INTERNATIONAL TRANSACTIONS IN OPERATIONAL RESEARCH





Intl. Trans. in Op. Res. 23 (2016) 551–591 DOI: 10.1111/itor.12199 INTERNATIONAL TRANSACTIONS IN OPERATIONAL RESEARCH

A research survey: review of flexible job shop scheduling techniques

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Received 16 October 2014; received in revised form 7 May 2015; accepted 9 July 2015

Abstract

In the last 25 years, extensive research has been carried out addressing the flexible job shop scheduling (JSS) problem. A variety of techniques ranging from exact methods to hybrid techniques have been used in this research. The paper aims at presenting the development of flexible JSS and a consolidated survey of various techniques that have been employed since 1990 for problem resolution. The paper comprises evaluation of publications and research methods used in various research papers. Finally, conclusions are drawn based on performed survey results.

Keywords: scheduling; flexible job shop; partial flexibility; total flexibility; heuristics; metaheuristics

1. Introduction

The last three decades have seen extensive development of efficient techniques to solve the flexible job shop scheduling problem (FJSSP). The scope and purpose of this paper is to present a survey of various techniques used for solving FJSSP using different objective functions. Numerous approaches have thus been investigated and these techniques are classified for ease of analysis.

The paper comprises the following major sections. The FJSSP is defined in Section 2. A summary of sources for the published papers and their year-wise distribution is presented in Section 3 and later a summary of various objective functions is given in Section 4. Section 5 gives a brief description of each technique used along with salient features of published work. Analysis and discussion of the survey results are presented in Section 6 and the paper concludes with a future research roadmap.

2. Flexible job shop scheduling problem

Job shop scheduling (JSSP) is a classical operations research problem that has been considered as a hard combinatorial optimization problem since the 1950s. In terms of computational complexity, JSSP is NP-hard in the strong sense (Garey and Johnson, 1979). Therefore, even for very small JSSP instances, an optimal solution cannot be guaranteed. In a job shop, every job may have a separate processing sequence. In the general JSSP, there is a finite set of n jobs to be processed on a finite set of m machines. Each job comprises a set of tasks that must be performed on a different machine and in specified processing times, in a given job-dependent order. A typical objective of this process is to minimize the total completion time required for all jobs or the makespan.

A flexible job shop problem (FJSSP) is an extension of the classical JSS problem that allows an operation to be processed by any machine from a given set of alternative machines. A general FJSSP may be formulated as follows:

- (1) There is a set of *n* jobs to be processed on *m* machines.
- (2) The set of *m* machines is noted as: $M = \{M_1, M_2, \dots, M_m\}$.
- (3) The job consists of a sequence of n_i operations as: $(O_{i,1}, O_{i,2}, \ldots, O_{i,ni})$.
- (4) Each operation has to be performed to complete the job. The execution of each operation j of a job $i(O_{i,j})$ requires one machine out of a set of given machines $M_{i,j}$. The time of operation $O_{i,j}$ running on $M_{i,j}$ is $p_{i,j,k}$. Typically following assumptions are considered in a general FJSSP:
 - (a) All machines are available at time t = 0.
 - (b) All jobs are available at time t = 0.
 - (c) Each operation can be processed by only one machine at a time.
 - (d) There are no precedence constraints among the operations of different jobs; therefore jobs are independent from each other.
 - (e) No pre-emption of operations is allowed, that is, an operation once started cannot be interrupted.
 - (f) Transportation time of jobs between the machines and time to setup the machine for processing a particular operation are included in the processing time.

Classical JSS requires sequencing of operations on fixed machines, whereas in flexible JSS the assignment of an operation is not fixed in advance and can thus be processed on a set of capable machines. Therefore, in FJSSP we not only deal with sequencing, but also with assignment of operations to suitable machines (routing). FJSSP is therefore more complex than JSSP as it considers the determination of machine assignment for each operation. The scheduling of jobs in FJSSP can be categorized into following two subproblems.

- 1. A routing subproblem where we have to select a suitable machine among the available ones to process each operation.
- 2. A scheduling subproblem, where assigned operations are sequenced on all selected machines to obtain a feasible schedule that minimizes a predefined objective.

Based on flexibility, Kacem et al. (2002a) has classified the FJSSP into following two subproblems.

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- 1. Total FJSSP: each operation can be processed on any of the "m" machines in the shop.
- 2. Partial FJSSP where some operations are only achievable on part of the available "m" machines in the shop.

Brucker and Schlie (1990) were the first to address FJSSP. Over the past 25 years, different methods and algorithms have been developed to solve this class of problem. Due to complexity of FJSSP, researchers have used a large number of techniques ranging from mathematical to various metaheuristics such as evolutionary algorithms (EAs), ant colony optimization (ACO), particle swarm optimization, and so on. This paper attempts to consolidate this research and present the findings.

3. Research survey results

A total of 404 distinct publications were found addressing the FJSSP. Some of the research papers presented more than one technique/algorithm to solve the problem that is categorized into 410 different applications. Selected time period of these research papers is between 1990 and February 2014. Articles were searched mainly on major databases such as SpringerLink, Science Direct, IEEE Xplore, Scopus, EBSCO, etc. and other web sources. All databases were searched for "flexible job shop" and "scheduling" in the title and keywords of the articles. The distribution of 404 papers in various sources is given in Table 1 and presented graphically in Fig. 1.

The references addressing the FJSSP has been categorized into five different types namely, journal, conference, book section, thesis, and report. The breakdown of 404 references into each of the mentioned reference category is tabulated in Table 1.

Due to large number of references available on the subject, this research work further concentrates on 191 journal papers only. The year-wise distribution of the 191 references is presented in Fig. 2. These 191 research papers address 197 different applications or techniques as some of these present more than one technique for problem resolution. Due to the practical nature of the problem, FJSSP has gained considerable importance during the last few years. It can be observed from Fig. 2 that relatively small number of papers addressed the problem in the first 10 years, which has grown tremendously from 2010 to 2013.

Table 1 Distribution of type of references

| Type of reference | Total references |
|-------------------|------------------|
| Book section | 30 |
| Conference paper | 167 |
| Journal article | 191 |
| Report | 4 |
| Thesis | 12 |
| Total | 404 |

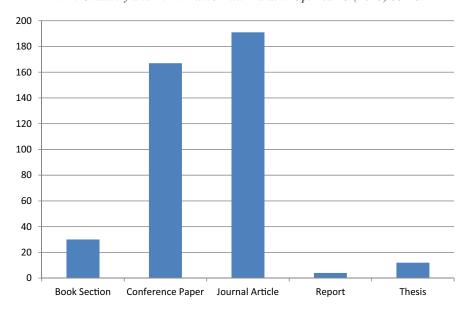


Fig. 1. Source distribution of reference papers.

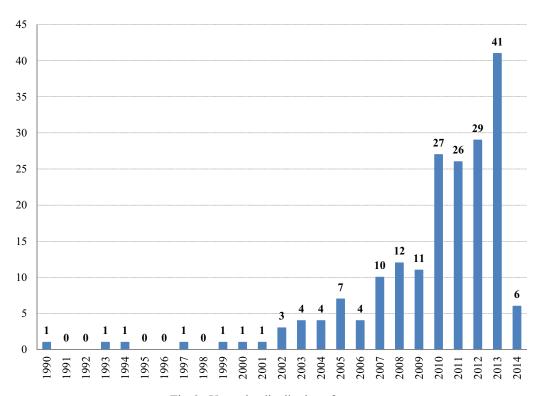


Fig. 2. Year-wise distribution of papers.

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Table 2 Well-known performance measures

| Objective function | Symbol | Type | Interpretation |
|-------------------------------------|-------------------------------|---------|---|
| Makespan or maximum completion time | C_{max} | Regular | The cost of a schedule depends on how long the processing system is devoted to the entire set of jobs |
| Mean completion time | $ar{C}$ | Regular | Average time taken to finish a single time |
| Maximum flow time | F_{max} | Regular | The longest time a job spends in the shop and schedule's cost is directly related to its longest job |
| Mean flow time | $ar{F}$ | Regular | Average time a single job spends in the shop and a schedule's cost is directly related to the average time it takes to process a single job |
| Total tardiness | T | Regular | Positive difference between the completion time and the due date of all jobs and is appropriate when early jobs bring no reward; there are only penalties incurred for late jobs |
| Average tardiness | $ar{T}$ | Regular | Average difference between the completion time and the due date of a single job |
| Total weighted tardiness | $\sum_{i=1}^{n} \alpha_i T_i$ | Regular | Weighted measure that recognizes that some jobs are more important than others |
| Maximum lateness | L_{max} | Regular | Typically used to check how well the due dates are respected, and is appropriate when there is a positive reward for completing a job early; the earlier a job is completed, the larger is the reward |
| Number of tardy jobs | n_T | Regular | Number of jobs that are late |
| Total workload of machines | \dot{W}_T | | Represents the total working time on all machines |
| Critical machine workload | $W_{\scriptscriptstyle M}$ | | Represents most workload among all machines, that is the machine with the maximum workload |

4. Objective functions used in FJSSP

Various researchers have addressed some of the commonly used objective functions that could be regular or irregular and these functions are tabulated in Table 2. An objective function is said to be regular if it is an increasing function, that is, it is always optimal to start (and complete) jobs as early as possible. French (1982) defines a regular measure (M) as a value to be minimized that can be expressed as a function of the job completion times (C_i) as given in relation 1. Thus:

$$M = f\left(C_1, C_2, \dots, C_n\right) \tag{1}$$

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Table 3
Survey results for various performance measures

| | Number of | | |
|--|-----------|------------|--|
| Performance measure | papers | Percentage | |
| Makespan | 88 | 44.67% | |
| Minimum of makespan, workload of most loaded machine, total workload of machines | 46 | 23.35% | |
| Minimum of makespan and mean tardiness | 5 | 2.54% | |
| Minimum of makespan and production costs | 4 | 2.03% | |
| Total tardiness | 3 | 1.52% | |
| Minimum of mean tardiness | 2 | 1.02% | |

and M increases only if at least one of the completion times increases as shown in relation 2. Thus if

$$M' = f(C'_1, C'_2, \dots, C'_n)$$
 (2)

then

$$M' > M$$
 only if $C'_i > C_i$ for at least one $i, 1 \le i \le n$.

In case of regular measures of performance, it is always aimed to finish an activity earlier, rather than later. Examples include the average or maximum job completion times (makespan), flow times, lateness, or tardiness. However, in case of nonregular measures it may not be preferred to finish all jobs as early as possible. A nonregular performance measure is usually not a monotone function of the job completion times, for example, in a just-in-time environment, finishing jobs too early may represent excess WIP. Table 2 gives some of the well-known performance measures used in FJSSP.

Table 2 only gives the single performance measures, whereas multiobjective performance measures in FJSSP have been adequately addressed by a large number of researchers, whereby two or more performance measures tabulated in Table 2 are optimized simultaneously. Survey results revealed that a total of 55 different objective functions have been used in various publications out of which 49 measures were only used once, which implies that out of 197 research papers only six performance measures contributed to almost 75% of the published research work. Out of these six, only three were single performance measures while others were multiobjective performance measures. The survey results with respect to the six performance measures are given in Table 3, makespan came out to be the most widely used performance measure. In 88 research papers (44.67%) makespan was used as the sole objective function, while in another 78 papers (39.59%) makespan is used in combination with another objective function. The cumulative number of citations for these papers solely using makespan as the objective function is 4210, while when makespan is used in combination with other objective functions this number is noted to be 3482.

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Table 4 Various techniques used to solve FJSSP

| Technique | Number of papers | Percentage | Citations |
|----------------------------|------------------|------------|-----------|
| Hybrid | 69 | 35.03% | 3572 |
| EAs | 47 | 23.86% | 1940 |
| Heuristic | 19 | 9.64% | 598 |
| TS | 12 | 6.09% | 1297 |
| Integer/linear programming | 10 | 5.08% | 233 |
| PSO | 8 | 4.06% | 60 |
| Miscellaneous techniques | 7 | 3.55% | 221 |
| NS | 6 | 3.05% | 187 |
| AIS | 5 | 2.54% | 110 |
| Mathematical programming | 4 | 2.03% | 26 |
| SA | 4 | 2.03% | 104 |
| ACO | 3 | 1.52% | 117 |
| GRASP | 2 | 1.02% | 32 |
| ABC | 1 | 0.51% | 48 |

5. Review of techniques used for FJSSP

Based on the research papers collected, various techniques used in these research publications can be broadly categorized into 14 major categories. Brief review of each of the technique is given in the subsequent sections. These categories along with number of papers in each addressing the FJSSP problem are tabulated in Table 4. It can be seen that almost 59% of the papers used hybrid techniques or EAs.

5.1. Ant colony optimization

ACO algorithm is a probabilistic metaheuristic for solving combinatorial problems, which was introduced in 1990s. ACO is a member of the ant colony algorithms family in swarm intelligence methods. ACO is inspired by the pheromone trail laying behavior of real ant colonies whereby it mimics ants' social behaviors in finding shortest paths. ACO was initially developed by Dorigo (1992) to solve travelling salesman problem. The first known application of ACO to JSS is attributed to Colorni et al. (1994).

Rossi and Dini (2007) with 109 citations applied ACO in FJSSP with a sequence-dependent setup and transportation time, in addition operation lag times have also been taken into consideration. The authors develop an effective pheromone trail coding and tailored ant colony operators to minimize makespan. The algorithm has been tested with the standard benchmarks and problems. Xing et al. (2008) with 13 citations propose a double-layered ACO, where the upper layer assigns operations to machines while the lower layer schedules operations on each machine. The authors consider multiobjective performance measure where objective is to simultaneously minimize makespan, workload of most loaded machine, and total workload of machines. Experimental results suggest that the proposed algorithm is a feasible and effective approach for the multiobjective FJSSP. Huang et al. (2013) with eight citations considered minimization of the sum of weighted earliness

and tardiness costs and proposed a two pheromone ACO approach for the FJSSP. The algorithm adds second pheromone group to the ant system in order to solve the scheduling problem faster. Computational results show that the proposed two pheromone ACO strategy yields better results as compared to traditional ACO and integer programming for a wide range of problems.

5.2. Artificial bee colony (ABC)

ABC algorithm is a swarm-based metaheuristic algorithm, introduced by Karaboga (2005) for optimization of numerical problems. The algorithm is motivated by the intelligent behavior of honey bees, that is, foraging behavior of honey bee colonies. The model consists of three essential components: employed, unemployed foraging bees, and food sources. The first two components, employed and unemployed foraging bees, search for rich food sources, the third component, close to their hive. The model also defines two leading modes of behavior that are necessary for self-organizing and collective intelligence: recruitment of foragers to rich food sources resulting in positive feedback and abandonment of poor sources by foragers causing negative feedback. The algorithm is specifically based on the model proposed by Tereshko and Loengarov (2005) for the foraging behavior of honey bee colonies. First reported instance of ABC algorithm application in JSS is attributed to Pansuwan et al. (2010).

Wang et al. (2012c) with 48 citations presented an effective ABC for solving the FJSSP with the criterion to minimize makespan. ABC algorithm stresses on the balance between global and local exploitation, thus special encoding and decoding schemes that are effective searching operators including hybrid initialization, crossover, mutation, local search, and population updating were well designed in the employed bee phase, onlooker bee phase, and scout bee phase. The proposed algorithm was capable to solve the FJSSP effectively, efficiently, and robustly, which has been demonstrated by simulation tests and comparisons to several existing algorithms.

5.3. Artificial immune system (AIS)

AIS are adaptive systems inspired by theoretical immunology and observed immune functions, principles, and models for applications to complex problem domains. The AISs are composed of intelligent methodologies and are inspired by the natural immune system to solve real-world problems. In computer science, AIS are a class of computationally intelligent systems inspired by the principles and processes of the vertebrate immune system. The algorithms typically exploit the immune system's characteristics of learning and memory for problem solution. The origins of AIS has its roots in the early theoretical immunology work by Farmer et al. (1986), whereas first known application of algorithm to scheduling was presented by Hart et al. (1998).

The research work addressing AIS application to FJSSP is shown in Table 5.

5.4. Evolutionary algorithms

Algorithms are a subset of evolutionary computation. EA is a generic population-based metaheuristic optimization algorithm that is inspired by natural evolution. EAs consist of several heuristics that are able to solve optimization tasks by imitating some aspects of natural evolution. An EA

Table 5 Applications AIS in FJSSP

| Article | Application | Algorithm and shop details | Objective function | Citations |
|----------------------------------|-------------|---|---|-----------|
| Bagheri et al. (2010) | Research | Artificial immune algorithm (AIA) that uses a combination of strategies is utilized for generating the initial population. Algorithm also employs multiple different mutations for reassigning and resequencing | Makespan | 100 |
| Akhshabi et al. (2011b) | Research | Clonal selection algorithm (CSA) to solve FJSSP | Makespan | 7 |
| Davarzani et al. (2012) | Research | Multi-objective AIA based on Pareto optimality based on an integrated approach to solve a stochastic FJSSP | Minimum of makespan, workload of most loaded machine, total workload of machines | 2 |
| Davoudpour and Azad (2012) | Research | AIS for solving multiobjective. Proposed approach efficiently solves large scale problems | Minimum of makespan, workload of most loaded machine, total workload of machines | _ |
| Sadrzadeh (2013) | Research | AIS for FJSSP with sequence-dependent setup times | Minimum of makespan and mean tardiness | 6 |

uses mechanisms inspired by biological evolution, such as reproduction, mutation, recombination, and selection. Candidate solutions to the optimization problem play the role of individuals in a population, while fitness function determines quality of the solutions. Evolution of the population takes place after repeated application of the above operators. Some of the well-known EA methods are listed below.

- 1. *Biogeography-based optimization* (*BBO*) it is an EA that optimizes a function by stochastically and iteratively improving candidate solutions with regard to a given measure of quality, or fitness function.
- 2. *Differential evolution* (*DE*) it is based on vector differences and is therefore primarily suited for numerical optimization problems.
- 3. *Evolution strategy (ES)* it works with vectors of real numbers as representations of solutions, and typically uses self-adaptive mutation rates.
- 4. Gene expression programming (GEP) GEP explores a genotype–phenotype system, where computer programs of different sizes are encoded in linear chromosomes of a fixed length.
- 5. Genetic Algorithms (GA) this is most popular type of EA. One seeks the solution of a problem in the form of strings of numbers, by applying operators such as recombination and mutation.

- 6. Genetic programming (GP) solutions are in the form of computer programs, and their fitness is determined by their ability to solve a computational problem.
- 7. Harmony Search (HS) population-based stochastic algorithm that mimics the behavior of a music orchestra when aiming at composing the most harmonious melody, as measured by aesthetic standards.
- 8. Learning classifier system (LCS) it is a machine learning technique that combines evolutionary computing, reinforcement learning, supervised or unsupervised learning, and heuristics to produce adaptive systems.
- 9. *Memetic Algorithms (MA)* EAs in which two search techniques are combined, genetic algorithms and some form of local search.
- 10. Estimation of distribution algorithm (EDA) EDAs are also called probabilistic model-building genetic algorithms. These are stochastic optimization methods that guide the search for the optimum by building and sampling explicit probabilistic models of promising candidate solutions.

The research work addressing EA application to FJSSP is shown in Table 6.

5.5. Greedy randomized adaptive search procedure (GRASP)

GRASP is a metaheuristic algorithm commonly applied to combinatorial optimization problems. It was first introduced by Feo and Resende (1989). Each iteration consists of two phases: construction and local search. The construction phase builds a solution, when this solution is not feasible, it is necessary to apply a repair procedure to achieve feasibility. However, when a feasible solution is obtained, its neighborhood is investigated until a local minimum is found during the local search phase (Resende and Ribeiro, 2010).

Rajkumar et al. (2010) with 16 citations consider FJSSP with nonfixed availability constraints where machines are unavailable due to preventive maintenance and proposed GRASP algorithm to minimize makespan, workload of most loaded machine, and total workload of machines. Rajkumar et al. (2011) with 16 citations present a GRASP algorithm to minimize makespan, workload of most loaded machine, and total workload of machines in an FJSSP with limited resource constraints. The main constraint of this scheduling problem is that each operation of a job must follow an appointed process order and each operation must be processed on an appointed machine.

5.6. Integer/Linear programming

An integer/linear programming problem is a mathematical optimization or feasibility program in which some or all of the variables are restricted to be integers. In many settings, the term refers to integer linear programming, in which the objective function and the constraints (other than the integer constraints) are linear. The method was first developed by Kantorovich (1940) for use during World War II to plan expenditures and returns in order to reduce costs to the army and increase losses to the enemy. The method was kept secret until 1947 when George B. Dantzig published the

Table 6 Applications of EAs in FJSSP

| Article | Application | Algorithm and shop details | Objective function | Citations |
|---------------------------|-------------|--|--|-----------|
| Jensen (2003) | Research | GA, rescheduling in a dynamic FJSSP with machine breakdowns | Makespan | 166 |
| Kis (2003) | Research | GA | Makespan | 55 |
| Kim et al. (2003) | Research | Symbiotic EA that simultaneously solves process planning and scheduling | Makespan | 237 |
| Jang et al. (2003) | Research | GA, FJSSP with multilevel job structures where end products are assembled from subassemblies or manufactured components | Total tardiness | 4 |
| Zhang and Gen (2005) | Research | Multistage operation-based GA to solve FJSSP | Minimum of makespan, workload of most loaded machine, total workload of machines | 236 |
| Chan et al. (2006) | Research | GA-based approach to solve a resource-constrained operations-machines assignment problem and flexible job-shop scheduling problem | Makespan | 64 |
| Ho et al. (2007) | Research | LEarnable Genetic Architecture (LEGA), that provides an effective integration between evolution and learning within a random search process | Makespan | 131 |
| Pezzella et al. (2008) | Research | GA that integrates different strategies for generating the initial population, selecting the individuals for reproduction, and reproducing new individuals | Makespan | 393 |
| Zandieh et al. (2008) | Research | GA | Makespan | 1 |
| Saad et al. (2008) | Research | GA with choquet integral for dealing with multiple criteria decision making | Minimum of makespan, workload of most loaded machine, total workload of machines, sum of weighted earliness and weighted tardiness, sum of production cost | 29 |
| Li and Huo (2009) | Industry | GA, FJSSP with setup times, consideration of maintenance planning and intermediate inventory restriction | Total setup time on machines and waiting time of jobs | 11 |

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Table 6 Continued

| Article | Application | Algorithm and shop details | Objective function | Citations |
|---------------------------------------|-------------|---|--|-----------|
| Lei (2010) | Research | Decomposition-integration GA, FJSSP with fuzzy processing times | Minimization of fuzzy makespan | 56 |
| De Giovanni and Pezzella (2010) | Research | GA, FJSSP in distributed manufacturing environment where jobs are processed by a system of several flexible manufacturing units (FMUs) | Makespan for all FMUs | 118 |
| Fattahi and Fallahi (2010) | Research | GA, dynamic FJSSP with addition of new jobs and machines | Minimum of efficiency and stability of schedules where (stability = starting time deviation + total deviation penalty) | 27 |
| Sun et al. (2010) | Research | GA, FJSSP with partial flexibility and JIT request | Makespan | 3 |
| Moradi et al. (2010) | Research | Learnable genetic architecture (LEGA), FJSSP with preventive maintenance | Makespan | 13 |
| Defersha and Chen (2010) | Research | Parallel GA, FJSSP with sequence-dependent setup time, attached or detached setup time, machine release dates, and time lag requirements between finish and start times of two successive operations of a job | Makespan | 17 |
| Rahmati and Zandieh (2011) | Research | Biogeography-based optimization (BBO) algorithm | Minimum of makespan, workload of most loaded machine, total workload of machines | 30 |
| Mati et al. (2011) | Research | GA, FJSSP with blocking constraints whereby buffer capacity between workstations is limited | Makespan | 9 |
| Al-Hinai and Elmekkawy (2011bb) | Research | Two-stage GA, FJSSP with random machine breakdowns | Makespan | 54 |
| Zhang et al. (2011) | Research | GA with global selection, local selection, and random selection procedures for generating initial solution | Makespan | 90 |
| Ida and Oka (2011) | Research | GA | Minimum of maximum workload | _ |

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Table 6 Continued

| Article | Application | Algorithm and shop details | Objective function | Citations |
|-------------------------------------|-------------|---|--|-----------|
| Farughi et al. (2011) | Research | MA, FJSSP with overlapping in operations where start of successor operation of job not necessarily starts after the finish of its predecessor | Makespan | - |
| Moradi et al. (2011) | Research | Nondominated sort GA (NSGA-II), FJSSP with fixed interval preventive maintenance | Minimum of makespan for the production part and the minimum of the system unavailability for maintenance part | 55 |
| Nicoara et al. (2011) | Research | GA (NSGA-II) with dynamic application of genetic operators and population partial re-initialization | Minimizing makespan, number of late operations, and average ratio of idle times in workshop | 28 |
| Lei (2012) | Research | Co-evolutionary GA (CGA) with a novel representation and crossover operator, the populations of job sequencing and machine assignment evolve independently and cooperate each other to approximate the best solutions of the problem | Minimum of fuzzy makespan | 20 |
| Nagamani et al. (2012) | Research | GA | Makespan | _ |
| Vaghefinezhad and Wong (2012) | Research | GA | Maximum of total profit by determining the amount of production in normal time and overtime | _ |
| Lee et al. (2012) | Research | GA for FJSSP with 'AND'/'OR' precedence constraints | Makespan | 63 |
| Rabiee et al. (2012) | Research | Four GAs (NSGA-II, NRGA, MOGA and PAES), statistical analysis is also conducted to analyze the performance of the algorithms in five metrics namely nondominated solution, diversification, mean ideal distance, quality metric and data envelopment analysis | Minimum of makespan and total operating cost | 20 |
| Chen et al. (2012) | Industry | Grouping GA (GGA), FJSSP with re-entrant processes. GGA consisting of machine selection and operation scheduling module applied in a real weapon production factory | Minimum of makespan, total tardiness and total machine idle time | 25 |
| | | | | Continued |

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Table 6 Continued

| Article | Application | Algorithm and shop details | Objective function | Citations |
|--------------------------|-------------|---|---|-----------|
| Pandian et al. (2012) | Research | GA (NSGA-II) with jumping of genes between chromosomes, FJSSP with AGVs | Minimum makespan and material flow | - |
| Yegane et al. (2012) | Research | Memetic algorithm (MA), FJSSP with pre-emption and overlapping in operation | Makespan | _ |
| Wang et al. (2012a) | Research | Bipopulation-based estimation of distribution algorithm (BEDA) | Makespan | 43 |
| Agrawal et al. (2012) | Research | Multiobjective GA | Minimum of the makespan and total machining time | 3 |
| Gen and Lin (2012) | Research | Multistage operation based GA (moGA), FJSSP with AGVs and setup times | Minimum of makespan, workload of most loaded machine, total workload of machines | 7 |
| Xiong et al. (2013) | Research | EA, FJSSP with random machine breakdowns and job release times | Minimum makespan and robustness of schedules | 23 |
| Chiang and Lin (2013) | Research | Simple EA to search for the set of Pareto-optimal solutions | Minimum of makespan, workload of most loaded machine, total workload of machines | 26 |
| Wang et al. (2013c) | Research | GA, FJSSP with machine disruptions and rescheduling with special chromosome that uses JIT routing heuristics taking into account the current state of the scheduling environment | Makespan | 1 |
| Na and Park (2014) | Research | GA, priority rules and local search rules applied to improve the performance of GA | Total tardiness | 4 |
| Nie et al. (2013) | Research | Gene expression programming (GEP), dynamic FJSSP with job release dates | Minimum of makespan, mean flow time, and mean tardiness | 15 |
| Yuan and Xu (2013c) | Research | Memetic algorithm (MA) based on NSGA-II | Minimum of makespan, workload of most loaded machine, total workload of machines | - |
| Zhang et al. (2013) | Research | Memetic algorithm (MA), dynamic FJSSP with random job arrivals | Minimum of makespan and mean tardiness | - |

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Table 6 Continued

| Article | Application | Algorithm and shop details | Objective function | Citations |
|-------------------------------|-------------|---|---|-----------|
| Rahmati et al. (2013) | Research | Nondominated sorting GA (NSGA-II) and nondominated ranking genetic algorithm (NRGA) | Minimum of makespan, workload of most loaded machine, total workload of machines | 14 |
| Gao et al. (2014) | Research | Discrete harmony search (HS) | Minimum of makespan and mean earliness and tardiness | 2 |
| Zambrano Rey et al. (2014) | Research | GA, FJSSP with distributed arrival-time, JIT dynamic scheduling | Minimum quadratic relationship between the earliness and tardiness penalties | 2 |

simplex method and John von Neumann developed the theory of duality as a linear optimization solution, and applied it in the field of game theory. In the postwar era, numerous industries found its use in their daily planning. Since then the method has had important application in various fields most notably being operations research, production planning, and scheduling. The significant applications of integer/linear programming are given in Table 7.

5.7. Neighborhood search (NS)

Variable neighborhood search is a relatively unexplored approach. VNS is a metaheuristic method for solving a set of combinatorial optimization and global optimization problems. It explores distant neighborhoods of the current incumbent solution, and moves from there to a new one if an improvement was noted. The local search method is applied repeatedly to solutions obtained from the neighborhood to local optima. VNS was first proposed by Mladenović and Hansen (1997). VNS was designed for approximating solutions of discrete and continuous optimization problems, it is aimed for solving linear, integer, mixed integer, and nonlinear program problems, etc. VNS applications in FJSSP are shown in Table 8.

5.8. Particle swarm optimization (PSO)

Particle swarm optimization (PSO) is a swarm-based optimization technique developed by Eberhart and Kennedy (1995) and Kennedy and Eberhart (1995). The algorithm is inspired by the flocking and schooling patterns of birds and fish. PSO algorithm is an adaptive algorithm based on a social-psychological metaphor; a population of individuals adapts by returning stochastically toward previously successful regions. PSO is an approach to problems whose solutions can be represented as a point in an *n*-dimensional solution space. A number of "particles" are randomly set into motion

Table 7
Applications of integer/linear programming in FJSSP

| Article | Application | Algorithm and shop details | Objective function | Citations |
|----------------------------|-------------|---|---|-----------|
| Thomalla (2001) | Research | Discrete-time integer programming (IP) based on the Lagrangian relaxation. FJSSP with just-in-time environment. | Minimum of sum of weighted quadratic tardiness | 62 |
| Gomes et al. (2005) | Industry | Integer linear programming. FJSSP with discrete parts manufacturing industries that operate on a make-to-order basis, parallel homogeneous machines, limited intermediate buffers and re-entrant machines | Costs derived from failing to meet the 'just in time' due dates, in-process inventory costs, and costs of orders not fully completed at the end of the scheduling horizon | 39 |
| Özgüven et al. (2010) | Research | MILP. FJSSP with routing and process plan flexibility | Makespan | 63 |
| Gomes et al. (2013) | Research | MILP. FJSSP in make-to-order industries, re-entrant processes and final assembly stage considered simultaneously | Minimum of weighted sum of order earliness, order tardiness and in-process inventory | 4 |
| Mousakhani (2013) | Research | MILP. FJSSP with sequence-dependent setup times | Total tardiness | 5 |
| Birgin et al. (2013) | Industry | MILP. An extended FJSSP is presented where precedence between operations of a job to be are given by an arbitrary directed acyclic graph rather than a linear order | Makespan | 3 |
| Roshanaei et al. (2013) | Research | MILP | Makespan | 6 |
| Hansmann et al. (2013) | Industry | MILP with B&B procedure. FJSSP with machine blockages or machines with restricted accessibility, that is, a busy machine blocks access to all succeeding machines | Makespan | - |
| Torabi et al. (2005) | Research | MINLP procedure for simultaneous determination of machine allocation, sequencing, economic lot-sizing, and scheduling decisions | Minimum the sum of setup and inventory holding costs per time unit without backlogging | 41 |
| Özgüven et al. (2012) | Research | Mixed integer goal programming. FJSSP with separable/nonseparable sequence-dependent setup times | Minimum makespan and balancing the workloads of the machines | 10 |

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Table 8
Applications of VNS in FJSSP

| Article | Application | Algorithm and shop details | Objective function | Citations |
|----------------------------------|-------------|--|---|-----------|
| Yazdani et al. (2010) | Research | Parallel variable neighborhood search (PVNS) to increases the diversification and exploration in search space | Makespan | 123 |
| Amiri et al. (2010) | Research | VNS | Makespan | 35 |
| Lei and Guo (2011) | Research | Swarm-based NS, FJSSP with fuzzy processing conditions | Fuzzy makespan | 7 |
| Bagheri and Zandieh (2011) | Research | VNS, FJSSP with sequence-dependent setup times | Minimum of makespan and mean tardiness | 15 |
| Zheng et al. (2012) | Research | Multiobjective swarm-based neighborhood search (MOSNS), FJSSP with fuzzy processing times | Minimum of fuzzy makespan and maximum machine workload | 6 |
| Lei and Guo (2014) | Research | VNS with two neighborhood-search procedures, Dual-resource (machines & workers) constrained FJSSP | Makespan | 1 |

through this space. At each iteration, they observe the "fitness" of themselves and their neighbors and "emulate" successful neighbors (those whose current position represents a better solution to the problem than theirs) by moving toward them. Various schemes for grouping particles into competing, semi-independent "flocks" can be used, or all the particles can belong to a single global flock. This extremely simple approach has been effective across a variety of problem domains. Applications of PSO in FJSSP are given in Table 9.

5.9. Simulated annealing (SA)

SA is a random-search technique that exploits an analogy to the process in which a metal cools and freezes into a minimum energy crystalline structure and search for a minimum in a more general system; it forms the basis of an optimization technique for combinatorial and other problems.

SA was developed in 1983 by Kirkpatrick and Vecchi (1983) to deal with highly nonlinear problems. SA approaches the global maximization problem similar to using a bouncing ball that can bounce over mountains from valley to valley. It begins at a high "temperature," which enables the ball to make very high bounces, it bounces over any mountain to access any valley, given enough bounces. As the temperature drops the ball cannot bounce so high, and it gets trapped in relatively small ranges of valleys. A generating distribution generates possible valleys or states to be explored. An acceptance distribution is also defined, which depends on the difference between the function value of the present generated valley to be explored and the last saved lowest

Table 9 Applications of PSO in FJSSP

| Article | Application | Algorithm and shop details | Objective function | Citations |
|--|-------------|---|---|-----------|
| Boukef et al. (2008) | Research | PSO | Makespan | 5 |
| Pongchairerks and Kachitvichyanukul (2009) | Research | PSO with multiple social learning topologies in its evolutionary process | Makespan | 14 |
| Liu et al. (2009) | Research | Multiswarm PSO | Minimum makespan of total flow time | 28 |
| Grobler et al. (2010) | Industry | PSO, FJSSP with sequence-dependent setup times, auxiliary resources and machine down time | Minimum of makespan, earliness/tardiness, and queue times | 5 |
| Mekni et al. (2011) | Research | Parameter-free PSO that does not require any parameter tuning (swarm size, sociometry, etc.) | Makespan | _ |
| Akhshabi et al. (2011a) | Research | PSO | Makespan | 2 |
| Mekni et al. (2012) | Research | Parameter-free PSO that does not require any parameter tuning | Minimum of makespan, workload of most loaded machine, total workload of machines | _ |
| Sadrzadeh (2013) | Research | PSO, FJSSP with sequence-dependent setup times | Minimum of makespan and mean tardiness | 6 |

valley. The acceptance distribution decides probabilistically whether to stay in a new lower valley or to bounce out of it. All the generating and acceptance distributions depend on the state of temperature.

Loukil et al. (2007) with 73 citations consider a real-life case study and propose an SA algorithm to minimize makespan, the mean completion time, maximal tardiness, and mean tardiness. Authors also consider various constraints such as batch production—production of several subproducts followed by the assembly of the final product, and possible overlaps for the processing periods of two successive operations of a same job. Fattahi et al. (2009) having 30 citations propose an SA algorithm to minimize makespan in an FJSSP with overlapping in operations. The authors argue that it can be easily adapted for other single objective optimization problems such as minimization of total weighted tardiness. Yazdani et al. (2009) also propose an SA algorithm to minimize makespan in FJSSP. In order to search the solution space, the authors use neighborhood structures of assignment of jobs to machines and sequencing of assigned jobs on a particular machine to generate neighboring solutions. Khalife et al. (2010) having one citation consider minimizing makespan, workload of most loaded machine, total workload of machines for an FJSSP with overlapping in operations, and propose an SA algorithm to solve the problem.

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5.10. Tabu search (TS)

TS is a metaheuristic method originally proposed by Glover (1986). It guides a local heuristic search procedure to explore the solution space beyond local optimality. TS is based on the premise that problem solving, to qualify as intelligent, must incorporate adaptive memory and responsive exploration. The adaptive memory feature of TS allows the implementation of procedures that are capable of searching the solution space economically and effectively. As local choices are guided by information collected during the search, TS contrasts with memory-less design that heavily rely on semi-random processes that implement a form of sampling. The emphasis on responsive exploration in TS, whether in a deterministic or probabilistic implementation, derives from the supposition that a bad strategic choice can often yield more information than a good random choice. Over a wide range of problem settings, strategic use of memory can make dramatic differences in the ability to solve problems. Application of TS in FJSSP is given in Table 10.

5.11. Mathematical programming

A mathematical model is an abstract model that uses mathematical language to describe the behavior of a system. Mathematical models are used particularly in the natural sciences and engineering disciplines such as physics, biology, and electrical engineering, but also in the social sciences such as economics, sociology, and political science; physicists, engineers, computer scientists, and economists use mathematical models most extensively. Eykhoff (1974) defined a mathematical model as "a representation of the essential aspects of an existing system (or a system to be constructed) presenting its knowledge in usable form." Mathematical models can take many forms, including but not limited to dynamical systems, statistical models, differential equations, or game theoretic models.

Elazeem et al. (2011) introduced a mathematical model of the primal problem of FJSSP to minimize the makespan and its dual problem (Abdou's problem). The authors show that the optimal value of Abdou's problem is a lower bound for the objective value of the primal problem. Sun et al. (2014) having one citation propose a new model for FJSSP with machine breakdown based on noncooperative game theory. The authors also propose an algorithm to find the Nash Equilibrium or near-Nash Equilibrium solutions. Robustness and stability are the two measures used by the authors to evaluate the quality of rescheduling. Experiments show that applying game theory can find solution with better performance for both robustness and stability as compared to other rescheduling strategies. Yulianty and Ma'ruf (2013) developed mathematical models for FJSSP with controllable processing time and expected downtime by using predictive approach with minimization of rescheduling cost and tardiness as the objective functions. The authors developed mathematical models to build an initial solution that determines job assignment and processing times of the jobs. The initial solution is later used for rescheduling jobs by considering expected downtime for each machine and downtime probability for each operation. Demir and Kürşat İşleyen (2013) having 25 citations consider various mathematical formulations of FJSSP compiled from literature and propose a time-indexed model to minimize makespan. The models are investigated in terms of binary variables that rely on using sequencing operations on machines named sequence-position,

Table 10 Applications of TS in FJSSP

| Article | Application | Algorithm and shop details | Objective function | Citations |
|--|-------------|--|---|-----------|
| Brandimarte (1993) | Research | Hierarchical approach based on TS by decomposing the problem into assignment and scheduling subproblems | Minimum of makespan and total weighted tardiness | 434 |
| Hurink et al. (1994) | Research | TS | Makespan | 240 |
| Mastrolilli and Gambardella (2000) | Research | TS by reducing the set of possible neighbors to a subset that always contains neighbor with the lowest makespan | Makespan | 370 |
| Kis (2003) | Research | TS | Makespan | 55 |
| Scrich et al. (2004) | Research | Hierarchical and multistart TS | Total tardiness | 73 |
| Saidi-Mehrabad and Fattahi (2007) | Research | TS with an adjacent pairwise interchange method to generate the neighborhoods, FJSSP with sequence-dependent setup times | Makespan | 18 |
| Ennigrou and Ghédira (2008) | Research | TS-based multiagent systems composed of three agent classes: job agents, resource agents, and an interface agent | Makespan | 18 |
| Bożejko et al. (2010) | Research | Parallel TS with two major modules: machine selection module executed sequentially, and operation scheduling module executed in parallel | Makespan | 6 |
| Vilcot and Billaut (2011) | Industry | TS | Minimum of makespan, maximum lateness and total tardiness | 17 |
| Bożejko et al. (2012) | Research | Parallel TS implemented on multi-GPU | Makespan | _ |
| Lee et al. (2012) | Research | TS | Makespan | 63 |
| Jia and Hu (2014) | Research | Path-relinking TS with back-jump tracking | Minimum of makespan, workload of most loaded machine, total workload of machines | 3 |

precedence, and time-indexed variables and compare the computational efficiency of models. Mixed integer linear programming (MILP) is used to solve these mathematical formulations.

5.12. Deterministic heuristics

Various authors have proposed deterministic heuristics for solving FJSSP. Given a particular input, deterministic heuristics always produce same output, with the underlying machine always passing

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through the same sequence of states. While nondeterministic heuristics explicitly involve a random variable. Different individuals using the same algorithm will get different answers. Metaheuristics such as GA, TS, PSO, ACO, SA, and so on, are all nondeterministic methods. Deterministic heuristics used by researchers are generally based on various dispatching rules. The summary of application of heuristics in FJSSP is given in Table 11.

5.13. Hybrid techniques

In the recent years hybrid techniques have become more popular among researchers as compared to pure heuristics or metaheuristics. These techniques combine one or more heuristic or metaheuristic to take advantage of their strengths. Applications of various hybrid techniques employed by researchers in FJSSP are given in Table 12.

5.14. Miscellaneous techniques

A single instance/application of a technique has been classified as a miscellaneous technique's category. Applications of various miscellaneous techniques in FJSSP are given in Table 13.

6. Analysis and discussion

Flexible JSS is considered to be an important area of research and thus falls among the most published class of problems in a manufacturing system. An optimal schedule is difficult to develop due to NP-hard nature of FJSSP. Due to the practical nature of this problem, a large number of researchers have addressed it and proposed various efficient techniques/methods. A total of 404 research publications were found having addressed the problem. In this survey, we have focused on 191 journal publications published from 1990 to February 2014. A total of 197 different solution techniques/methods were used in these papers.

The techniques/methods proposed in 191 journal papers have been categorized in 14 major categories namely: ACO, ABC, AIS, EAs, GRASP, Heuristics, Hybrid Techniques, Integer/Linear Programming, Mathematical Modeling, Miscellaneous Techniques, NS, PSO, SA, and TS.

This survey revealed the following facts:

- 1. Publications addressing FJSSP has grown significantly from 2010 onwards that signifies the importance of the problem under review.
- 2. Country-wise analysis was also performed to ascertain the country of research publication origination. The publications are attributed to a country based on the address of its first author. Country-wise distribution of papers is tabulated in Table 14. The results reveal that China leads the table with a total of 59 publications with Iran following at second and France at third positions with 44 and nine papers, respectively. A total of 213 research papers addressing FJSSP are also presented in various level of conferences. Country-wise distribution of work presented

Table 11 Application of deterministic heuristics in FJSSP

| Article | Application | Algorithm and shop details | Objective function | Citations |
|---|-------------|--|--|-----------|
| Brucker and Schlie (1990) | Research | Polynomial algorithm, 2 Job FJSSP | Makespan | 289 |
| Golenko- Ginzburg and Laslo (2004) | Research | Dispatching rules based on a cyclic coordinate descent method, and a simulation model comprising a modified decision-making rule for cost objectives | Minimum of average scheduling expanses within the time period | - |
| Wu and Weng (2005) | Research | Agent-based heuristic | Minimum of earliness and tardiness | 47 |
| Alvarez-Valdes et al. (2005) | Industry | Heuristic algorithm based on priority rules, FJSSP with no-wait constraints | Minimum the total cost associated to the finishing times of the jobs corresponding to final products | 53 |
| Chen et al. (2007) | Industry | Heuristic based on priority rules, FJSSP with re-entrant processes where certain machines are visited more than once | Minimum of maximum lateness and average tardiness | 35 |
| Fahmy et al. (2007) | Research | Heuristic with rank matrices (Latin rectangles) to insert new jobs in schedules, FJSSP with reactive rescheduling | Efficiency of the revised schedules in terms of the mean flow time, resulting system nervousness, and the required solution time | 2 |
| Shi-Jin et al. (2008) | Research | Heuristic based on filtered beam search algorithm (Ow and Morton, 1988) | Minimum of makespan, workload of most loaded machine, total workload of machines | 21 |
| Chen and Frank Chen (2008) | Research | Heuristic based on priority rules, FJSSP under demand changes | Maximum utilization of all machine | 5 |
| Taghavi-Fard and Saidy (2009) | Research | Heuristic implementing Aker's graphical algorithm (Akers and Friedman, 1955), FJSSP with machine nonavailability constraints | Makespan | _ |
| Baykasoğlu and Özbakır (2010) | Research | Effect of dispatching rules on the scheduling performance through grammars | Minimum of mean tardiness | 19 |

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Table 11 Continued

| Article | Application | Algorithm and shop details | Objective function | Citations |
|------------------------------|-------------|--|--|-----------|
| Wang and Yu (2010) | Research | Heuristic based on filtered beam search algorithm (Ow and Morton, 1988), FJSSP with preventive maintenance | Minimum of makespan, workload of most loaded machine, total workload of machines | 41 |
| Ham et al. (2011) | Research | Real-time scheduling heuristic based on binary integer programming, FJSSP with real-time scheduling | Makespan | 8 |
| Liu and Zhang (2012) | Research | Online scheduling based on dispatching rules | Makespan | _ |
| Lee et al. (2012) | Research | Heuristic | Minimum of makespan and mean flow time | 63 |
| Ziaee (2013) | Research | Heuristic based on a constructive procedure | Makespan | 2 |
| Calleja and Pastor (2013) | Industry | Heuristic based on priority rules, FJSSP with transfer batches | Minimum of average tardiness | 2 |
| Doh et al. (2013) | Research | Heuristic with priority rules | Minimum of makespan, total flow time, mean tardiness, the number of tardy jobs, and the max tardiness | 8 |
| Pérez and Raupp (2014) | Research | Newton's method (Fliege et al., 2009) based hierarchical heuristic algorithm | Minimum of makespan, workload of most loaded machine, total workload of machines | 3 |
| Sadaghiani et al. (2014) | Research | Pareto archive floating search procedure | Minimum of makespan, workload of most loaded machine, total workload of machines | _ |

and published in conferences is tabulated in Table 15. Of 404 papers collected during the study, a major chunk of papers almost 54% papers were from only two countries, that is, China (41.58%) and Iran (12.38%).

3. Papers on FJSSP have appeared in 84 different journals. Of a total of 191 research papers, 66% of the papers appeared in 21 journals with a cumulative impact factor of 193.79 as mentioned in

Table 12 Application of hybrid techniques in FJSSP

| Article | Application | Algorithm and shop details | Objective function | Citations |
|---------------------------------------|-------------|---|--|-----------|
| Dauzère-Pérès and Paulli (1997) | Research | TS with disjunctive graph model | Makespan | 232 |
| Mesghouni et al. (1999) | Research | GA + constraint logic programming | Minimum of Makespan, SD of workload of the resource, mean completion time of the manufacturing orders and SD of the completion time of the manufacturing orders | 27 |
| Baykasoğlu (2002) | Research | SA based on linguistic and Giffler and Thompson (Giffler and Thompson, 1960) priority rule | Makespan | 45 |
| Kacem et al. (2002a) | Research | Localization approach + controlled GA | Minimum of makespan, workload of most loaded machine, total workload of machines | 116 |
| Kacem et al. (2002b) | Research | Pareto-optimality based fuzzy EA | Minimum of makespan, workload of most loaded machine, total workload of machines | 351 |
| Tanev et al. (2004) | Industry | GA + priority dispatching rules | Minimum ratio of tardy jobs, variance of the flow time, amount of mold changes and maximum efficiency of machines | 59 |
| Baykasoğlu et al. (2004) | Research | TS based on grammars and Giffler and Thompson priority rule | Min of makespan and mean tardiness | 55 |
| Xia and Wu (2005) | Research | PSO + SA | Minimum of makespan, workload of most loaded machine, total workload of machines | 478 |
| Gao et al. (2006) | Research | GA + local search (LS). FJSSP with maintenance constraints | Minimum of makespan, workload of most loaded machine, total workload of machines | 79 |
| Butt and Hou-Fang (2006) | Research | GA + scheduling rules | Makespan | _ |
| Imanipour and Zegordi (2006) | Research | TS + backward procedure heuristic | Minimum of total weighted earliness/tardiness | 7 |

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Table 12 Continued

| Article | Application | Algorithm and shop details | Objective function | Citations |
|---|-------------|---|---|-----------|
| Liouane et al. (2007) | Research | ACO + TS | Makespan | - |
| Zribi et al. (2007) | Research | GA + local search (LS) | Makespan | 26 |
| Gao et al. (2007) | Research | GA + local search (LS) procedure based on shifting bottleneck | Minimum of makespan, workload of most loaded machine, total workload of machines | 123 |
| Fattahi et al. (2007) | Research | TS + SA | Makespan | 133 |
| Ho and Tay (2008) | Research | GA + guided local search | Minimum of makespan, workload of most loaded machine, total workload of machines | 48 |
| Gao et al. (2008) | Research | GA + variable neighborhood descent | Minimum of makespan, workload of most loaded machine, total workload of machines | 270 |
| Tay and Ho (2008) | Research | Genetic programming + dispatching rules | Minimum of makespan, mean tardiness, and mean flow time | 190 |
| Li et al. (2008) | Research | PSO + TS | Makespan | 8 |
| Girish and Jawahar (2009) | Research | GA + ACO | Makespan | 22 |
| Gholami and Zandieh (2009) | Research | GA + simulation. Dynamic FJSSP with stochastic breakdowns | Makespan Minimum of mean tardiness | 56 |
| Zhang et al. (2009) | Research | PSO + TS | Minimum of makespan, workload of most loaded machine, total workload of machines | 197 |
| Xing et al. (2010) | Research | ACO + knowledge model | Makespan | 142 |
| Motaghedi- Larijani et al. (2010) | Research | GA + hill climbing | Minimum of makespan, workload of most loaded machine, total workload of machines | 13 |
| Li et al. (2010c) | Research | GA + VNS | Minimum of makespan, workload of most loaded machine, total workload of machines | 5 |

Table 12 Continued

| Article | Application | Algorithm and shop details | Objective function | Citations |
|--|-------------|---|---|-----------|
| Wang et al. (2010) | Research | GA based on immune and entropy principle | Minimum of makespan, workload of most loaded machine, total workload of machines | 61 |
| Lan et al. (2010) | Research | GA with immune mechanism and SA strategy | Minimum of makespan and production costs | _ |
| Frutos et al. (2010) | Research | Memetic algorithm based on the NSGAII acting on two chromosomes with SA for local search (LS) | Minimum of makespan and production costs | 25 |
| Rajabinasab and Mansour (2010) | Research | Pheromone-based multiagent scheduling system. FJSSP with stochastic job arrivals, uncertain processing time, and unexpected machine breakdowns | Various performance measures such as mean flow time, maximum flow time etc. | 45 |
| Li et al. (2010b) | Research | PSO + TS | Makespan | 10 |
| Mahdavi et al. (2010) | Research | Simulation-based DSS | Makespan | 40 |
| Li et al. (2010a) | Research | TS + VNS | Minimum of makespan, workload of most loaded machine, total workload of machines | 81 |
| Li et al. (2011c) | Research | TS + public critical block neighborhood structure | Makespan | 68 |
| Xing et al. (2011) | Research | GA + ACO | Makespan | 14 |
| Al-Hinai and Elmekkawy (2011a) | Research | GA + local search (LS) | Makespan | 22 |
| Jiang et al. (2011) | Research | GA + SA | Minimum of makespan and total workload of machines | 5 |
| Gutiérrez and García- Magariño (2011) | Research | GA with heuristics | Makespan | 23 |
| Li et al. (2011d) | Research | Pareto-based ABC algorithm | Minimum of makespan, workload of most loaded machine, total workload of machines | 19 |
| Li et al. (2011b) | Research | Pareto-based discrete ABC | Minimum of makespan, workload of most loaded machine, total workload of machines | 71 |
| | | | | Continue |

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Table 12 Continued

| Article | Application | Algorithm and shop details | Objective function | Citations |
|--|-------------|---|---|-----------|
| Li et al. (2011a) | Research | Pareto-based local search embedding a VNS-based self-adaptive strategy | Minimum of makespan, workload of most loaded machine, total workload of machines | 12 |
| Moslehi and Mahnam (2011) | Research | PSO + local search | Minimum of makespan, workload of most loaded machine, total workload of machines | 127 |
| Tavakkoli- Moghaddam et al. (2012) | Research | Duplicate GA where initial population is generated randomly in the GA to find a semi-active solution which then evolves by using TS in a certain neighborhood structure as an active solution. Secondary population of the DGA is composed of all active solutions, finally GA is used for the convergence the secondary population | Minimum of makespan and total idleness | - |
| Xiong et al. (2012) | Research | GA + local search (LS) based on critical path theory | Minimum of makespan, workload of most loaded machine, total workload of machines | 4 |
| Dalfard and Mohammadi (2012) | Research | GA + SA. FJSSP with parallel machines and maintenance constraints | Minimum of makespan, average completion time, and tardiness with penalty | 13 |
| Zhang et al. (2012) | Research | GA + TS. FJSSP with transportation constraints and bounded processing times | Makespan | 25 |
| Barzegar et al. (2012a) | Research | GS algorithm + color petri nets | Makespan | 5 |
| Karimi et al. (2012) | Research | Knowledge-based VNS | Makespan | 5 |
| Wang et al. (2012b) | Research | Pareto-based ABC + local search based on critical path | Minimum of makespan, workload of most loaded machine, total workload of machines | 20 |
| Li et al. (2012) | Research | Shuffled frog-leaping algorithm + local search | Minimum of makespan, workload of most loaded machine, total workload of machines | 27 |

Table 12 Continued

| Article | Application | Algorithm and shop details | Objective function | Citations |
|--|-------------|---|---|-----------|
| Li and Pan (2012) | Research | TS + discrete chemical reaction optimization. FJSSP with maintenance activity | Minimum of makespan, workload of most loaded machine, total workload of machines | 15 |
| Wang et al. (2013b) | Research | ABC + VNS. FJSSP with fuzzy processing times | Minimum of fuzzy makespan | 14 |
| Roshanaei et al. (2013) | Research | AIS + SA | Makespan | 6 |
| Yuan and Xu (2013a) | Research | DE algorithm + local search | Makespan | 9 |
| Xu et al. (2013) | Research | Discrete DE algorithm + local search (LS) strategy based on critical path. FJSSP with lot splitting with capacity constraints | Minimum of total production costs, setup costs and tardiness penalty costs | 1 |
| Shao et al. (2013) | Research | Discrete PSO + SA | Minimum of makespan, workload of most loaded machine, total workload of machines | 15 |
| Nagamani et al. (2013) | Research | EA + hill climbing algorithm | Minimum of makespan, workload of most loaded machine, total workload of machines | _ |
| Zhou and Zeng (2013) | Research | GA + SA | Minimum of makespan and sum of SD of processing workload for all working centers | - |
| Jalilvand- Nejad and Fattahi (2013) | Research | GA + SA | Minimum the total cost including delay costs, setup costs and holding costs | 5 |
| Shahsavari- Pour and Ghasemisha- bankareh (2013) | Research | GA + SA | Minimum of makespan, workload of most loaded machine, total workload of machines | 5 |
| Geyik and Dosdoğru (2013) | Research | GA + simulation. Dynamic FJSSP | Total of average flow times | 1 |
| Araghi et al. (2013) | Research | GA + VNS with affinity function. FJSSP with sequence-dependent setup times | Makespan | 2 |
| Barzegar and Motameni (2013) | Research | Gravitational search (GS) algorithm + PSO | Makespan | _ |

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Table 12 Continued

| Article | Application | Algorithm and shop details | Objective function | Citations |
|------------------------|-------------|--|--|-----------|
| Yuan and Xu (2013b) | Research | Harmony search (HS) + large neighborhood search (LNS) to improve solutions found by HS | Makespan | 7 |
| Yuan et al. (2013) | Research | HHS + local search | Makespan | 17 |
| Farughi et al. (2013) | Research | Memetic algorithm (MA) + critical path method. FJSSP with overlapping operations | Makespan | 1 |
| Wang et al. (2013a) | Research | Pareto-based estimation of distribution algorithm (EDA) + local search strategy based on critical path | Minimum of makespan, workload of most loaded machine, total workload of machines | 15 |
| He et al. (2013) | Research | Pareto-optimal based immune clone algorithm with Nash equilibrium. FJSSP with machine breakdown | Minimum of makespan, stability based on end time of each operation and stability based on machine allocation for each operation | _ |
| Li et al. (2014) | Research | ABC + TS. FJSSP with maintenance activities | Minimum of makespan, workload of most loaded machine, total workload of machines | 9 |

Table 16 while remaining 34% papers appearing in 64 journals have a cumulative impact factor of 38.27.

- 4. Most of the researchers focused on the algorithm development as compared to application in real-world industrial problems. Of 191 journal papers, 179 papers (93.72%) were pure research oriented while only 12 papers (6.28%) addressed the real-world industrial applications.
- 5. The research articles addressed both single-objective and multiobjective performance measures to evaluate the schedule. A total of 53.09% of the papers addressed single objective while 46.91% were multiobjective performance measures.
- 6. Minimization of makespan or total completion time had been the most widely used performance measure. Makespan was used in a total of 166 papers, that is, 84.26%. Of these 166 papers, 88 papers (44.67%) of all papers addressed makespan as the single performance measure while 78 papers used makespan in combination with other performance measures. Among multiobjective performance measures minimization of makespan, workload of most loaded machine, and total workload of machines have been the most widely used performance measure with 23.86% of the papers using this performance measure.

Table 13
Application of miscellaneous techniques in FJSSP

| Article | Application | Algorithm and shop details | Objective function | Citations |
|-------------------------|-------------|--|---|-----------|
| Jansen et al. (2005) | Research | Approximation Algorithm. Job release & delivery times, pre-emption | Makespan | 44 |
| Xing et al. (2009a) | Research | Efficient search method | Minimum of makespan, workload of most loaded machine, total workload of machines | 62 |
| Xing et al. (2009b) | Research | Simulation | Minimum of makespan, workload of most loaded machine, total workload of machines | 86 |
| Hmida et al. (2010) | Research | Climbing depth-bound discrepancy search (CDDS) approach based on ordering heuristics, involves two types of discrepancies, operation selection and resource allocation, and uses the block notion to build neighborhood structures that define relevant variables on which discrepancies are applied | Makespan | 39 |
| Barzegar et al. (2012b) | Research | Gravitational Search (GS) | Makespan | 2 |
| Sharma (2013) | Research | Attribute-oriented data mining technique | Minimum of makespan, workload of most loaded machine, total workload of machines | _ |
| He and Sun (2013) | Research | Idle time insertion and route changing strategy combined with right-shift policy. FJSSP with machine breakdown, route changing, and right-shift strategies | Makespan | 12 |

- 7. None of the research claims their technique/method is superior to others in the domain of FJSSP. Almost 80% of the citations, that is, 6809 of total of 8545 were attributable to only three techniques namely: hybrid, EAs, and TS.
- 8. Hybrid techniques/methods have been the most popular methods with almost 35.03% of the publications relying on hybrid techniques. Hybrid methods have been used to take advantage

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Table 14 Country-wise distribution of research papers in journals

| Country | Total |
|---|---------|
| China | 59 |
| Iran | 44 |
| France | 9 |
| India | 8 |
| Turkey, Tunisia | 7 |
| Canada, South Korea | 5 each |
| Brazil, Germany, Italy, Singapore, Taiwan | 04 each |
| Japan, Spain, USA | 03 each |
| Poland, Portugal | 02 each |
| Argentina, Denmark, Egypt, Hong Kong, Hungary, Indonesia, Israel, Malaysia, | 01 each |
| Pakistan, Romania, Saudi Arabia, South Africa, Switzerland, Thailand | |

Table 15 Country-wise distribution of research papers other than journals

| Country | Total |
|--|--------|
| China | 109 |
| France | 17 |
| Poland | 9 |
| USA | 8 |
| Japan, Singapore | 7 each |
| Iran, Turkey | 6 each |
| India, Malaysia, Taiwan, Thailand, Tunisia | 4 each |
| Belgium, Canada, Germany | 3 each |
| Portugal, Sweden | 2 each |
| Brazil, Denmark, Egypt, Holland, Indonesia, Italy, Oman, Pakistan, Saudi Arabia, South Africa, Spain | 1 each |

of the strengths of each of the methods to find better solutions. A total of 3572 citations for 69 papers have been noted addressing hybrid techniques.

- 9. Second-most popular technique in terms of total number of papers and citations is EAs. Various EA techniques were used in 23.86% of publications. A total of 1940 citations using EAs toward problem resolution have been noted.
- 10. Among all techniques used by various researchers whether hybrid or pure, GAs have been the most popular technique, used by 34% of the publications. A total of 2660 citations have been recorded for the publications that have used GAs as a standalone or combination technique. Second-most popular technique is TS with total of 2150 citations.
- 11. Most of the research papers have addressed simple FJSSP, while only 70 papers (35.53%) considered different scenarios such as setup and transportation times, maintenance, machine breakdown, job/machine ready times, fuzzy/uncertain processing times, overlapping operations, and re-entrant flexible job shop. In these real-like situations GA has been the most

Table 16 Paper distribution in different journals

| Journal | Number of papers | Impact factor |
|---|------------------|---------------------------|
| International Journal of Production Research | 28 | 1.323 |
| International Journal of Advanced Manufacturing Technology | 17 | 1.779 |
| Journal of Intelligent Manufacturing | 10 | 1.142 |
| Computers & Industrial Engineering | 9 | 1.690 |
| International Journal of Production Economics | 8 | 2.081 (2012) ^a |
| Computers & Operations Research | 7 | 1.718 |
| Applied Soft Computing | 6 | 2.679 |
| Applied Mathematical Modeling | 5 | 2.158 |
| European Journal of Operational Research | 5 | 1.843 |
| Annals of Operations Research | 4 | 1.103 |
| Expert Systems with Applications | 4 | 1.965 |
| IEEE Transactions on Systems, Man, and Cybernetics, Part C | 3 | 1.526 |
| Journal of Applied Sciences | 3 | _ |
| Journal of Information and Computational Science | 3 | = |
| International Journal of Computers Communications & Control | 2 | 0.694 |
| Journal of Manufacturing Systems | 2 | 1.847 |
| Knowledge-Based Systems | 2 | 3.058 |
| Mathematical Problems in Engineering | 2 | 1.082 |
| Mathematics and Computers in Simulation | 2 | 0.856 |
| Production Planning & Control | 2 | 0.991 |

^aTitle suppressed in 2013 JCR (http://admin-apps.webofknowledge.com/JCR/static_html/notices/notices.htm).

popular technique used by 24 researchers with total of 749 citations. Among these 24 papers, seven papers used GA in combination with other techniques.

7. Conclusions

The study presented here considered the review of solution techniques/methods published in the literature to solve FJSSPs. The paper presents a critical and comprehensive overview of the research trends in this area.

FJSSPs are an extension of typical JSS problems and are considered difficult to solve due to their NP-hard nature. Researchers and practitioners have tried to develop efficient solution techniques/methods during the last 25 years. With the advancement in computational power, the techniques/methods to solve FJSSP have become more and more efficient and powerful. Metaheuristics have been used more widely as compared to other methods. The most popular techniques have been hybrid methods as these techniques take advantage of the strengths of each of the methods to find better solutions. However, among all techniques, genetic algorithms have been the most widely used in this domain.

In the study, a range of methods have been surveyed. None of the methods has been adjudged the best by any researchers. Most of the work is focused on testing the developed algorithm on benchmark or generated problems. Relatively less work has been reported on practical problem solutions as compared to pure research.

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