
Differential Evolution and Simulated Annealing Technique for Design and Developing of Process Plan to Perform Optimization on Finite **Capacity Scheduling for Simultaneous System of Machines**

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Abstract Differential evolution and simulated annealing technique for simultaneous scheduling of machines is defined as any computer program that uses advanced mathematical algorithm for logic to perform optimization and simultaneous on finite capacity scheduling and others. This work addresses the problem of simultaneous system of machine and Automated Guided Vehicles in Flexible Manufacturing System so as to minimize make span and mean tardiness. Nontraditional approach is most effective algorithm that aims to converge and give optimal solution in a shorter time. Therefore, in this work a suitable simultaneous scheduling mechanism is design to generate accurate scheduling using simulated annealing and differential evolution methodology, which gives optimum solution for the bench mark problem chosen from literature

Key Words: Automated Guided Vehicles, Make-span, Tardiness, Simultaneous Scheduling, Differential **Evolution, Simulated Annealing**

1. INTRODUCTION

The objective of scheduling is to find a way to assign and sequence the use of these shared resources such that production constraints are satisfied and production costs are minimized. Simultaneous scheduling in the manufacturing environment can be defined as the process of deciding what happens when and where. In scheduling theory, it is often assumed that the time taken to move jobs from one machine to another is negligible. But in many real life situations, this Movement can have a significant effect on the complete time of the jobs, thus adding a parameter to the optimization function. This work looks at the situation where the job travelling time between machines is taken into account. Author, therefore, present a differential evolution (DE) and simulated annealing (SA) for the simultaneous scheduling of machines and AGVs in an FMS for process plan design. Scheduling is the process of allocating shared resources over time for competing activities is known as scheduling. It has been the subject of a significant amount of literature in the operations research field. A flexible manufacturing system is a highly automated manufacturing system well suited for the simultaneous production of a wide variety of part types in low to mid volume quantities at a low cost while maintaining a high quality of the finished products.

1.2 LITERATURE SURVEY

The importance of the material handling system for the efficiency of the overall system has been emphasized by several researchers.

Medikondu.Nageswararao et al., proposed an efficient and optimized Automated Guided Vehicles (AGVs) operation plays a critical role in improving the performance of a Flexible Manufacturing System (FMS). It is proven that the method is capable to provide better solution compared to others [1].

Noboru Murayama and Seiichi Kawata. Focused on simultaneous scheduling of processing machines and multiple-load automated guided vehicles and proposed a simulated annealing method for the simultaneous scheduling problems of machines and multiple-load AGVs to obtain relatively good solutions for a short time [2].

K. V. Subbaiah et al., proposed a Simultaneous scheduling problem using a dynamic programming approach. They tested different machines and AGV scheduling rules in FMS against the mean flow time criterion [3].

I. A. Chaudhry et al., proposed a Production scheduling is concerned with the efficient allocation of resources over time for manufacturing. Scheduling problems arise whenever a common and finite set of resources (labour, material and equipment) must be used to make a variety of different products during the same period of time [4].

B. Siva Prasad Reddy and C.S.P. Rao proposed the interactions between the machine scheduling and the scheduling of Material Handling System in a given Flexible Manufacturing System (FMS) layout with an objective of minimization of makespan and to study the overall feasibility of generation of optimal simultaneous schedules [5].

2. MATERIALS AND METHODS

2.1 Methodology

Authors have considered 2 different layouts and 2 job sets consisting of 1- 10 different job sets and operations on machines to be performed. The problem is formulated as a nonlinear mixed integer programming model. Its objective is makespan minimization, mean makespan, mean tardiness and CPU time.



Figure 1: DE and SA Method

2.1 Objectives function

Operation completion time = $O_{ij} = T_{ij} + Pij$ J = Job operation i = Job $T_{ij} = Traveling time$ Pij = Operation processing time Tardiness = $T_i = C_i - D_i$ where $D_i = It$ is due date Mean Tardiness = $\frac{l}{n} \sum_{i=1}^{n} T_i$ Where, n = number of jobs; Job Completion Time = $C_i = \sum_{i=1}^{n} O_{ij}$ Make span = C_{max} . i.e., $c_1, c_2...c_n$

2.3 Algorithms Used

The algorithm applied for the present study is the differential evolution (DE) and simulated algorithm (SA).

Simulated Annealing: Annealing is the process of obtaining low energy states of a solid in heat treatment. Annealing

process starts with melting the solid by heat treatment. Particles constituting the solid are arranged according to that heat treatment. Then, temperature is decreased which results in minimum energy state.

Differential Evolution: In this study, authors propose a DE algorithm, in which both trial vector generation strategies and their associated control parameter values in generating promising solutions. Consequently, a more suitable generation strategy along with its parameter settings can be determined adaptively to match different phases of the search/evolution. The performance of the DE algorithm is extensively evaluated using codes.

3. RESULT AND DISCUSSION

Table 1: Comparison of make-span for optimum solution using literature review & proposed methods for t/p > 0.25

Problem Number	STW	UGA	AGA	PGA	SFHA	DE	SA
EX39	110	105	106	105	94	69	68
EX310	143	143	141	139	127	80	95
EX49	125	123	122	122	126	59	74
EX410	171	164	159	159	173	73	110

Inference 1.1: Table 2 shows that the comparison of makespan for optimum solution using literature review and proposed methods (DE and SA) for t/p >0.25. Make-span for differential evolution can find better solutions than those of literature review and proposed method of simulated annealing algorithm, for every stage of EX39, EX310, EX49 and EX410 problems

Table 2: Comparison of make-span for DE and SAfor t/p ratio >0.25

Problem No.	DE	SA
EX39	69	68
EX310	80	95
EX49	59	74
EX410	73	110

Inference 1.2: From table 6.8 it is observed that the comparison of make-span for optimum solution using proposed methods (DE and SA) for t/p > 0.25. Make-span for



differential evolution can find better solutions than those of simulated annealing algorithm, for every stage of EX39, EX310, EX49 and EX410 problems.

3.1 Charts for benchmark problems



Chart - 1: Comparison of Make-span Performance for literature review for t/p ratio >0.25

Inference 1.2: Chart 1 shows that the graph plotted problem number v/s Make-span for optimum solution using literature review and proposed methods for t/p > 0.25. Mean Tardiness for DE can find better solutions than those of SA algorithm, for problem numbers.



Chart -2: Convergence of Make-span Performance for problem no EX39, for t/p >0.25

Inference 1.3: Chart 2 shows that the Convergence graph plotted No. of Generations v/s Make-span for optimum solution using proposed methods for t/p > 0.25. Make-span for DE can find better solutions than those of SA algorithm, for Problem No. EX39.



Chart 4: Convergence of Makespan Performance for problem no EX310, for t/p >0.25

Inference 1.4: Chart 4 shows that the Convergence graph plotted No. of Generations v/s Makespan for optimum solution using proposed methods for t/p > 0.25. Makespan for DE can find better solutions than those of SA algorithm, for Problem No. EX310.



Chart - 3: Convergence of Make-span Performance for problem no EX410, for t/p >0.25

Inference 1.4: Chart 3 shows that the Convergence graph plotted No. of Generations v/s Make-span for optimum solution using proposed methods for t/p >0.25. Make-span for DE can find better solutions than those of SA algorithm, for Problem No. EX410







Inference 1.5: Chart 4 shows that the Convergence graph plotted No. of Generations v/s Makespan for optimum solution using proposed methods for t/p >0.25. Makespan for DE can find better solutions than those of SA, for Problem No. EX49.

4. DISCUSSION ABOUT OBTAINED RESULTS

In all these problems there are four machines, two layouts, two job set and 2vehicles. Table 1 consists of problems whose t/p ratios are greater than 0.25 and charts 1 to N consists of graphs for particular problem whose t/p ratios are greater than 0.25. From table 1 for all the 4 benchmark problems, it gives better results using proposed algorithm of differential evolution. From charts 1 to N for all the 4 benchmark problems, the graph charts show the performance of DE is more effective than SA and other five literature algorithms. A code is used to designate the example problems which are given in the tables. The digits that follow EX indicate the layout and job set. Looking DE closer in all tables and figures (graph charts).

5. CONCLUSIONS

Efficient scheduling of all the resources plays a vital role in achieving the required targets in a flexible manufacturing system. The following conclusions are obtained during results and discussion.

For a number of well-studied problems in combinatorial optimization of DE, SA and five literature algorithms usually leads to the conclusion that DE algorithm is more efficient and more effective than literature algorithms and SA.

- DE has potential of finding minimum Make-span, mean Make-span and mean tardiness than other five literature algorithms.
- For convergence chart of all the problems, from chart 1 to 4 (graph chart) and from table 1 and 2, DE gives effective results like- minimum Make-span, Mean Make-span and Mean tardiness when compared with proposed algorithm of SA.

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