

# Solving University Timetabling Problem Using Harmony Search: a Case Study in STMIK Mikroskil

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Harmony Search  
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## ABSTRACT

A University Timetabling Problem (UTP) belongs to NP-complete (Nondeterministic Polynomial-Complete) problem category since it has huge search space that consists of many students, classrooms, lectures, and courses permutation. Due to that, finding an optimal schedule using exact algorithm is impossible to be done under a reasonable time.

Harmony Search is a heuristic method that can solve NP-Complete problem such as UTP in reasonable time. However because of its heuristic nature, it can only provide solution which is close to the perfect solution. Therefore, to prove that a schedule is a feasible solution, several constraints have to be satisfied. This research used Harmony Search to solve the UTP whereby the dataset, constraints, and experiments are acquired and performed in STMIK Mikroskil.

This research also evaluates the speed performance and fitness value of Harmony Search using different parameter configuration. From its finding, the HMCR parameter value that is best suited for scheduling is in range of 0.7 to 0.95. The PAR value however, did not sufficiently contribute to the speed and fitness value.

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## 1. INTRODUCTION

A University Timetabling Problem (UTP) is a placement of a set of courses in predefined time period [1]. UTP consists of lecture and exam scheduling where both of them face common problem of finding a schedule over a limited time period while avoiding conflicts and satisfying a number of side-constraint [2]. UTP has long been known as NP-complete problem class [3]. NP-complete problem class can't be solved using exact algorithm since it has a huge search space whereby iterating each possible solutions can't be done under a reasonable time period. However, heuristic algorithms such as Genetic Algorithm [4], Particle Swarm Optimization [5], Harmony Search (HS) [6] and many others can be used to solve NP-complete problem. These heuristic algorithms can provide an optimal solution to UTP. HS is chosen as the preferred method of this research due to its performance that is better than other heuristic algorithms [6].

This research is performed in STMIK Mikroskil environment. All the data that are used as input to HS are collected from STMIK Mikroskil. The resulting schedule also have to conform to the constraints that are stated in the STMIK Mikroskil academic rules. The contribution from this research is majorly directed toward the academic improvement of STMIK Mikroskil and the evaluation of speed performance and fitness value of HS using different parameter configurations that are best suited for scheduling purpose.

## 2. LITERATURE REVIEW

University Timetabling Problem has been investigated in several researches such as Al-Betar et al. [7], Nguyen et al. [8], and Chen and Shih [9]. In Al-Betar et al. [7]'s work, HS is used with multi-pitch adjusting rate. Nguyen et al [8] used a hybrid algorithm of HS and Bees [10] algorithm. Chen and Shih [9]

used constriction Particle Swarm Optimization with local search. HS is chosen as the preferred method in solving UTP in STMIK Mikroskil due to its good performance in Socha Benchmark<sup>2</sup> [7].

HS is inspired by musician's music improvisation [11]. It follows the pattern of a musician during improvisation processes which are: (1) famous tune retrieval from his or her memory; (2) adjusting the pitch of the aforementioned tune and replays it again; (3) composing new or random notes. Geem [11] formalize those three components into: usage of harmony memory, pitch adjusting and randomization. The algorithm of HS is presented in Algorithm 1.

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**Algorithm 1** Harmony Search algorithm [11].

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1: begin
2:   Define objective function  $f(x)$ ,  $x=(x_1, x_2, \dots, x_d)^T$ 
3:   Define harmony memory accepting rate ( $r_{accept}$ )
4:   Define pitch adjusting rate ( $r_{pa}$ ) and other parameters
5:   Generate Harmony Memory with random harmonies
6:   while (  $t < \text{max number of iterations}$  )
7:     while (  $i \leq \text{number of variables}$  )
8:       if ( $\text{rand} < r_{accept}$ ), Choose a value from HM for the variable  $i$ 
9:       if ( $\text{rand} < r_{pa}$ ), Adjust the value by adding certain amount
10:      end if
11:    else Choose a random value
12:    end if
13:  end while
14:  Accept the new harmony (solution) if better
15: end while
16: Find the current best solution

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Geem et al. [6] have done benchmarking in HS with other heuristic algorithms by using Braken and McCormick [12]'s problem. The problem statement is as follows.

$$\text{Minimize } f(x) = (x_1 - 2)^2 + (x_2 - 1)^2$$

$$\text{Subject to } g_1(x) = 0$$

$$g_2(x) \geq 0$$

with

$$g_1(x) = x_1 - 2x_2 + 1$$

$$g_2(x) = -x_1^2/4 - x_2^2 + 1$$

The components  $x_1$  and  $x_2$  are both initialized randomly with uniform distribution over the range  $[-10.0, 10.0]$ . The results are then compared with results from other researches, such as: Homaifer et al. [13]'s comparison of using exact solution, Genetic Algorithm (GA) [4], and Generalized Reduced Gradient (GRG) [14]; and Fogel [15] using Evolutionary Programming [16].

Table 1 shows the comparison of results from various methods. HS shows good result with relative error between the algorithm results and the exact function value only +0.2153. Although GRG shows better result from the HS, however, it violates the inequality constraint.

Table 1. Comparison of Results from various methods for Braken and McCormick [12]'s problem [6]

	EXACT	GRG	GA	EP	HS (1)	HS (2)
$f(x)$	1.3935	1.3934	1.4339	1.3772	1.3771	1.3965
%	0.0000	-0.0072	+2.8992	-1.1697	-1.1769	+0.2153
$x_1$	0.82288	0.8229	0.8080	0.8350	0.8348	0.8290
$x_2$	0.91144	0.9115	0.8854	0.9125	0.9124	0.9080
$g_1$	$7.05 \times 10^{-9}$	$1.0 \times 10^{-4}$	$3.7 \times 10^{-2}$	$1.0 \times 10^{-2}$	$1.0 \times 10^{-2}$	$1.3 \times 10^{-2}$
$g_2$	$1.73 \times 10^{-8}$	$-5.2 \times 10^{-5}$	$5.2 \times 10^{-2}$	$-7.0 \times 10^{-3}$	$-6.7 \times 10^{-3}$	$3.7 \times 10^{-3}$

### 3. RESEARCH METHOD

This section presents a generalization of quantitative research method that are divided into three distinct phases which are: (1) data and constraint acquisition from STMIK Mikroskil, (2) harmony memory and HS parameters initialization, and (3) HS processing. The result evaluation is presented in the next section.

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<sup>2</sup> <http://iridia.ulb.ac.be/msampels/tt.data/>

### 3.1. Data and Constraint Acquisition Phase

STMIK Mikroskil have four departments comprises of two undergraduate degree (S-1), and two vocational degree (D-3). The undergraduate degree further comprises of Computer Science department and Information System department, while the vocational degree comprises of Informatics Management department and Computerized Accounting department. Each department have their own curriculum, lecturer, and student data. However, several lecturers can teach different departments as long as they have the course prerequisite and there is no conflict between two or more taken courses in lecture schedule.

Since STMIK Mikroskil have two lecture sessions which are morning and afternoon sessions, therefore there must be two schedules generated for each period. There are also hard constraints and soft constraints that have to be satisfied. The hard constraints and soft constraints can be seen in Table 2.

Table 2. Hard constraint and soft constraint specification for scheduling criterias

Name of Constraint	Morning	Afternoon
<i>Hard Constraint</i>		
Each lecturer can only teach for maximum 18 units in one semester.	✓	✓
Each lecturer can not teach two courses in the same class.	✓	✓
Each lecturer can only teach for maximum two different courses in one semester	✓	✓
Afternoon classes must have at least one day schedule free.	✗	✓
<i>Soft Constraint</i>		
In assigning lecturer, homebased lecturers have priority over other lecturers.	✓	✓
Monday until Friday are given higher priorities over other Saturday.	✓	✗
It's preferable that lecturer can teach five units in one day.	✓	✓
It's preferable that student can take classes with total of five units in one day.	✓	✓
For religious purpose, it's preferable that on Friday student can take classes with total of three units where units allocation is on the first until the third hours of day.	✓	✗

### 3.2. Harmony Memory and HS Parameters Initializing Phase

In this phase, Harmony Memory (HM) is initialized. The first step in initialization is to determine the HM size. The second step is to convert lectures, courses, available days, and status of lecturers (homebased or non-homebased) dataset into several matrixes. The following is the matrixes of Harmony Memory.

$$\text{Lecturer\_matrix} = \begin{bmatrix} \text{lid}_1 & \text{lname}_1 & c_1 & s_1 & \text{dtm}_1 & \text{dta}_1 \\ \text{lid}_2 & \text{lname}_2 & c_2 & s_2 & \text{dtm}_2 & \text{dta}_2 \\ \text{lid}_3 & \text{lname}_3 & c_3 & s_3 & \text{dtm}_3 & \text{dta}_3 \\ \dots & \dots & \dots & \dots & \dots & \dots \\ \text{lid}_n & \text{lname}_n & c_n & s_n & \text{dtm}_n & \text{dta}_n \end{bmatrix}$$

where:

lid<sub>i</sub> = i<sup>th</sup> lecture id

lname<sub>i</sub> = i<sup>th</sup> lecture name

c<sub>i</sub> = course that can be taught by i<sup>th</sup> lecture

s<sub>i</sub> = status of i<sup>th</sup> lecture

dtm<sub>i</sub> = date and time that is available to i<sup>th</sup> lecture in morning session

dta<sub>i</sub> = date and time that is available to i<sup>th</sup> lecture in afternoon session

$$\text{Semester} = \begin{bmatrix} [\text{cid}_{11} & c_{12} & \text{sum}_{13}] & \dots & [\text{cid}_{1n} & c_{1n} & \text{sum}_{1n}] & s_1 & \text{sm}_1 & \text{sa}_1 \\ [\text{cid}_{21} & c_{22} & \text{sum}_{23}] & \dots & [\text{cid}_{2n} & c_{2n} & \text{sum}_{2n}] & s_2 & \text{sm}_2 & \text{sa}_2 \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ [\text{cid}_{n1} & c_{n2} & \text{sum}_{n3}] & \dots & [\text{cid}_{nn} & c_{nn} & \text{sum}_{nn}] & s_n & \text{sm}_n & \text{sa}_n \end{bmatrix}$$

where:

cid<sub>ij</sub> = j<sup>th</sup> course id of i<sup>th</sup> semester

c<sub>ij</sub> = j<sup>th</sup> course name of i<sup>th</sup> semester

sum<sub>ij</sub> = sum of j<sup>th</sup> course unit of i<sup>th</sup> semester

s<sub>i</sub> = i<sup>th</sup> semester

sm<sub>i</sub> = sum of morning classes of i<sup>th</sup> semester

sa<sub>i</sub> = sum of afternoon classes of i<sup>th</sup> semester

$$\text{Department\_matrix} = [\text{dcm}_1 \quad \text{dcm}_2 \quad \dots \quad \text{dcm}_n]$$

where:

dcm<sub>i</sub> = course of i<sup>th</sup> department

$$\text{Class\_matrix} = \begin{bmatrix} [\text{cc}_{11} & \text{lc}_{12}] & \dots & [\text{cc}_{1n} & \text{lc}_{1n}] & \text{tc}_1 & \text{dcm}_1 & \text{cc}_1 & \text{lc}_1 \\ [\text{cc}_{21} & \text{lc}_{22}] & \dots & [\text{cc}_{2n} & \text{lc}_{2n}] & \text{tc}_2 & \text{dcm}_2 & \text{cc}_2 & \text{lc}_2 \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ [\text{cc}_{n1} & \text{lc}_{n2}] & \dots & [\text{cc}_{nn} & \text{lc}_{nn}] & \text{tc}_n & \text{dcm}_n & \text{cc}_n & \text{lc}_n \end{bmatrix}$$

where:

cc<sub>ij</sub> = j<sup>th</sup> course of i<sup>th</sup> class

lc<sub>ij</sub> = j<sup>th</sup> lecturer of i<sup>th</sup> class

tc<sub>i</sub> = lecturing time of i<sup>th</sup> class (morning and afternoon)

$dc_i$  = department of  $i^{th}$  class  
 $sc_i$  = semester of  $i^{th}$  class  
 $c_i$  =  $i^{th}$  class

For the two parameter of HS, namely: HMCR and PAR, this research use the recommended value as seen in Geem [11]'s work. The recommended value for HMCR is between 0.1 to 0.4 while for PAR is between 0.7 to 0.95.

### 3.3. Harmony Search Processing Phase

In this phase, each class of morning and afternoon session schedule are generated through the improvisation process of HS. Steps of the improvisation process is as follows.

1. Generate random value between 0.0 to 1.0. The value will be compared to HMCR value. If it is higher or equal with HMCR, then a new random schedule is generated. However, if it is smaller than HMCR then randomly take the previous schedule in HM.
2. Generate another random value between 0.0 to 1.0. The value will be compared to PAR value. If it is higher or equal with PAR, then check whether there are collision in the schedule. When there are no conflict detected in the schedule, the schedule will be used as the new solution. If it is lower than PAR then the schedule will be adjusted in accordance to the bandwidth value. Bandwidth value is generated by random with value from 0 to the number of available days.
3. If all the schedule in Harmony Memory have been used and all of them contain conflict then, a new schedule will be regenerated and return to step 1.
4. However, if the schedule have reached termination criterias, the improvisation process will be stopped. There are two termination criterias available: (1) if the generated schedule have satisfied all the hard constraint, or (2) have run 500 iteration.

## 4. RESULTS AND ANALYSIS

This section presents the results that were obtained and the analysis of those results. Since HS has two parameters (HMCR and PAR), this research also evaluate which configuration of parameter is the best for scheduling purpose. For evaluation, the program is run 25 times where each run has 500 iteration termination criteria. The academic data that are used in each run is as follows.

1. Three department data are used, namely Computer Science, Information System, and Computerized Accounting department.
2. Computer Science department has 13 classes in morning session and 9 classes in afternoon session.
3. Information System department has 8 classes in morning session and 7 classes in afternoon session.
4. Computerized Accounting department has 12 classes in morning session and 12 classes in afternoon session.
5. Total lecturers that are available for three departments is 84 people.

Table 3 shows the speed performance and fitness value for the best schedule obtained using different parameter configuration. Every generated schedule that is stated in Table 3 have satisfied all the hard constraint.

Table 3. Speed performance and fitness value for the best schedule obtained using different parameter configuration

Parameter Configuration	HMCR = 0.8	HMCR = 0.5	HMCR = 0.8	HMCR = 0.5
	PAR = 0.4	PAR = 0.4	PAR = 0.7	PAR = 0.7
Time (s)	61.656	156.125	49.453	149.797
Fitness value (0~1)	0.98904	0.98780	0.98699	0.98699

From the result, the best parameter for HMCR is in the range of 0.7 to 0.95. The PAR value doesn't seem to significantly affect the result.

## 5. CONCLUSION

In this research, our implementation of Harmony Search has fulfilled every hard constraint of UTP problem in STMIK Mikroskil. This means that the generated schedule is feasible for usage in real academic situation. It is also found that HMCR value with range of 0.7 to 0.95 is best suited for scheduling purpose. For future work, it is hoped that the scheduler application scope can be widen into classroom allocation problem.



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