



Techniques to Restrict an Interval of a Lower Bound in Fuzzy Scheduling Problems

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Article information

Article history:

Received 2 September 2022 ,
Accepted 30 November 2022 ,
Available online 1 June 2023 ,

Keywords:

Fuzzy Processing Times, Fuzzy
Due dates, Maximum Fuzzy
Tardiness, Lower Bounds, Efficient
solutions.

Abstract

In this paper, a fuzzy scheduling problem with triangular fuzzy numbers for processing times and due dates is provided. Each of the n operations is to be performed on a single machine without interruption and then becomes ready for processing at time zero. We generalized some ideas by giving a definition and a theorem to find a relation between the fuzzy lower bound and the fuzzy optimal solution for a problem with number of efficient solutions that minimizes total fuzzy completion time and maximum fuzzy tardiness. Also the results show how to choose the best defuzzification method in fuzzy scheduling

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DOI: [10.33899/IQJOSS.2023.178678](https://doi.org/10.33899/IQJOSS.2023.178678) , ©Authors, 2023, College of Computer and Mathematical Science, University of Mosul.

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1. Introduction

Fuzzy multi-objective scheduling problems are mainly focused on two criteria and they occur often in the real life situations. Solving multi-objective functions is difficult in the sense that we must deal with two various objectives without possessing any prior knowledge of their respective importance. Finding efficient solutions (Pareto set) is one of the approaches that solves this type of problems. From this set the decision maker will select one of these solutions (Heide and David, 2008). In the literature the Pareto set has been studied in the field of the optimization theory by many authors. In convex optimization, Ward looked at the construction of efficient sets (Ward, 1989) Lowe et al. Characterized the set of quasi-efficient solutions to a multiple objective problem (Lowe et al., 1984). For the minimum spanning tree problem, Steiner and Radzik determined all efficient solutions (Steiner and Radzik, 2008). In scheduling problems, the paper of Van Wassenhove and Gelders was the first one that dealt with finding the efficient solutions (Van Wassenhove and Gelders, 1980). Zinchenko studied the structure of Pareto set of some vector problems in scheduling (Zinchenko, 2002). Using the standard boundary intersection technique, Jia and Ierapetritou found Pareto optimum solutions for scheduling difficulties (Jia and Ierapetritou, 2007)). For equal processing times Lazarev et al. Found the Pareto set for

jobs with respect of two criteria (Lazarev, 2015), and for a shop scheduling problem Nguyen and Bao computed an efficient solution by using genetic algorithm (Nguyen and Bao, 2016). Some papers are related to this direction, starting with the non-fuzzy case Jabbar and Ramadhan firstly introduced such relation. Later Ramadhan and Begard introduced a relation regarding the bi-criteria problem namely maximum tardiness and maximum earliness (Amin and Ramadhan, 2021). Hassan et al., generalized the idea to three criteria to minimize the sum of total completion time, maximum earliness and maximum tardiness (Dara et al., 2022). In the fuzzy environment Ramadhan presented the same idea to minimize the sum of total fuzzy completion time and maximum earliness in addition to find all the efficient solutions (Ramadhan, 2021).

The focus of this paper is on the structure of efficient solutions where The processing times and deadlines are triangular, fuzzy numbers. The multi-objective problem is to minimize total fuzzy completion time and maximum fuzzy tardiness. We introduce a new definition for fuzzy numbers which is called q -strongly positive fuzzy numbers, and a theorem which finds a relation between the fuzzy lower bound and the fuzzy optimal solution with number of efficient solutions. This relation restricts the fuzzy lower bound through number of efficient solutions.

1. Definitions and Notations

Definition 1: A triangular fuzzy number $\tilde{K} = (k^L, k, k^U)$ can be represented by three components, where k^L represents the lower bound of the fuzzy number, k^U the upper bound for the number. $\mu_{\tilde{K}}(x)$ specifies a membership function for a triangular fuzzy number $\tilde{K}$, where

$$\mu_{\tilde{K}}(x) = \begin{cases} 0 & , \quad x < k^L, \\ \frac{x-k^L}{k-k^L} & , \quad k^L \leq x \leq k, \\ \frac{k^U-x}{k^U-k} & , \quad k < x \leq k^U. \end{cases}$$

Assume $\tilde{K}$ and $\tilde{G}$ be two triangular fuzzy numbers, with $\tilde{K} = (k^L, k, k^U)$ and $\tilde{G} = (g^L, g, g^U)$, then the addition of two fuzzy numbers is

$$3) \quad \tilde{K} \oplus \tilde{G} = (k^L + g^L, k + g, k^U + g^U), \text{ which is also triangular fuzzy number,}$$

and the subtraction is

$$ii) \quad \tilde{K} \ominus \tilde{G} = (k^L - g^U, k - g, k^U - g^L), \text{ which is also triangular fuzzy number (Hsien, 2010).}$$

Definition 2: The procedure that converts a fuzzy number to its crisp value is called defuzzification. For a triangular fuzzy number $\tilde{K} = (k_1, k_2, k_3)$. A triangular fuzzy number's centroid point is $D(\tilde{K}) = \frac{k_1 + k_2 + k_3}{3}$ (Cheng, 1998).

Consider two triangular fuzzy production time $\tilde{p}_1$ and $\tilde{p}_2$ (due dates) where $\tilde{p}_1 = (p_1^L, p_1, p_1^U)$ and $\tilde{p}_2 = (p_2^L, p_2, p_2^U)$. Using this method we say that $\tilde{p}_1 < \tilde{p}_2$ if $D(\tilde{p}_1) < D(\tilde{p}_2)$. A special case will occur when $p_1^U < p_2^L$, in this case the two numbers are comparable and there is no need to use ranking methods to map them to crisp values. The majority of defuzzification procedures result in a rational number, So, let d be the rational number's denominator, which plays an important part in this paper. Consider a problem p with two any criteria f and g to be minimized simultaneously, then

Definition 3: A sequence $\pi^* \in \Pi$ is efficient solution for p if $\exists$ no $\pi \in \Pi$ s.t. $f(\pi) \leq f(\pi^*)$ and $h(\pi) \leq h(\pi^*)$, where at least one relation holds with strict inequality.

Definition 4: A set of all the efficient solutions for a problem is called Pareto set.

The notations used in this paper are as follows:

$N: \{1, 2, 3, \dots, n\}$,

Π : all possible schedules,

$\tilde{p}_j = (p_j^L, p_j, p_j^U)$: fuzzy processing time for job j , and they are triangular fuzzy number (TFN),

$\tilde{d}_j = (d_j^L, d_j, d_j^U)$: fuzzy due date for job j , and they are triangular fuzzy number (TFN),

$\tilde{c}_j$: fuzzy completion time for job j ,

$\tilde{L}_j = \tilde{c}_j \ominus \tilde{d}_j$: fuzzy lateness for job j ,

$\tilde{T}_j = \tilde{\max}\{\tilde{c}_j \ominus \tilde{d}_j, \tilde{0}\}$: fuzzy tardiness of job j , where $\tilde{0} = (0, 0, 0)$,

$\tilde{T}_{\max} = \tilde{\max}\{\tilde{T}_j\}$: maximum fuzzy tardiness,

K^c : the crisp value of the fuzzy number $\tilde{K}$,

d : the denominator of a rational number,

EDD (early due date): tasks are arranged in ascending order of d_j ,

SPT (Shortest processing time): tasks are sequenced in ascending order of p_j ,

The fuzzy lower bound ($\tilde{LB}$) is a value of the objective function that is less than or equal to the fuzzy optimum value.

$\tilde{UB}$ (Fuzzy upper bound): an objective function value larger than or equal to the fuzzy optimum value.

2. Background of the Problem

A set N of n jobs to be processed on a one- machine. Each one has a fuzzy processing time $\tilde{p}_j$ which is a triangular fuzzy number $\tilde{p}_j = (p_j^L, p_j, p_j^U)$, and a triangular fuzzy due date $\tilde{d}_j = (d_j^L, d_j, d_j^U)$, for $j = 1, 2, \dots, n$. At time zero, all jobs are accessible, and the machine can only process one task at a time, and a job's execution cannot be stopped. A schedule is made by placing tasks in a certain sequence such that the fuzzy completion time $\tilde{c}_j$ of each job j may be calculated. In fact, employment processing timelines are unpredictable. As a result, each job's completion time is unknown.

Smith proposed an approach for solving a single machine scheduling issue that reduced overall completion time while ensuring that all tasks were finished on time (Smith, 1965). Van Wassenhove extended the idea to find the Pareto set of the simultaneous problem $1/F(\gamma^1, \gamma^2)$ which is a function of two cost criteria where $\gamma^1 = \sum_{j=1}^n c_j$ and $\gamma^2 = T_{\max}$ and without constraints on jobs. The solution of this problem is difficult and sometimes is not possible, this means, there is in general no π which minimizes γ^1 and γ^2 , as a result, we're looking for a sequence that provides a fair solution to both goals. (if such a sequence exists). To define such a sequence, Van Wassenhove and Gelders introduced the concept of efficiency in scheduling problems (Van Wassenhove and Gelders, 1980).

3. Needed Calculations

Consider a scheduling problem with n -jobs with one machine and processing times that are considered to be fuzzy. Let $\tilde{p}_j$ be the job j fuzzy processing time. The following formulas can be used to compute the fuzzy completion time of jobs:

$$\tilde{c}_1 = \tilde{p}_1,$$

$$\tilde{c}_2 = \tilde{c}_1 \oplus \tilde{p}_2, \tag{1}$$

$$\tilde{c}_j = \tilde{c}_{j-1} \oplus \tilde{p}_j, \text{ for } j = 1, \dots, n.$$

A penalty is charged if a task is done beyond its due date; nevertheless, if a job is performed before its due date, it is deemed early (Hsien, 2010). The difference between the fuzzy completion time and the fuzzy due date of this job is a fuzzy maximum of zero, and the fuzzy tardiness of a work in a given sequence is a fuzzy maximum of zero, which implies

$$\tilde{T}_j = \widetilde{\max}\{\tilde{c}_j \ominus \tilde{d}_j, \tilde{0}\}, \quad (2)$$

where the fuzzy completion time $\tilde{c}_j = (c_j^L, c_j, c_j^U)$, (1) may be achieved for each job j , and maximum fuzzy tardiness is

$$\tilde{T}_{\max} = \widetilde{\max}\left\{\widetilde{\max}\{\tilde{c}_j \ominus \tilde{d}_j, \tilde{0}\}\right\} \quad (3)$$

4. Efficient Solutions and Optimal Solution

With the number of effective solutions, Jabbar and Ramadhan discovered a relationship between the lower bound and the optimization method for a problem which was minimizing total completion time and maximum tardiness (Jabbar and Ramadhan, 2006). In the case where all the inputs data are are fuzzy numbers, we generalized the case by introducing new definition. The problem is

$$1/F(\gamma^1, \gamma^2), \quad (4)$$

where $\gamma^1 = \sum_{j=1}^n \tilde{c}_j$ and $\gamma^2 = \tilde{T}_{\max}$. This problem is in simultaneously form and has efficient solutions, one of the efficient solutions will be fuzzy optimal for the sum of the problem, i.e.,

$$1/(\sum_{j=1}^n \tilde{c}_j + \tilde{T}_{\max}). \quad (5)$$

For the problem (5) let the fuzzy lower bound $\tilde{LB} = \sum_{j=1}^n \tilde{c}_j(\text{SPT}) \oplus \tilde{T}_{\max}(\text{EDD})$, the fuzzy upper bound $\tilde{UB} = \sum_{j=1}^n \tilde{c}_j(\text{SPT}) \oplus \tilde{T}_{\max}(\text{SPT})$ and $\widetilde{\text{opt}}$ be the fuzzy optimal value. To find our results we introduce the following.

Definition 5: If $(k^L + k + k^U) - (g^L + g + g^U) \geq m$, where m may be any positive integer greater than or equal to one, two fuzzy numbers $\tilde{K} = \{k^L, k, k^U\}$ and $\tilde{G} = \{g^L, g, g^U\}$ are q -strongly positive.

Think of the sets of maximum fuzzy tardiness $S = \{\tilde{T}_{\max}(\pi_i)\}$ and total fuzzy completion time $S_1 = \{\sum_{j=1}^n \tilde{c}_j(\pi_i)\}$, where π_i is an efficient solution for each $i = 1, \dots, k$, and both sets have two components that are q -strongly positive. Implying that $\tilde{T}_{\max}(\pi_i)$ and $\tilde{T}_{\max}(\pi_{i+1})$ are q -strongly positive numbers, and $\sum_{j=1}^n \tilde{c}_j(\pi_{i+1})$ and $\sum_{j=1}^n \tilde{c}_j(\pi_i)$ are q -strongly positive numbers too.

Theorem 1 (New): If the problem has 3-strong positive numbers efficient solutions for problem (4), then there exists a fuzzy number $\tilde{r}$ such that $\tilde{LB} \oplus \tilde{r} = \widetilde{\text{opt}}$ and $r^c \in [Q_1 - 1, (\tilde{Q}_2 \oplus \frac{1}{d})^c]$ where Q_1 =number of efficient solutions and $\tilde{Q}_2 = \tilde{T}_{\max}(\text{SPT}) \ominus \tilde{T}_{\max}(\text{EDD})$.

Proof. Since $\tilde{LB} \leq \widetilde{\text{opt}}$, so there exists $\tilde{r}$ such that $\tilde{LB} \oplus \tilde{r} = \widetilde{\text{opt}}$ this proves the first part. Now, to show that $r^c \in [Q_1 - 1, (\tilde{Q}_2 \oplus \frac{1}{d})^c]$ or $Q_1 - 1 \leq r^c \leq (\tilde{Q}_2 \oplus \frac{1}{d})^c$. We have $\tilde{r} = \widetilde{\text{opt}} \ominus \tilde{LB} \leq \tilde{UB} \ominus \tilde{LB}$

$$= \tilde{T}_{\max}(\text{SPT}) \ominus \tilde{T}_{\max}(\text{EDD})$$

$$= \tilde{Q}_2 \leq \tilde{Q}_2 \oplus \frac{1}{d},$$

$$\text{implies that } r^c \leq (\tilde{Q}_2 \oplus \frac{1}{d})^c.$$

For $Q_1 - 1 \leq r^c$ use mathematical induction on Q_1 .

If $Q_1 = 1$, In other words, there is just one effective solution, which is SPT then

$$\tilde{r} = \sum_{j=1}^n \tilde{c}_j(\text{SPT}) \oplus \tilde{T}_{\max}(\text{SPT}) \ominus \sum_{j=1}^n \tilde{c}_j(\text{SPT}) \ominus \tilde{T}_{\max}(\text{EDD}),$$

$$= \tilde{T}_{\max}(\text{SPT}) \ominus \tilde{T}_{\max}(\text{EDD}), \text{ since } \tilde{T}_{\max}(\text{SPT}) \text{ and } \tilde{T}_{\max}(\text{EDD}) \text{ are equal, so } r^c = 0.$$

Thus $r^c = 0 = Q_1 - 1$, which proves the case $Q_1 = 1$.

If $Q_1 = 2$, Such that, there are only two effective solutions, that are SPT and σ , instance. Since $Q_1 = 2$, so $Q_1 - 1 = 1$. The following are two scenarios:

3- If SPT is optimal then

$$\begin{aligned}\tilde{r} &= \sum_{j=1}^n \tilde{c}_j(\text{SPT}) \oplus \tilde{T}_{\max}(\text{SPT}) \ominus \sum_{j=1}^n \tilde{c}_j(\text{SPT}) \ominus \tilde{T}_{\max}(\text{EDD}), \\ &= \tilde{T}_{\max}(\text{SPT}) \ominus \tilde{T}_{\max}(\text{EDD}), \text{ since } \tilde{T}_{\max}(\text{SPT}) \text{ and } \tilde{T}_{\max}(\text{EDD}) \text{ are 3-strongly positives, so the difference between } \\ &\tilde{T}_{\max}^c(\text{SPT}) \text{ and } \tilde{T}_{\max}^c(\text{EDD}) \text{ is } \geq 1. \text{ Thus } r^c \geq 1 \geq Q_1 - 1, \text{ As a result, the theorem holds for } Q_1 = 2.\end{aligned}$$

b- If σ is optimal then

$$\begin{aligned}\tilde{r} &= \sum_{j=1}^n \tilde{c}_j(\sigma) \oplus \tilde{T}_{\max}(\sigma) \ominus \sum_{j=1}^n \tilde{c}_j(\text{SPT}) \ominus \tilde{T}_{\max}(\text{EDD}), \\ &= \sum_{j=1}^n \tilde{c}_j(\sigma) \ominus \sum_{j=1}^n \tilde{c}_j(\text{SPT}), \text{ since they are 3-strongly positives, so } r^c \geq 2 \geq Q_1 - 1, \\ &\text{so it is true for } Q_1 = 2.\end{aligned}$$

If $Q_1 = 3$, means we have three efficeient solutions namelly, SPT, σ and σ_1 . Since $Q_1 = 3$, so $Q_1 - 1 = 2$. We have the following three cases:

3- If SPT is optimal then

$$\begin{aligned}\tilde{r} &= \sum_{j=1}^n \tilde{c}_j(\text{SPT}) \oplus \tilde{T}_{\max}(\text{SPT}) \ominus \sum_{j=1}^n \tilde{c}_j(\text{SPT}) \ominus \tilde{T}_{\max}(\text{EDD}), \\ &= \tilde{T}_{\max}(\text{SPT}) \ominus \tilde{T}_{\max}(\text{EDD}), \text{ since } \tilde{T}_{\max}(\text{SPT}) \text{ and } \tilde{T}_{\max}(\text{EDD}) \text{ are 3-strongly positives, so the difference between } \\ &\tilde{T}_{\max}^c(\text{SPT}) \text{ and } \tilde{T}_{\max}^c(\text{EDD}) \text{ is } \geq 2. \text{ Thus } r^c \geq 2 \geq Q_1 - 1, \text{ so it is true for } Q_1 = 3.\end{aligned}$$

b- If σ is optimal then

$$\begin{aligned}\tilde{r} &= \sum_{j=1}^n \tilde{c}_j(\sigma) \oplus \tilde{T}_{\max}(\sigma) \ominus \sum_{j=1}^n \tilde{c}_j(\text{SPT}) \ominus \tilde{T}_{\max}(\text{EDD}), \\ &= \sum_{j=1}^n \tilde{c}_j(\sigma) \ominus \sum_{j=1}^n \tilde{c}_j(\text{SPT}) \oplus \tilde{T}_{\max}(\sigma) \ominus \tilde{T}_{\max}(\text{EDD}), \text{ since each of the difference is 3-strongly positive, so } r^c \geq \\ &2 \geq Q_1 - 1, \text{ so it is true for } Q_1 = 3.\end{aligned}$$

c- If σ_1 is optimal then

$$\begin{aligned}\tilde{r} &= \sum_{j=1}^n \tilde{c}_j(\sigma_1) \oplus \tilde{T}_{\max}(\sigma_1) \ominus \sum_{j=1}^n \tilde{c}_j(\text{SPT}) \ominus \tilde{T}_{\max}(\text{EDD}), \\ &= \sum_{j=1}^n \tilde{c}_j(\sigma_1) \ominus \sum_{j=1}^n \tilde{c}_j(\text{SPT}) \oplus \tilde{T}_{\max}(\sigma_1) \ominus \tilde{T}_{\max}(\text{EDD}), \\ &= \sum_{j=1}^n \tilde{c}_j(\sigma_1) \ominus \sum_{j=1}^n \tilde{c}_j(\text{SPT}), \text{ since they are 3-strongly positives, so } \\ &r^c \geq 2 \geq Q_1 - 1, \text{ so it is true for } Q_1 = 3.\end{aligned}$$

Assume that the theorem holds for $Q_1 = k$, such that, for the k most efficient solutions SPT, σ , σ_1 , ..., σ_{k-2} . Let $Q_1 = k + 1$, that means, there are $k + 1$ efficient solutions SPT, σ , σ_1 , ..., σ_{k-2} , σ_{k-1} . Whether any of the first k optimal processes is the optimum, and the theorem holds fo $Q_1 = k$ then we get $Q_1 - 1 \leq r^c$.

If σ_{k-1} is the most efficient solution, then

$$\tilde{r} = \sum_{j=1}^n \tilde{c}_j(\sigma_{k-1}) \oplus \tilde{T}_{\max}(\sigma_{k-1}) \ominus \sum_{j=1}^n \tilde{c}_j(\text{SPT}) \ominus \tilde{T}_{\max}(\text{EDD}),$$

implies that

$$\tilde{r} = \sum_{j=1}^n \tilde{c}_j(\sigma_{k-1}) \ominus \sum_{j=1}^n \tilde{c}_j(\text{SPT}), \text{ and then } r^c \geq k. \text{ Thus it is true for } Q_1 = k + 1. \blacksquare$$

Corollary 1: If the efficient solutions are not 3-strongly positive for the problem (4), then $\exists$ a fuzzy number $\tilde{r}$ s.t. $\tilde{LB} \oplus \tilde{r} = \tilde{opt}$ and $r^c \in [\frac{Q_1-1}{d}, (\tilde{Q}_2 \oplus \frac{1}{d})^c]$ where Q_1 = number of efficient solutions and $\tilde{Q}_2 = \tilde{T}_{max}(SPT) \ominus \tilde{T}_{max}(EDD)$.

Proof. We prove only the second part which is $Q_1 - 1 \leq r^c$.

This will be done by the same way of the above theorem. Since they are not q-strongly positive, so the difference between each of any two one is less than m. If $m = 1$ which is the worst case, then $r^c \geq K_i - 1, \forall i, i = 1, \dots, k$.

It's vital to note that the q and d values are the same, so the theorem depends strongly on the defuzzification method. ■

To illustrate the theorem we give here an example.

Consider the following data. For simple calculation we have the due dates as cris number.

J	1	2	3
$\tilde{p}_j$	(2, 3, 4)	(4, 5, 7)	(4, 5, 9)
$\tilde{d}_j$	7	5	2

There are three efficient solutions for the problem, and by using the mentioned defuzzification method the efficient solutions for this problem are: sequence (1, 2, 3) with $\sum_{j=1}^3 \tilde{c}_j = (18, 26, 35)$ and $\tilde{T}_{max} = (8, 13, 18)$ which is SPT- rule, sequence (1, 3, 2) with $\sum_{j=1}^3 \tilde{c}_j = (18, 28, 37)$ and $\tilde{T}_{max} = (5, 10, 15)$, and sequence (3, 2, 1)) with $\sum_{j=1}^3 \tilde{c}_j = (22, 34, 45)$ and $\tilde{T}_{max} = (3, 8, 13)$. Now

$$\tilde{LB} = \sum_{j=1}^3 \tilde{c}_j(SPT) \oplus \tilde{T}_{max}(EDD) = (18, 26, 35) \oplus (3, 8, 13) = (21, 34, 48),$$

$$\tilde{UB} = \sum_{j=1}^3 \tilde{c}_j(SPT) \oplus \tilde{T}_{max}(SPT) = (18, 26, 35) \oplus (8, 13, 18) = (26, 39, 53),$$

$\tilde{opt} = (18, 28, 37) \oplus (5, 10, 15) = (23, 38, 52)$, it is the sum of one the efficient solutions for $\sum_{j=1}^n \tilde{c}_j \oplus \tilde{T}_{max}$ which is the sequence (1, 2, 3), and $\tilde{r} = \tilde{opt} - \tilde{LB} = (23, 38, 52) \ominus (21, 34, 48) = (-25, 4, 31)$, $\tilde{r} = \frac{10}{3}$. Q_1 = number of efficient solutions, $Q_1-1 = 3-1 = 2$, and $\tilde{Q}_2 = \tilde{T}_{max}(SPT) - \tilde{T}_{max}(EDD) = (8, 13, 18) \ominus (3, 8, 13) = (-5, 5, 15)$. $(\tilde{Q}_2 \oplus \frac{1}{d})^c = (\frac{-14}{3}, \frac{16}{3}, \frac{46}{3})^c = \frac{48}{3}$. So, for this example $\frac{10}{3} \in [2, \frac{48}{3}]$.

5. Conclusions

The fuzzy lower bound and the fuzzy optimum solution with the number of efficient solutions were discovered to have a significant relationship in this paper. This relationship conceptually explains the difference between a fuzzy optimum solution and a fuzzy lower bound, allowing new algorithms and approaches to be developed to discover heuristic solutions to these issues. Furthermore, the defuzzification approach has an important role in limiting the gap between the fuzzy optimum solution and the fuzzy lower bound. For a given theorem, multiple approaches provide different intervals.

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¹زارة التربية ، المديرية العامة لتربية كرميان ، معهد الاعداد والتطوير التربوي ، العراق.^{2,3}قسم الرياضيات ، كلية العلوم ، جامعة كرميان ، إقليم كردستان ، العراق.⁴قسم الرياضيات ، كلية العلوم ، جامعة السليمانية ، إقليم كردستان ، العراق

الخلاصة: في هذا البحث تم دراسة مسالة جدولة الماكنة الضبابية مع اعداد ضبابية مثلية لافقات البدء وافقات الانتهاء . كل عمل من الاعمال تنجز على ماكنة واحدة بدون انقطاع و تكون حاضرة للبدء في الوقت الصفري . عممنا بعض الافكار باعطاء تعريف ونظرية لايجاد علاقة بين الحد الادنى الضبابي والحل الامثل الضبابي للمسالة مع عدد الحلول الكفوءة والتي تصغر مجموع اوقات الاتمام الضبابي و اكبر تاخير ضبابي . ايضا النتائج تظهر كيفية اختيار افضل طريقة للترتيب في الجدولة الضبابية .

الكلمات المفتاحية: أوقات المعالجة الضبابية ، تواريخ الاستحقاق غير الواضحة ، الحد الأقصى للتأخير الضبابي ، الحدود الدنيا ، الحلول الفعالة .