balakrishnan, Cheng, Conway, & Lau, 2003; Kusiak & Heragu, 1987). Static layout problem which is also facility layout problem is formulated as Quadratic Algorithm Problem (Koopmans & Beckmann, 1957); the changes brought about by flow through the facility layout permits the static layout problem to be modeled as dynamic layout problem. Factors associated with the changes in flow are product design change, product addition or removal, production equipment replacement, short product life cycle, production volume change and associated changes in schedule (Heragu, 1992). Metaheuristics approach such as genetic algorithm, tabu search, simulated annealing and other hybrid metaheuristic approach are the solution methods employed in solving dynamic layout problem as evident in the works of Balakrishnan and Cheng (2000), Baykasoğlu and Gindy (2001), Bos (1993), Kaku and Mazzola (1997), McKendall and Shang (2006a, 2006b), Wu, Chung, and Chang (2008) and Balakrishnan et al. (2003). The poudo yam flour processing plant layout is a static layout constituting the yam selection and weighing, washing, peeling and slicing, parboiling, drying, sieving and packaging sections and having the capacity to process 50.25 kg of yam tubers to 12.25 kg of poudo yam per throughput production (Ayodeji, Olabanji, & Adeyeri, 2012) and subsequent automation of the process plant (Ayodeji, Khubulani, Oduetse, & Mohammed, 2015) which reduced the processing time from 6 h 12 min to 5 h 19 min. Prior to the dynamic layout model for poudo yam flour processing plant (Ogunsua, 2016), the existing static layout does not permit flexible rearrangement of machines, hence modelling the system as a quadratic algorithm problem which is a complex combinatorial problem does not suffice. The modelling of dynamic facility layout to; introduce some measure of flexibility at the regions of bottleneck, ensure effectiveness in the material handling, increase the per throughput production at optimal production time; and some sets of design equations to optimally populate the dynamic layout model at optimal production time consequently ensuring economic viability of the entire process plant by preventing redundancy at the region of bottleneck in the system (the parboiling and drying sections) constitute the latest improvements on the processing plant (Ogunsua, 2016). In view to sequentially arrange and implement the design equations to optimally populate the dynamic layout model for the existing processing plant, a dynamic programming software called *Poudosim* was developed with the intent to the sort out necessary design equations and data, to forecast material flow above the static

Table 1. Machine process time at reference machine configuration

Machines	Average wt. before processing (kg)	Average weight after processing (kg)	Rate of machine	Time (min)
Washing	$W_1 = 4.102 W_0$	$W_2 = 95.52\% W_1$	3.891 kg/min	$T_1 = \frac{W_1}{3.891}$
Peeling and slicing	W_2	$W_3 = 95.31\% W_2$	12.28 kg/min	$T_2 = \frac{W_2}{12.28}$
Parboiling	W_3	$W_4 = 1.0197 W_3$	45.75 [77 min]	$T_3 = b * 77$
Conveyor	W_4	$W_5 = W_4$	31.8 kg/min	$T_4 = \frac{W_4}{31.8}$
Drying	W_5	$W_6 = 26.36\% W_5$	0.16 kg/min	$T_5 = \frac{W_5}{0.16}$
Milling	W_6	$W_7 = 99.84\% W_6$	3.72 kg/min	$T_6 = \frac{W_6}{3.72}$
Sieving	W_7	$W_0 = 99.76\% W_6$	22.8 kg/min	$T_7 = \frac{W_7}{22.8}$

The total time taken from washing to sieving (T) = $T_1 + T_2 + T_3 + T_4 + T_5 + T_6 + T_7$

Note: b is the number of throughput through the process plant.

Table 2. Machine process time at optimal machine configuration

Machines	Average weight before processing (kg)	Average weight after processing (kg)	Time (min)
Washing	$W_1 = 4.102W_0$	$W_2 = 95.52\% W_1$	$T_1 = \frac{T_1}{N_1}$
Peeling and slicing	W_2	$W_3 = 95.31\% W_2$	$T_2 = \frac{T_2}{N_2}$
Parboiling	W_3	$W_4 = 1.0197 W_3$	$T_3 = \frac{T_3}{N_3}$
Conveyor I	W_4	$W_5 = W_4$	$D_1 = \frac{T_4 \times [3000 + 833.33(N_3 - 1)]}{3000}$
Drying	W_5	$W_6 = 26.36\% W_5$	$T_5 = \frac{T_5}{N_5}$
Conveyor II	W_6	W_6	$D_2 = \frac{T_4 \times 483.33(N_5 - 1)}{3000}$
Milling	W_6	$W_7 = 99.84\% W_6$	$T_6 = \frac{T_6}{N_6}$
Sieving	W_7	$W_0 = 99.76\% W_7$	$T_7 = \frac{T_7}{N_7}$

Total time taken from washing to sieving $Z = \frac{T_1}{N_1} + \frac{T_2}{N_2} + \frac{T_3}{N_3} + D_1 + \frac{T_5}{N_5} + D_2 + \frac{T_6}{N_6} + \frac{T_7}{N_7}$

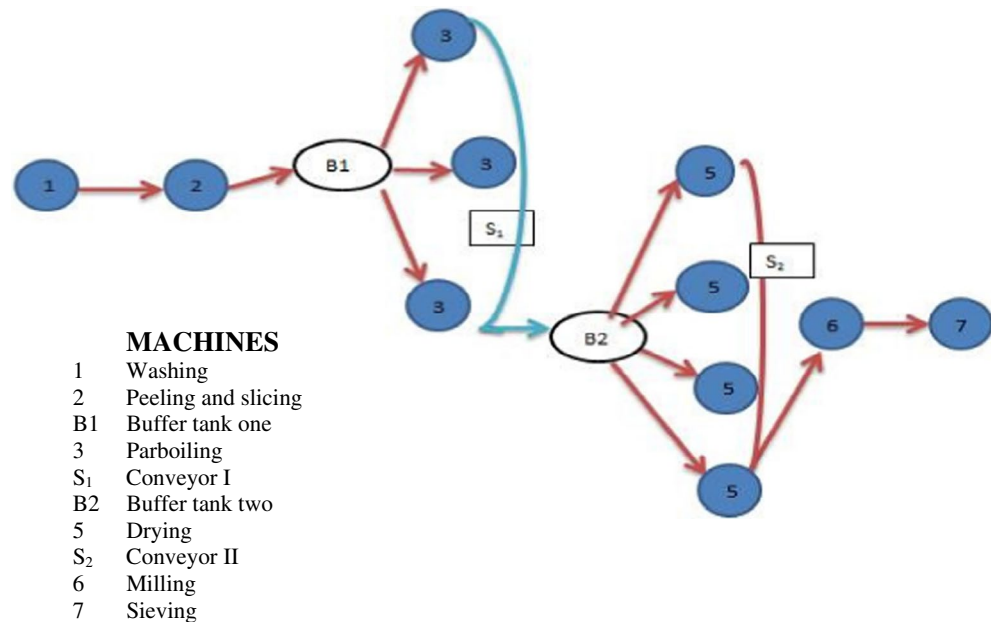
reference frame of the production facility, and iteratively generate the optimal machine configuration within cost constraint consequently preventing unending increase of service points at the regions of bottlenecks and ensuring economic viability of the entire processing plant.

2. Methodology

2.1. Dynamic programming design

Using Equations (1)–(9) and model formulation in Tables 1 and 2 by Ogunsua (2016), flowcharts were designed to generate the material flow through the process plant, generate the optimal machine configuration and forecast process time at reference and optimal machine configurations as shown in Figures 2 and 3. The flowcharts were used to create software (*Poundosim*) in java programming language. The Governing Design Equation constitute following in view to populate the dynamic facility model as shown in Figure 1: the material handling equation which extends the length of conveyor I and the introduced conveyor II as stated in Equations (1) and (2); the objective function equation which forecast the process time at each of the units in the process plant as stated in Equations (3)–(5); the cost constraint equations which prevents unending increase of the service

Figure 1. Improved dynamic facility layout model.



points at the region of bottlenecks specifically at the parboiling and drying section as shown in Equations (6)–(8); and the weight constraint equation which prevents redundancy in the system specifically at the parboiling section as shown in Equation (9).

$$S_1 = 3000 + I_3(N_3 - 1) \quad (1)$$

$$S_2 = I_5(N_5 - 1) \quad (2)$$

S_1 represents the length of conveyor I; S_2 represents the length of conveyor II; L_3 represents the extra length of the conveyor I per number of parboiler (833.33 mm/parboiler) when $N_3 > 1$ (Ajamu, 2014); L_5 represents the introduced length of the conveyor II per number of dryers (483.33 mm/dryer) when $N_5 > 1$ (Afolami, 2014).

$$T = \frac{T_1}{N_1} + \frac{T_2}{N_2} + \frac{T_3}{N_3} + D_1 + \frac{T_5}{N_5} + D_2 + \frac{T_6}{N_6} + \frac{T_7}{N_7} \quad (3)$$

$$D_1 = \frac{T_4 \times [3000 + 883.33](N_3 - 1)}{3000} \quad (4)$$

$$D_2 = \frac{T_4 \times 483.33(N_5 - 1)}{3000} \quad (5)$$

where T represents the total process time per throughput.

T_1, T_2, T_3, T_5, T_6 and T_7 represents the process time of the washing, peeling and slicing, parboiling, drying, milling and sieving machines respectively.

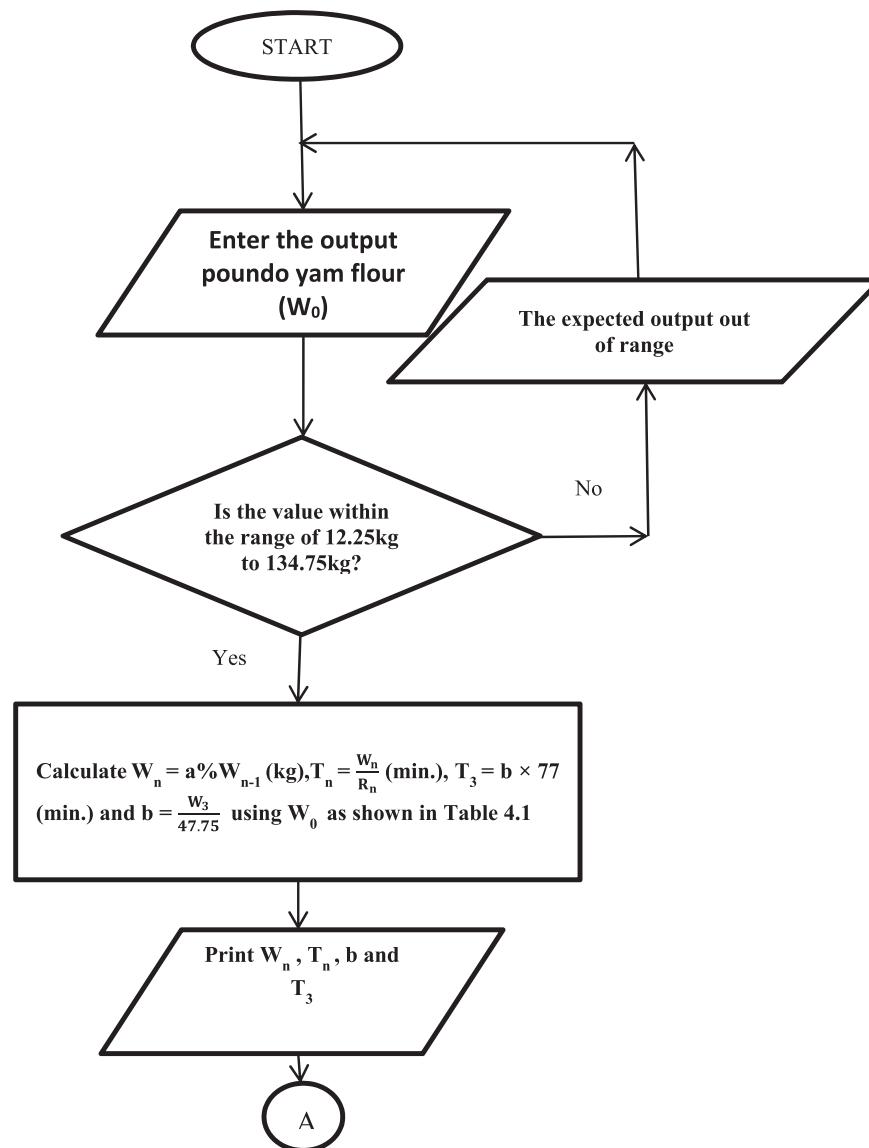
N_1, N_2, N_3, N_5, N_6 and N_7 represents the number of washing, peeling and slicing, parboiling, and drying.

D_1 represents the time of material transport through conveyor I at the parboiler.

D_2 represents the time of material transport through conveyor II at the dryer.

$$C_1N_1 + C_2N_2 + C_3N_3 + C_4 + AL_3(N_3 - 1) + C_5N_5 + AL_5(N_5 - 1) + C_6N_6 + C_7N_7 \leq W_0 \times \text{Market cost} \quad (6)$$

Figure 2. Reference machine configuration flowchart.



$$A = C_4/L_4 \quad (7)$$

where $C_1, C_2, C_3, C_4, C_5, C_6$ and C_7 represent the equivalent cost per throughput of the washing, peeling and slicing, parboiling, conveyor, drying, milling and sieving machines respectively.

N_1, N_2, N_3, N_5, N_6 and N_7 is the number of washing, peeling and slicing, parboiling, conveyor, drying, milling and sieving machine respectively.

W_0 represents the weight of pouno yam flour produced per throughput = 12.25 kg/throughput (Ayodeji et al., 2015).

L_3 is the extra length of conveyor per parboiler to convey materials from the parboiling machine when $N_3 > 1$.

L_s is the extra length of conveyor per dryer to convey dried chips of yam when the number of drying machine $N_s > 1$.

$$C_n = \frac{T_n}{T} \left(YC + LC + \frac{K}{N_i} \right) + ec_n + \frac{FC_n}{N_i} \quad (8)$$

$N_i = 648$ throughputs to breakeven (Ogunsua, 2016).

Where C_n is the cost constraint of machine- n .

T_n is the process time per throughput of machine- n .

T is the total process time per throughput of the entire process plant.

FC_n is the fixed cost of machine- n taking inflation rate into consideration from 2014 to 2016.

K is the constant fixed cost of the control system and the contingencies.

$$N_3 \leq b$$

$$b = \frac{W_3}{47.75} \text{ (approximated to the nearest whole number)} \quad (9)$$

where N_3 is the number of parboiling machine in the process plant.

W_3 is the weight input into the machine at location 3 in the process plant (Parboiling machine).

b is the number of throughput approximated to the nearest whole number.

3. Result and discussion

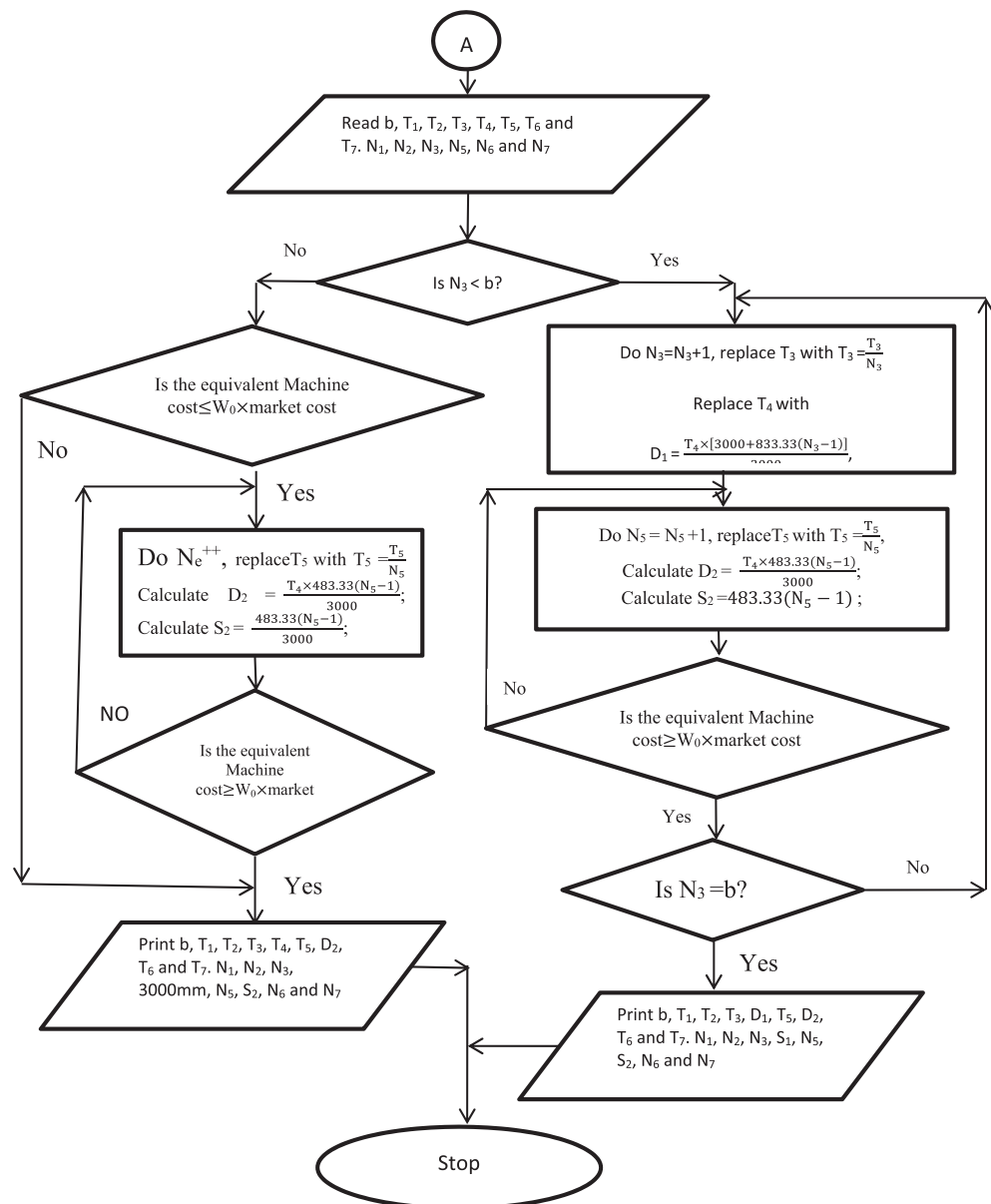
3.1. Reference machine configuration flowchart

The Reference Machine Configuration Flowchart as shown in Figure 2 takes the input (W_o) which is expected pondo yam flour to be produced from the process plant. It uses the formulated model equations in Table 1 to estimate the weight of material flow through the machines in the process plant and forecast the process time at reference machine configuration.

3.2. Optimal machine configuration flowchart

The Optimal Machine Configuration Flowchart generates the optimal machine configuration for a specified throughput capacity. It increases the service points at the region of bottlenecks; the par-boiling and the drying section. The product of the market cost per throughput production and the specified weight (W_o) of pondo yam flour expected from the process plant is set as a constraint for the machine cost contribution as shown in Figure 3 for the drying machine while the weight is set as constraint for the number of parboiling machine using the number of throughput (b) as stated in Equation (9).

Figure 3. Optimal machine configuration flowchart.



3.2.1. Equivalent machine cost contribution

The Machine Cost Contribution per throughput is as stated in Equation (10) using Equations (6)–(8). The detailed estimate is as shown in Tables 3 and 4 (Ogunsua, 2016).

$$426.57N_1 + 201.49N_2 + 1814.65N_3 + 4668.85N_5 + 119.31N_6 + 66.55N_7 + 46.65 \leq W_0 \times \text{Market cost} \quad (10)$$

3.3. Dynamic programming result

The software developed (*Poundosim*) was tested with values within the range of 12.25–134.75 kg of poundo yam flour per throughput capacity at an increment of 12.25 kg. The machine processing time for each of the machines in the process plant as generated by the software at specified ranges of input at reference and optimal machine configurations are as shown Table 5. At reference machine configuration when number of parboiling (N_3) and drying machine (N_5) is one as shown in Table 5, the

Table 3. Machine energy cost per throughput

S/N	Machine	Powered component	Power (kW)	Total power (kW)	Time per throughput (h)	Total energy (kWh)	Cost (₦) (ec_n)
1	Washing	Pump	3.2309	3.2309	0.215	0.6946	29.17
2	Peeling and slicing	Peeling shaft	3.2800	3.2800	0.055	0.1804	7.56
3	Parboiling	Shaft	0.9321	2.4321	1.283	3.1204	131.06
		Heater	1.5000				
4	Conveyor	Drive shaft	2.2300	2.2300	0.025	0.0500	2.10
5	Drying	Heater	3.000	3.4321	3.600	12.3557	518.94
		Shaft	0.4321				
6	Milling	Shaft	0.9694	0.9694	0.055	0.0533	2.24
7	Sieving	Vibrator	0.5966	0.5966	0.009	0.0054	0.23

Table 4. Equivalent machine cost contribution per throughput (C_n)

Machine	Fixed cost per throughput $\frac{FC_n}{N_i}$ (₦)	Energy cost per throughput (ec_n) (₦)	$\frac{T_n}{T} \left(YC + LC + \frac{K}{N_i} \right)$ (₦)	Equivalent cost per throughput (C_n) (₦)
Washing	155.07	29.17	242.33	426.57
Peeling and slicing	123.00	7.56	70.93	201.49
Parboiling	217.34	131.06	1,442.17	1,790.57
Conveyor	56.30	2.10	28.30	86.70
Drying	93.49	514.94	4,044.65	4,654.08
Milling	55.16	2.24	61.91	119.31
Sieving	56.30	0.23	10.02	66.55

length of the conveyor I (S_1) and conveyor II (S_2) remains at 3,000 and 0 mm respectively in line with the process plant design (Ayodeji et al., 2012); an indication that while the parboiling and drying machine is one, there is no need for the extension of conveyor I length and the introduction of conveyor II. However, with the increment in the number of parboiling and drying machine, the length of conveyor I increase to accommodate the increasing number of parboiling machine while there was introduction of conveyor II and an increase in the length of the conveyor with increasing number of drying machines. There was also a significant reduction in the process time when the total process time at optimal machine configuration was compared with the total process time at reference machine configuration for specified throughput capacity as shown in Table 5.

Where $T_1, T_2, T_3, D_1, T_5, T_6, D_2$ and T_7 represents the process time of the washing, peeling and slicing, parboiling, conveyor 1, drying, milling, introduced conveyor 2 and sieving machines respectively.

N_1, N_2, N_3, N_5, N_6 and N_7 represents the number of washing, peeling and slicing, parboiling, drying.

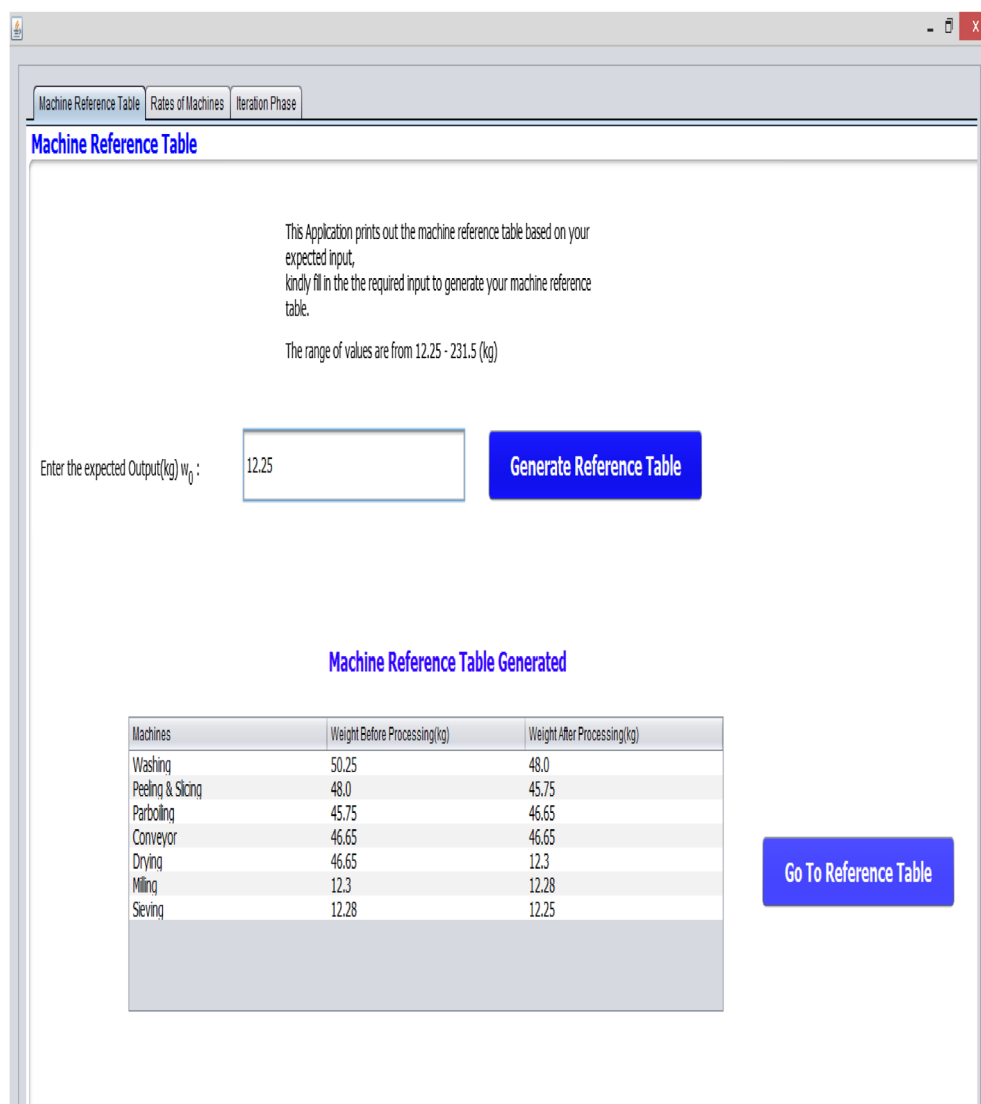
S_1 and S_2 represents the length of conveyor I at the parboiling machine and introduced conveyor II at the drying machine.

Table 5. Reference and optimal machine configuration and corresponding machine process time

			Reference machine configuration								Optimal machine configuration										
S/N	Input (kg)	Out-put (kg)	N ₁	N ₂	N ₃	S ₁ (mm)	N ₅	S ₂ (mm)	N ₆	N ₇	Total time (min)	N ₁	N ₂	N ₃	S ₁	N ₅	S ₂	N ₆	N ₇	Total time (min)	%Time reduction
1	50.25	12.25	1	1	1	3,000	1	0	1	1	314.81	1	1	1	3,000	2	483.33	1	1	207.07	34.22381754
2	100.5	24.5	1	1	1	3,000	1	0	1	1	630.26	1	1	2	3,833.33	3	966.66	1	1	267.09	57.62225113
3	150.75	36.75	1	1	1	3,000	1	0	1	1	945.4	1	1	3	4,666.66	4	1,449.99	1	1	310.1	67.19906918
4	201	49	1	1	1	3,000	1	0	1	1	1,260.57	1	1	4	5,499.99	5	1,933.32	1	1	347.23	72.45452454
5	251.25	61.25	1	1	1	3,000	1	0	1	1	1,575.64	1	1	5	6,333.32	6	2,416.65	1	1	382.04	75.75334467
6	301.5	73.5	1	1	1	3,000	1	0	1	1	1,890.77	1	1	6	7,166.65	7	2,899.98	1	1	416.12	77.99203499
7	351.75	85.75	1	1	1	3,000	1	0	1	1	2,205.89	1	1	7	7,999.98	9	3,866.64	1	1	430.88	80.46684105
8	402	98	1	1	1	3,000	1	0	1	1	2,521.06	1	1	8	8,833.31	10	4,349.97	1	1	467.56	81.45383291

			Reference machine configuration process time								Optimal machine configuration process time										
S/N	Input (kg)	Out-put (kg)	T ₁	T ₂	T ₃	D ₁	T ₅	D ₂	T ₆	T ₇	Total time (min)	T ₁	T ₂	T ₃	D ₁	T ₅	D ₂	T ₆	T ₇	Total time (min)	%Time reduction
1	50.25	12.25	12.59	3.91	77	1.49	215.97	0	3.31	0.54	314.81	12.59	3.91	77	1.49	107.99	0.24	3.31	0.54	207.07	34.22381754
2	100.5	24.5	25.83	7.82	154	2.98	431.94	0	6.61	1.08	630.26	25.83	7.82	77	3.81	143.98	0.96	6.61	1.08	267.09	57.62225113
3	150.75	36.75	38.74	11.73	231	4.47	647.92	0	9.92	1.62	945.4	38.74	11.73	77	6.95	161.98	2.16	9.92	1.62	310.1	67.19906918
4	201	49	51.66	15.64	308	5.96	863.94	0	13.22	2.15	1,260.57	51.66	15.64	77	10.93	172.79	3.84	13.22	2.15	347.23	72.45452454
5	251.25	61.25	64.57	19.54	385	7.45	1,079.86	0	16.53	2.69	1,575.64	64.57	19.54	77	15.73	179.98	6	16.53	2.69	382.04	75.75334467
6	301.5	73.5	77.49	23.45	462	8.94	1,295.83	0	19.83	3.23	1,890.77	77.49	23.45	77	21.36	185.12	8.64	19.83	3.23	416.12	77.99203499
7	351.75	85.75	90.4	27.34	539	10.43	1,511.81	0	23.14	3.77	2,205.89	90.4	27.34	77	27.81	167.98	13.44	23.14	3.77	430.88	80.46684105
8	402	98	103.32	31.27	616	11.93	1,727.78	0	26.45	4.31	2,521.06	103.32	31.27	77	35.13	172.78	17.3	26.45	4.31	467.56	81.45383291

Plate 1. Poundsim interface showing material flow through the process plant.



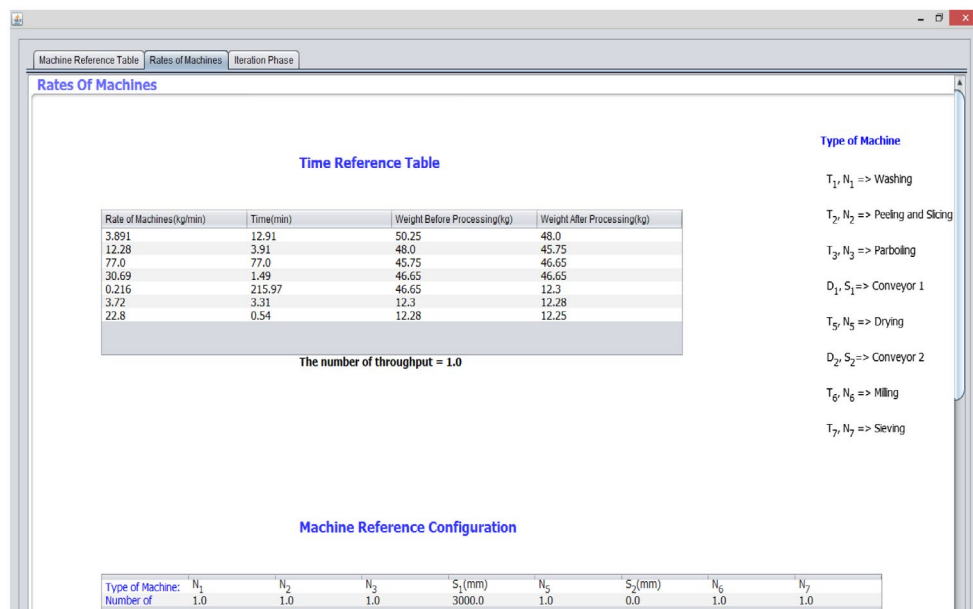
The screenshot shows the 'Machine Reference Table' tab of the Poundsim software. It contains instructions for generating a reference table based on an expected output. A text box for 'Enter the expected Output(kg) w_0 :' contains the value '12.25'. A blue button labeled 'Generate Reference Table' is next to it. Below this, a section titled 'Machine Reference Table Generated' displays a table with three columns: 'Machines', 'Weight Before Processing(kg)', and 'Weight After Processing(kg)'. The table lists seven machines: Washing, Peeling & Slicing, Parboiling, Conveyor, Drying, Milling, and Sieving, with their respective weights before and after processing. A blue button labeled 'Go To Reference Table' is positioned to the right of the table.

Machines	Weight Before Processing(kg)	Weight After Processing(kg)
Washing	50.25	48.0
Peeling & Slicing	48.0	45.75
Parboiling	45.75	46.65
Conveyor	46.65	46.65
Drying	46.65	12.3
Milling	12.3	12.28
Sieving	12.28	12.25

3.4. The Poundsim software

The interface displaying the reference machine configuration and process time, optimal machine configuration and process time result for the developed software is as shown in Plates 1, 2 and 3. In line with the design of the flowcharts in Figure 2, the software generates the material flow through the process plant using model formulation Table 1. Upon the input of the expected poundo yam flour output (W_0) within the range 12.25–134.75 kg as shown in Plate 1, the Generate Reference Table button generates the material flow through the process plant; subsequently, the Go To Reference Table button switches the top menu from Machine Reference Table icon to the Rate of Machines icon in which the Machine Reference Table displays the material flow through the process plant, the machine ratings and the reference machine configuration and corresponding process time is as shown in Plate 2. On the Machine reference table icon, the Go To Iteration button executes the

Plate 2. Poundsim software interface showing (a) machine reference table and (b) machine reference configuration and corresponding machine process time.



Rates Of Machines

Time Reference Table

Rate of Machines(kg/min)	Time(min)	Weight Before Processing(kg)	Weight After Processing(kg)
3.891	12.91	50.25	48.0
12.28	3.91	48.0	45.75
77.0	77.0	45.75	46.65
30.69	1.49	46.65	12.3
0.216	215.97	12.3	12.28
3.72	3.31	12.28	12.25
22.8	0.54		

The number of throughput = 1.0

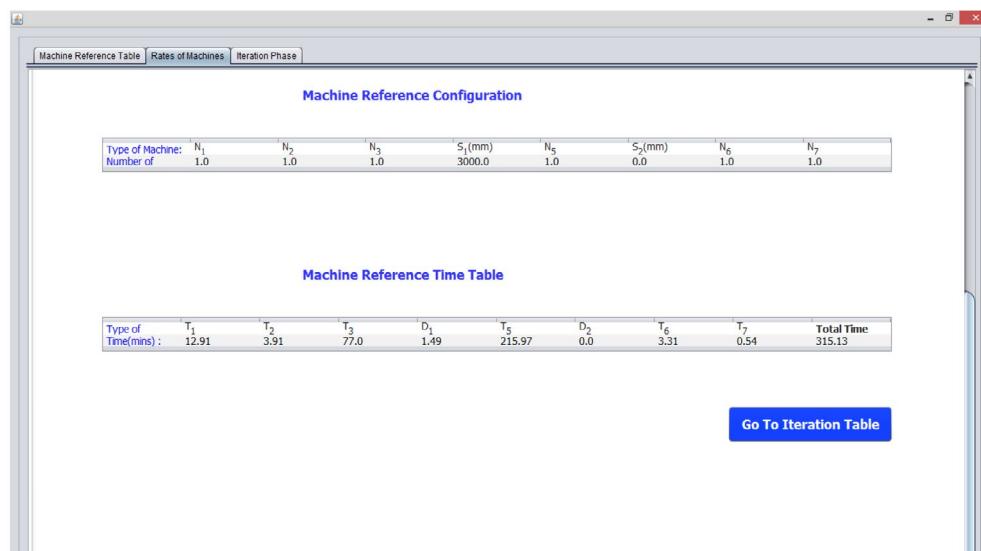
Machine Reference Configuration

Type of Machine:	N ₁	N ₂	N ₃	S ₁ (mm)	N ₅	S ₂ (mm)	N ₆	N ₇
Number of	1.0	1.0	1.0	3000.0	1.0	0.0	1.0	1.0

Type of Machine

- T₁, N₁ ==> Washing
- T₂, N₂ ==> Peeling and Sking
- T₃, N₃ ==> Parboiling
- D₁, S₁ ==> Conveyor 1
- T₅, N₅ ==> Drying
- D₂, S₂ ==> Conveyor 2
- T₆, N₆ ==> Milling
- T₇, N₇ ==> Sieving

(a)



Machine Reference Configuration

Type of Machine:	N ₁	N ₂	N ₃	S ₁ (mm)	N ₅	S ₂ (mm)	N ₆	N ₇
Number of	1.0	1.0	1.0	3000.0	1.0	0.0	1.0	1.0

Machine Reference Time Table

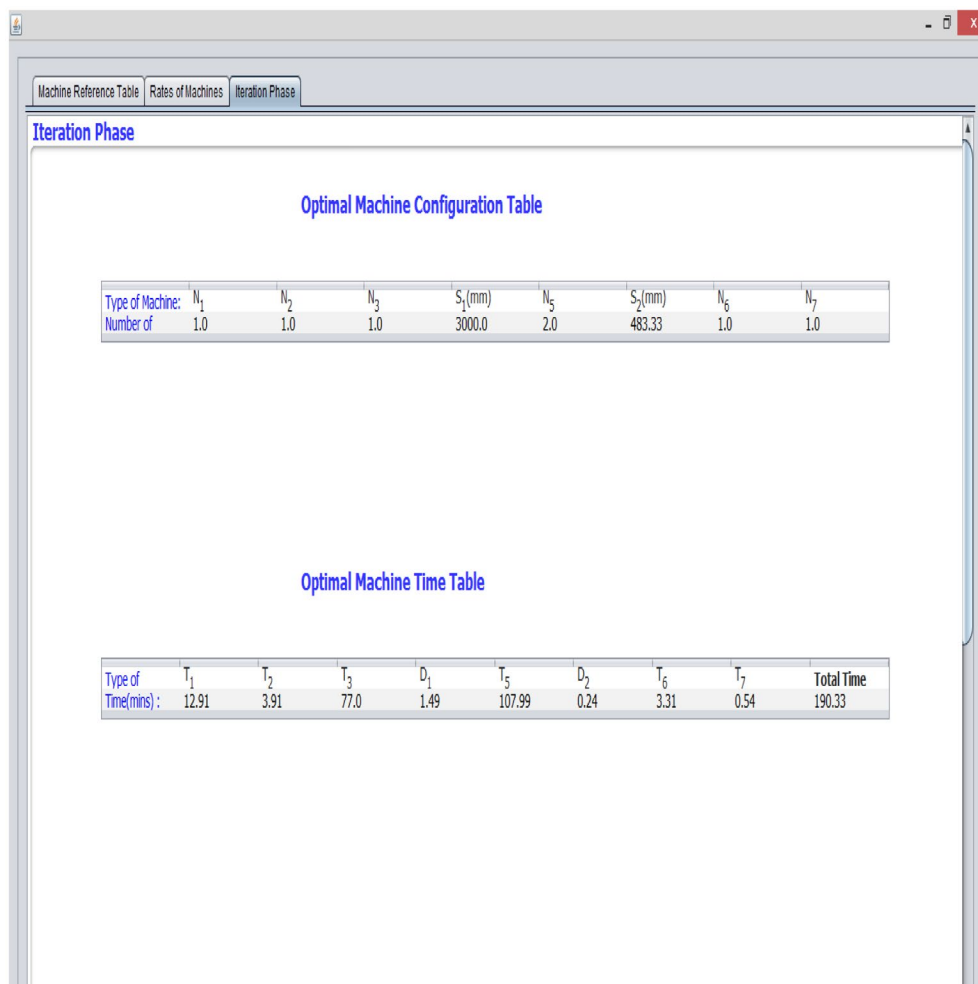
Type of	T ₁	T ₂	T ₃	D ₁	T ₅	D ₂	T ₆	T ₇	Total Time
Time(mins) :	12.91	3.91	77.0	1.49	215.97	0.0	3.31	0.54	315.13

[Go To Iteration Table](#)

(b)

programming command for the dynamic programming design of the flow chart in Figure 3 to generate the optimal machine configuration at initial specified input (W_0) and switches the icon menu from Rate of Machine to Iteration Phase where the optimal machine configuration and the corresponding process time is displayed as shown in Plate 3.

Plate 3. Poundsim interface showing the optimal machine configuration and corresponding machine process time.



The screenshot shows the 'Iteration Phase' window of the Poundsim software. It contains two tables: 'Optimal Machine Configuration Table' and 'Optimal Machine Time Table'.

Optimal Machine Configuration Table

Type of Machine:	N_1	N_2	N_3	$S_1(\text{mm})$	N_5	$S_2(\text{mm})$	N_6	N_7
Number of	1.0	1.0	1.0	3000.0	2.0	483.33	1.0	1.0

Optimal Machine Time Table

Type of	T_1	T_2	T_3	D_1	T_5	D_2	T_6	T_7	Total Time
Time(min):	12.91	3.91	77.0	1.49	107.99	0.24	3.31	0.54	190.33

4. Conclusion

The developed *poundsim* software was able to sort out the necessary data, forecast material flow through the processing plant and implement the design equations to optimally populate the existing dynamic layout model. At inputs within the range of 12.25–134.75 kg pondo yam flour at an increment of 12.25 kg, the software generated optimal machine configuration with a considerable reduction in total processing time compared with the reference machine configuration at equivalent throughput capacity consequently bringing the machine configuration matrix and the economic viability of the processing plant into context. Although the software is tailored to pondo yam flour processing plant, the software application could be broadened to handle other processing plant with similar attributes.

Funding

The authors received no direct funding for this research.

Author details

S.P. Ayodeji¹

E-mail: ayodejisesantut@gmail.com

M.K. Adeyeri¹

E-mail: adeyerimichaeltut@gmail.com

A. Ogunsua¹

E-mail: wolesijohn@gmail.com

¹ Department of Mechanical and Production Engineering, The Federal University of Technology, Akure, Nigeria.

Citation information

Cite this article as: Development of dynamic layout model for pondo yam flour processing plant, S.P. Ayodeji, M.K. Adeyeri & A. Ogunsua, *Cogent Engineering* (2017), 4: 1336872.

References

Afolami, K. A. (2014). *Development and performance evaluation of washing, peeling and parboiling machines for a pondo yam flour process plant* (Unpublished ME thesis). Department of Mechanical Engineering, Federal University of Technology, Akure.

- Ajamu, T. A. (2014). *Development and performance evaluation of drying, milling and sieving machines for a pondo yam flour process plant*. (Unpublished ME thesis). Department of Mechanical Engineering, Federal University of Technology, Akure.
- Alan, R. M. J., & Shang, J. (2004). Hybrid ant systems for the dynamic facility layout problem. *Computers & Operations Research*, 33, 790–803.
- Amine, D., Henri, P., & Sonia, H. (2007). Facility layout problems: A survey. *Annual Reviews in Control*, 31, 255–267.
- Ayodeji, S. P., Khubulani, M., Oduetse, M., & Mohammed, O. O. (2015). A control system for a pondo yam flour processing plant. *African Journal of Science, Technology, Innovation and Development*, 7, 192–200.
<https://doi.org/10.1080/20421338.2015.1040285>
- Ayodeji, S. P., Olabanji, O. M., & Adeyeri, M. K. (2012). Design of a process plant for the production of pondo yam. *International Journal of Engineering (IJE)*, 6, 10–24.
- Balakrishnan, J., & Cheng, C. H. (2000). Genetic search and the dynamic layout problem. *Computers and Operations Research*, 27, 587–593.
[https://doi.org/10.1016/S0305-0548\(99\)00052-0](https://doi.org/10.1016/S0305-0548(99)00052-0)
- Balakrishnan, J., Cheng, C. H., Conway, D. G., & Lau, C. M. (2003). A hybrid genetic algorithm for the dynamic plant layout problem. *International Journal of Production Economics*, 86, 107–120.
[https://doi.org/10.1016/S0925-5273\(03\)00027-6](https://doi.org/10.1016/S0925-5273(03)00027-6)
- Baykasoğlu, A., & Gindy, N. N. Z. (2001). A simulated annealing algorithm for dynamic layout problem. *Computers and Operations Research*, 28, 1403–1426.
[https://doi.org/10.1016/S0305-0548\(00\)00049-6](https://doi.org/10.1016/S0305-0548(00)00049-6)
- Bos, J. (1993). Zoning in forest management a quadratic assignment problem solved by simulated annealing. *Journal of Environmental Management*, 37, 127–145.
- Heragu, S. (1992). Recent models and techniques for solving the layout problem. *European Journal of Operational Research*, 57, 136–144.
[https://doi.org/10.1016/0377-2217\(92\)90038-B](https://doi.org/10.1016/0377-2217(92)90038-B)
- Kaku, B., & Mazzola, J. B. (1997). A Tabu-search heuristic for the Plant Layout Problem. *INFORMS Journal on Computing*, 9, 374–384.
<https://doi.org/10.1287/ijoc.9.4.374>
- Koopmans, T. C., & Beckmann, M. (1957). Assignment problems and the location of economic activities. *Econometrica*, 25, 53–76. <https://doi.org/10.2307/1907742>
- Kusiak, A., & Heragu, S. S. (1987). The facility layout problem. *European Journal of Operational Research*, 29, 229–251.
[https://doi.org/10.1016/0377-2217\(87\)90238-4](https://doi.org/10.1016/0377-2217(87)90238-4)
- McKendall Jr., A. R., & Shang, J. (2006a). Hybrid ant systems for the dynamic facility layout problem. *Computers and Operations Research*, 33, 790–803.
<https://doi.org/10.1016/j.cor.2004.08.008>
- McKendall Jr., A. R., & Shang, J. (2006b). Simulated annealing heuristics for the dynamic facility layout problem. *Computers and Operations Research*, 33, 2431–2444.
<https://doi.org/10.1016/j.cor.2005.02.021>
- Ogunsua, A. (2016). *Design of dynamic facility layout and virtual monitoring system for pondo yam flour processing plant* (Unpublished ME thesis). Department of Mechanical Engineering, Federal University of Technology, Akure.
- Subodh, B. P., & Kuber, S. S. (2014). Productivity improvement in plant by using systematic layout planning (Slp) - A case study of medium scale industry. *International Journal of Research in Engineering and Technology*, 3(4).
- Wu, T. H., Chung, S. H., & Chang, C. C. (2008). Hybrid simulated annealing algorithm with mutation operator to the cell formation problem with alternative process routings. *Expert Systems with Applications*, 36, 3652–3661.



© 2017 The Author(s). This open access article is distributed under a Creative Commons Attribution (CC-BY) 4.0 license.

You are free to:

Share — copy and redistribute the material in any medium or format
 Adapt — remix, transform, and build upon the material for any purpose, even commercially.
 The licensor cannot revoke these freedoms as long as you follow the license terms.

Under the following terms:

Attribution — You must give appropriate credit, provide a link to the license, and indicate if changes were made.
 You may do so in any reasonable manner, but not in any way that suggests the licensor endorses you or your use.
 No additional restrictions



You may not apply legal terms or technological measures that legally restrict others from doing anything the license permits.

Cogent Engineering (ISSN: 2331-1916) is published by Cogent OA, part of Taylor & Francis Group.

Publishing with Cogent OA ensures:

- Immediate, universal access to your article on publication
- High visibility and discoverability via the Cogent OA website as well as Taylor & Francis Online
- Download and citation statistics for your article
- Rapid online publication
- Input from, and dialog with, expert editors and editorial boards
- Retention of full copyright of your article
- Guaranteed legacy preservation of your article
- Discounts and waivers for authors in developing regions

Submit your manuscript to a Cogent OA journal at www.CogentOA.com

