

The Decision Support System (DSS) Application to Determination of Diabetes Mellitus Patient Menu Using a Genetic Algorithm Method

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Abstract. Composition of foods containing sugar in people with Diabetes Mellitus should be balanced, so an app is required for facilitate the public and nutritionists in determining the appropriate food menu with calorie requirement of diabetes patient. This research will be recommended to determination of food variation for using Genetic Algorithm. The data used is nutrient content of food obtained from *Tabel Komposisi Pangan Indonesia* (TKPI). The requirement of caloric value the patient can be used the PERKENI 2015 method. Then the data is processed to determine the best food menu consisting of energy (E), carbohydrate (K), fat (L) and protein (P) requirements. The system is compared with variation of Genetic Algorithm parameters is the total of chromosomes, Probability of Crossover (Pc) and Probability of Mutation (Pm). Maximum value of the probability generation of crossover and probability of mutation will be the more variations of food that will come out. For example, patient with gender is women aged 61 years old, height 160 cm, weight 55 kg, will be resulted number of calories: (E=1621.4, K=243.21, P=60.80, L=45.04), with the gene=4, chromosomes=3, generation=3, Pc=0.2, and Pm=0.2. The result obtained is the three variants: E=1607.25, K=198.877, P=95.385, L=47.508), (E=1633.25, K=196.677, P=85.885, L=55.758), (E=1630.90, K=177.455, P=85.245, L=64.335).

1. Introduction

In 2005 there were 200 million people with diabetes and is expected to increase to 333 million people in 2025. Countries like India, China, the United States, Japan, Indonesia, Pakistan, Bangladesh, Italy, Russia and Brazil is the top 10 countries with the largest population of diabetes [1]. The development of Diabetes Mellitus disease causes concern throughout the world, especially people who have nutritional transition marked by reduced traditional food intake and increased intake of instant food or fast food.

In the American Diabetes Association (ADA) research concluded that there are no more effective nutrients than carbohydrate, protein and fat. The proportion of macro nutrition should be tailored to each individual need to meet metabolic goals in a person with Diabetes Mellitus [2]. Diabetes Mellitus occurs sustainably, nutritional therapy in Diabetes Mellitus is effective for improving blood sugar control and other metabolic functions. In general, the energy needs of individuals with Diabetes Mellitus is not different from normal people.

Paradigm that happened in society is patient with Diabetes Mellitus should not consume sugar, but sugar is very needed for body metabolism process. Metabolism

that occurs in the body will produce energy that will be needed for the activity. Lack of intake of sugar in a person will result in the occurrence of hypoglycemia, a condition in which blood sugar levels dropped dramatically. It would be dangerous and even fatal. Diabetes Mellitus sufferers should continue to consume sugar in sufficient quantities, no more and no less.

Therefore, the determination of energy needs in people with Diabetes Mellitus is considered very important. How to determine it is by using the formula of the PERKENI 2015. The problem that exists today is that people still rarely use the formula and more often use the formula approximately. So the provision of diet to patients with Diabetes Mellitus is not accurate. Hospitals have set the PERKENI 2015 formula but when determining the portion of the menu still uses the approximate formula because it is too complex, and it takes time for the community or nutritionist to calculate the energy needs of each patient each day.

The Genetic Algorithm has been widely used in various optimizations, including numerical optimization and combinatorial optimization problems. The advantages of Algorithm methods from other methods are capable of handling various types of objective functions with constraints on discrete, continuous, or mixed search spaces. The advantages possessed by

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Genetic Algorithm are selection evolution operations that work probabilistically against several possible solutions to make genetic algorithm very effective in global optimization. The genetic algorithm has a fairly good flexibility and is sufficiently possible to combine it with other methods and make a more efficient implementation of a particular problem [3].

From the problems and conditions that have been discussed earlier, it is necessary an information system to make the determination of food menu of people with diabetes mellitus using Genetic Algorithm. Genetic Algorithm is used to randomize food based on the calories needed.

2. Theoretical Framework

2.1. Diabetes Mellitus

Diabetes Mellitus (DM) is a collection of symptoms that arise in someone who has increased blood glucose levels due to deficiency of insulin hormone in absolute or relative. The implementation of the diet should be accompanied by physical exercise and changes in food behavior. According to the consensus of Management of Diabetes Mellitus in Indonesia by the association of Endocrinology of Indonesia, Diabetes Mellitus disease is divided into 4 groups, namely: Type I and Diabetes Mellitus, Gestational Diabetes Mellitus, and other types [4].

Determination of food for people with Diabetes disease aims to maintain blood glucose levels to near normal, maintain normal serum lipid levels, and provide enough energy to maintain or achieve normal weight. The dietary requirements of inner Diabetes disease [4] are as follows:

- a. Foods with sufficient energy correspond to the total energy required by the patient to achieve and maintain a normal weight.
- b. Food with enough protein, ie 10-15% of total energy needs in a day. Fat is 20-25% of total energy needs in a day. And carbohydrates are the rest of the total energy requirement, which is 60-70%.
- c. The use of pure sugar in beverages and foods is not allowed unless the amount is a bit as a spice. When the glucose level is controlled, it is allowed to consume pure sugar to 5% of total energy needs.
- d. Vitamins and minerals.
- e. Distribution of food in a day:
 1. Breakfast: 20% x Total Energy
 2. Food interlude morning: 10-15% x Total Energy
 3. Lunch: 30% x Total Energy
 4. Interlude afternoon snack: 10-15% x Total Energy
 5. Dinner: 25% x Total Energy

2.1.1 Food List and Nutrient Composition

Data of nutritional composition of food used in this research taken from the Indonesian Food Composition Table (TKPI) and report on the results of research on the composition of food nutrients carried out in the Center

for Research and Development of Nutrition and Food Ministry of Health.

2.1.2 Consensus Formulas PERKENI 2015

Consensus formula 2015 is used to calculate the nutritional needs of patients with Diabetes Mellitus. The formula is as follows [5]:

1. Energy

The main components that determine energy demand are Basal Metabolic Rate (AMB) or Basal Metabolic Rate (BMR) influenced by age, gender, weight, and height. To determine BMR as follows:

$$\begin{aligned} BMR \text{ for men} &= 30 \times \text{Ideal weight} \\ BMR \text{ for women} &= 25 \times \text{Ideal weight} \end{aligned} \quad (1)$$

With the calculation of Ideal Weight using Broca's modified formula:

- For men again with Height ≥ 160 cm and women with Height ≥ 150 cm. The formula used is:

$$\text{Ideal weight} = 90\% \times (\text{Height} - 100) \quad (2)$$

- For men again with Height < 160 cm and women with Height < 150 cm. The formula used is:

$$\text{Ideal weight} = \text{Height} - 100$$

$$\text{Energy} = (\text{BMR} + \text{Activity Factor}) - \text{Age Factor} + \text{Stress Factor} \pm \text{Body Correction Factor} \quad (3)$$

Table 1. Activity Factor

Activity Level	Formula
Bedrest	10% from BMR
Mild	20% from BMR
Medium	30% from BMR
Heavy	40% from BMR

Table 2. Stress Factor

Stress Factor	Value
No stress, normal nutritional status	1.2-1.3
Mild Stress: inflammation of the gastrointestinal tract, cancer, effective surgery, trauma, post minor surgery	1.3-1.4
Medium stress: sepsis, bone surgery, burns, liver disease, major surgery post	1.4-1.5
High Stress: HIV AIDS + complications, multisystem surgery, Pulmonary TB + complications	1.5-1.6
Very Heavy Stress: Severe head trauma	1.7

Table 3. Age Factor

Age	Formula
0-40 years old	0% from BMR
40-59 years old	5% from BMR
60-69 years old	10% from BMR
≥ 70 years old	15% from BMR

Table 4. Body Correction

Nutrition Status	Formula
Normal	0% from BMR
Malnutrition	+ 20% from BMR
Obesity	- 20% from BMR

With the formula as:

$$Nutritional\ Status = \frac{Actual\ Weight}{Ideal\ Weight} \times 100\% \quad (4)$$

Information:

Status result > 110 = Obesity

Status result 90-100 = Normal

Status result < 90 = Malnutrition

1. Protein

Protein requirement normally is 15% of total energy requirement. The protein equation is as follows:

$$Protein = \frac{15\% \times Total\ Energy\ (TE)}{4} \quad (5)$$

2. Fat

While for normal fat requirement is 25% of total energy requirement. The fat equation on Harris Benedict method is as follows:

$$Fat = \frac{25\% \times Total\ Energy\ (TE)}{9} \quad (6)$$

3. Carbohydrate

While for normal carbohydrate needs is 60% of total energy needs, or residual energy after being reduced by energy derived from protein and fat. The equation determines carbohydrates are as follows:

$$Carbohydrate = \frac{60\% \times Total\ Energy\ (TE)}{4} \quad (7)$$

Diet used as part of Diabetes Mellitus management is controlled based on energy, protein, fat and carbohydrate content. As a guideline used 8 types of Diabetes Mellitus Diet as can be seen in Table 5 [4]:

Table 5. Type Diet Diabetes Melitus

Type Diet	Energy (Kcal)
I	1100
II	1300
III	1500
IV	1700
V	1900
VI	2100
VII	2300
VIII	2500

2.2. Genetic Algorithm

The genetic algorithm is a heuristic search algorithm based on the mechanism of biological evolution. The diversity in biological evolution is a variation of the chromosomes between individual organisms. The Genetic algorithm was developed by John Holland from Michigan University. John Holland says if any problem (natural or artificial) associate with adaptation can be formulated in genetic terms. The genetic algorithm is a simulation of the evolutionary process of Darwin and the genetic operation of the chromosomes.

In this algorithm, the search technique is done simultaneously on a number of solutions known as the population. Individual in one population are called chromosomes. This chromosomes is a solution that is still a symbol. The initial populations randomly constructed, while the next population is the result of chromosomes evolution through an iteration called by the term generation. In each generation, the chromosomes will go through the evaluating process using a measuring instrument called a fitness function.

The fitness value of a chromosomes will show the quality of the chromosomes in that population. The next generation known as term child (off-spring) formed from the combined 2 generation chromosomes now acting as the parent using a cross carrier (crossover), in addition to the cross carrier, a chromosomes may also be modified by a mutation operator. The population of new generation is formed by selecting the fitness value of the parent chromosomes and the fitness value of a child's chromosomes (offspring), as well as rejecting the other chromosomes so the population size (the number of chromosomes in a population) still constant. After going through several generations, this algorithm will converge to the best chromosomes. In general, a simple genetic algorithm flow diagram as shown in the figure 1:

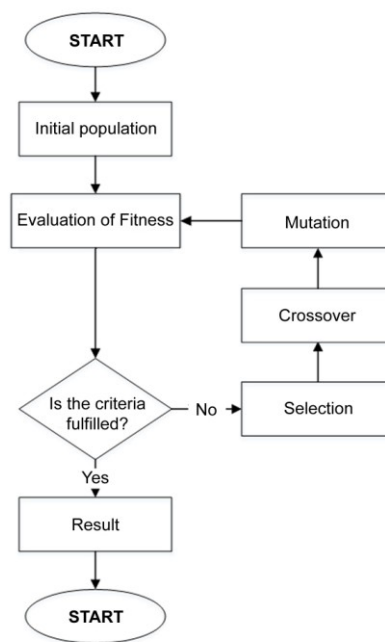


Fig 1. Genetic Algorithm Cycle

To implement genetic algorithm, the main components to do are:

1. Generate the initial population

Before generating the initial population, it must first determine the number of individuals in the population. For example the number of individuals is N. After that, it generates only initial populations that have individual beginnings at random.

2. Evaluation function

There are 2 things to do in evaluating chromosomes, that is: evaluate objective function (objective function) and conversion of objective function to fitness function. In general, the fitness function is derived from an objective function with a non-negative value.

3. Selection

Selection will determine which individuals will be selected for recombination and how offspring is formed from the selected individual. The first step in this selection is the fitness value search. Each individual in a selection vessel will accept the probability of reproduction that depends on its own objective value on the objective value on an individuals in the container. This fitness value will be used in the next selection stage.

A commonly used selection method is roulette-wheel. This selection aims to provide greater reproductive opportunities for members of high-fitness populations to reproduce. The workings of the roulette-wheel method are as follows:

a. Calculate the total fitness (F):

$$TotFitness = \sum F_k ; k = 1,2, \dots, popsize \quad (8)$$

b. Calculate the relative fitness of each individual:

$$P_k = F_k / TotFitness \quad (9)$$

c. Calculate cumulative fitness

$$Q_k = Q_{k-1} + P_k ; k = 2,3, \dots, popsize \quad (10)$$

d. Select the parent that will be the candidate for the crossover:

1. Generate random r
2. If $Q_k \leq r$ and $Q_{k+1} > r$, then select the chromosome into (k+1) as the parent candidate.

4. Crossover

Crossover (crossing) is done on two chromosomes to produce a child's chromosome (offspring). The child's chromosomes that are formed will inherit some of the parental chromosomal properties. In crossover there is one parameter that is very important that is probability of crossover (Pc). Probability of crossover show the ratio of the children generated in each generation to the size of the population. Example the population size (popsize = 100), while the probability of crossover (Pc = 0.25), means there is expected 25 chromosomes of 100 chromosomes in the population will be crossover.

5. Mutation

Mutation is the process of modifying a child's chromosomes randomly. Mutations will create new individuals by altering one or more genes present in a chromosome. Mutations play a role in replacing the missing genes of the population due to the selection process and allow for the emergence of genes not present in the initial population. In the mutation there is one parameter that is very important is the probability of mutation (Pm). Probability of mutation show the percentage of total genes in the population that will undergo a mutation. To make a mutation, we must first calculate the total number of genes in the population. Then generate a random number that will determine which position will be mutated. Example the population size (popsize = 100), each chromosome has a length of 20 genes, then the total gene is 100 x 20 = 2000 genes. If the mutation probability (Pm = 0.01), means that there is expected (1/100) x 2000 = 20 genes will mutate.

3. Methodology

This research uses Genetic Algorithm method to optimize calorie and calorie needs of food so get the appropriate menu variation. Food data used in this study is Indonesian Food Composition Table (TKPI) and report on the results of research on the composition of food nutrients carried out at the Center for Research and Development of Nutrition and Food Ministry of Health.

In an effort to obtain optimal results, this research is done through the stages to be done, namely:

1. Literature review.

At this stage the process of collecting literature such as international and national journals in accordance with the topic of research, as well as articles relevant to the topic of research.

2. Interview.

Observation is done by conducting interview process to nutritionist of General Hospital of Kudus.

3. Analysis and Design of Information System

The design of this system is the conceptualization stage, which is a stage that requires analysis in the design of the system (software) so that the system is made according to the needs of the user. The diagram used in the design of this application is the Unified Language.

4. System implementation

At this stage will implement the system into the application to be built using PHP and MySQL database based web.

Flow Chart on the Implementation of Genetic Algorithm for Decision Support System Food Menu Determination of Diabetes Mellitus Patients is shown in figure 2.

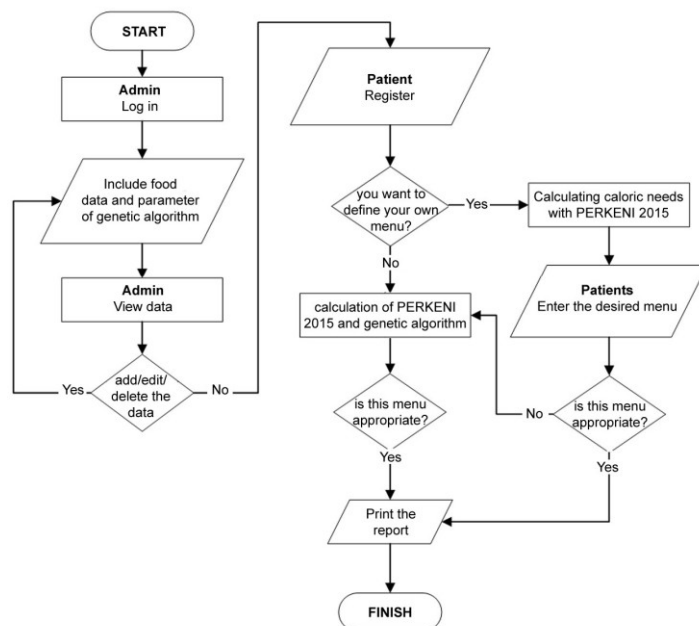


Fig 2. Flow Chart

$$\text{Nutritional Status} = \frac{\text{Actual Weight}}{\text{Ideal Weight}} \times 100\%$$

$$= \frac{55}{54} \times 100\%$$

$$= 101\% = \text{Normal}$$

$$\begin{aligned} \text{Energy} &= (\text{BMR} + \text{Activity Factor}) - \text{Age} \\ &\quad \text{Factor} + \text{Stress Factor} \pm \text{Body} \\ &\quad \text{Correction Factor} \\ &= (1350 + (30\% \times 1350)) - (10\% \times 1350) + 1.4 - 0 \\ &= (1350 + 405) - 135 + 1.4 - 0 \\ &= 1621.4 \end{aligned}$$

$$2. \text{Carbohydrate} = \frac{60\% \times \text{Total Energy}}{4} = 243.21$$

$$3. \text{Protein} = \frac{15\% \times \text{Total Energy}}{4} = 60.80$$

$$4. \text{Fat} = \frac{25\% \times \text{Total Energy}}{9} = 45.04$$

b. Process of Initialization Phase and Evaluation Function

The size of the population depends on the problem to be solved. After the population size is determined, then initialization of the chromosomes in the population is performed. Chromosome initiation is done randomly. Initial population by taking samples of 4 kinds of foods as genes and 3 menus as chromosomes to be done manual calculations.

There are two things to do in evaluating chromosomes, namely evaluating the value of objective

4. Result and Discussion

4.1. Research result

The results of the research were conducted by comparing the parameters used in the Genetic Algorithm, so it can form a conclusion of the use of suitable parameters for a relatively large amount of data. Food menu determination system can provide information on which food menu is suitable to serve as a menu for people with diabetes mellitus with the calories needed.

4.2. Discussion

Before calculating the food menu using Genetic Algorithm, first calculate caloric requirement using PERKENI 2015.

Examples of patient test data used:

Name : Siti Fatimah
 Age : 61 years old
 Sex : Women
 Height : 160 cm
 Weight : 55 kg
 Activity : Medium
 Stress factor : Medium

a. Calculate the Nutritional Needs

1. Energy

$$\begin{aligned} \text{BMR} &= 25 \times \text{Ideal weight} \\ \text{Ideal weight} &= 90\% \times (\text{TB}-100) \\ &= 90\% \times (160-100) \\ &= 54 \\ \text{BMR} &= 25 \times 54 \\ &= 1350 \end{aligned}$$

function and conversion of objective function into fitness function. The number of calories here as an objective function.

The fitness function is lowered by a non-negative value. At the fitness value here, the minimum caloric difference is the optimization to be achieved in the determination of the food menu. So the smallest caloric difference will get the value of fitness with the greatest value. The initial population show in table 6, objective function show in table 7 and fitness value in table 8.

Table 6. Initial Population

Chromosome	Random Menu
K[1]	Rice, Sea Cucumber Dendeng, Fried Tempeh, Spinach Steamed
K[2]	Brown Rice, Rendang, Fried Tofu, Papaya Salad
K[3]	Savory Rice, Peda Banjar, Fried Tofu, Boiled Bean

Table 7. Objective Function

Chromosome	Total Calories of Each Food				Total of Calories
	FO K[1]	FO K[2]	FO K[3]	FO K[4]	
FO K[1]	135	231	175	30	571
FO K[2]	111.75	96,5	57.5	49	314.75
FO K[3]	142.5	156	57.5	30	386

Table 8. Fitness Value

Chromosome	Total of Calories	Deviation of Calories	Fitness Value (F)
Fitness K[1]	571	$ 324,28 - 571 = 246,72$	$\frac{1}{246,72} = 0,0041$
Fitness K[2]	314.75	$ 324,28 - 314,75 = 9,53$	$\frac{1}{9,53} = 0,1038$
Fitness K[3]	386	$ 324,28 - 386 = 61,72$	$\frac{1}{61,72} = 0,0162$
Total Fitness			0.1241

c. Stage of Selection

After obtaining the value of the cultivated fitness (Q) and the random value (R), an approximation of R values to the Q value is altered to obtain a new chromosome. The following stage of selection in a table 9 and result of selection in table 10.

d. Stage of Crossover

Crossovers are performed on two chromosomes to produce a child's chromosome (offspring). The child's chromosomes that are formed will inherit some of the parental chromosomal properties. Probability of crossover (Pc) is 0.5 then the value below 0.5 will be selected to be the parent chromosome. The following stage of crossover in a table 11 and result of crossover in table 12.

Table 9. Stage of Selection

Chromosome	Fitness Relative (P)	Fitness Cumulative (Q)	Random Value (R)
K[1]	$\frac{0,0041}{0,1241} = 0,0330$	0.0330	0.3553
K[2]	$\frac{0,1038}{0,1241} = 0,8364$	$0,0330 + 0,8364 = 0,8694$	0.9198
K[3]	$\frac{0,0162}{0,1241} = 0,1305$	$0,8694 + 0,1305 = 0,9999$	0.5786

Table 10. Result of Selection

Random Value (R)	Fitness Cumulative (Q)	Comparative Result	New Chromosome
0.3553	0.0330	K[2] from the previous population	Red Rice, Rendang, Tahu goreng, Papaya Salad
0.9198	0.8694	K[3] from the previous population	Savory Rice, Peda banjar, Fried Tofu, Boiled Bean
0.5786	0.9999	K[2] from the previous population	Red Rice, Rendang, Fried Tofu, Papaya Salad

Random Value (R)
0.1197
0.4297
0.6950

Table 11. Stage of Crossover

Selected Chromosome	Population
K[1]	Brown Rice, Rendang, Fried Tofu, Papaya Salad
K[2]	Savory Rice, Peda banjar, Fried Tofu, Boiled Bean

Table 12. Result of Crossover

Chromosome	New Population from Crossover Outcome	Total Calories
K[1]	Brown Rice, Rendang, Fried Tofu, Boiled Bean	295.75
K[2]	Savory Rice, Peda Banjar, Fried Tofu, Papaya Salad	405
K[3]	Brown Rice, Rendang, Fried Tofu, Papaya Salad	314.75

e. Stage of Mutation

The number of chromosomes that have mutations in one population is determined by the probability of mutation (Pm). In this mutation it is possible to emerge a new chromosome that had not originally appeared in the initial population. Each chromosome has a length of 4 genes, then the total gene is $3 \times 4 = 12$ genes. If the probability of mutation (Pm = 0.5), the expected means is $0,5 \times 12 = 6$. It is expected that there are 6 genes that will have mutations. The following calculation of the mutation stage is shown in table 13 and result of mutation in table 14

Table 13. Stage of Mutation

Random from 1 to 20	Random Mutation
Random number = 2 1 st Chromosome, 2 nd index	Rendang → Fried Chicken
Random number = 4 1 st Chromosome, 4 th index	Boiled Bean → Cap cai
Random number = 5 Kromosom ke-2 index ke-1	Savory Rice → Rice
Random number = 6 2 nd chromosome, 2 nd	Peda Banjar → Fried Fish

Random from 1 to 20 index	Random Mutation
Random number = 7 2 nd chromosome, 3 rd index	Fried Tofu → Fried Tempeh
Random number = 8 2 nd chromosome, 4 th index	Papaya salad → Boiled Cassava Leaves

Table 14. Result of Mutation

Chromosome	Result of Mutation	Calories	Fitness
K[1]	Brown Rice, Fried Chicken, Fried Tofu, Cap cai	365.25	0.02441
K[2]	Rice, Fried Fish, Fried Tempeh, Boiled Cassava Leaves	551.5	0.00440
K[3]	Brown Rice, Rendang, Fried Tofu, Papaya Salad	314.75	0.10493

So, one of selected food menu is K[3]. Calculated to the desired generation.

4.3. Testing System

Below is a picture of the system implementation:

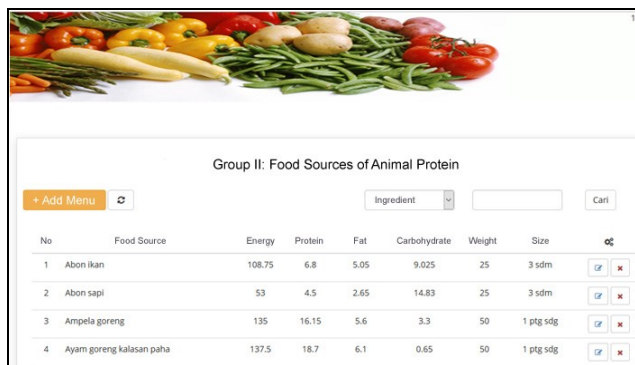


Fig 3. Pages of Food Data

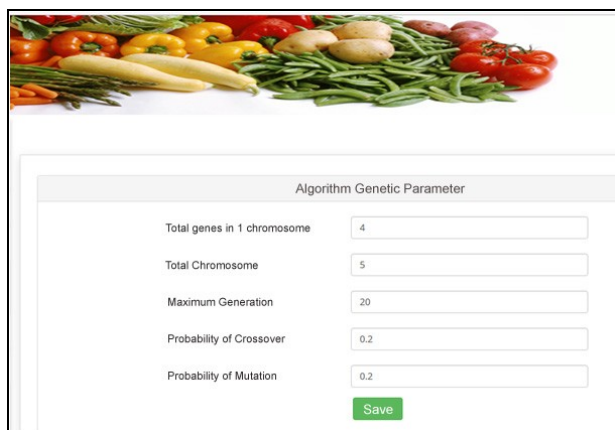


Fig 4. Pages of Parameter Genetic Algorithm

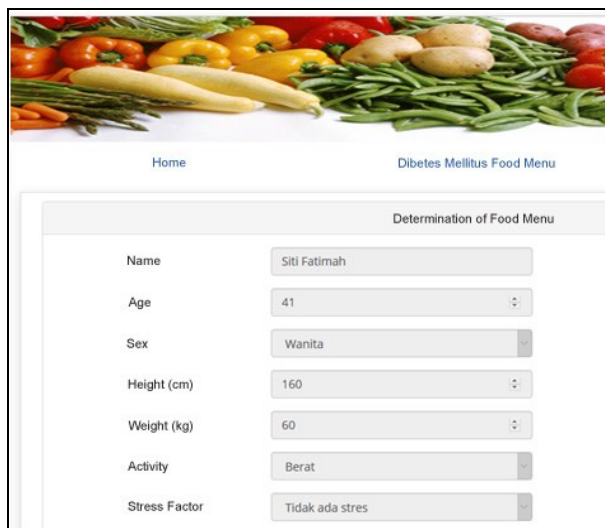


Fig 5. Pages of Patient Data

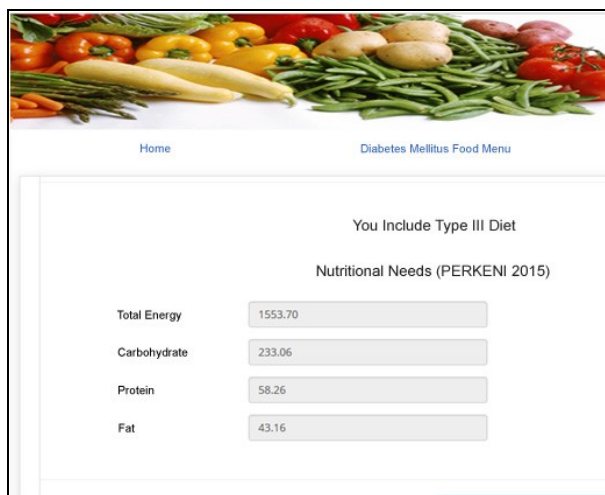


Fig 6. Pages of Patient Calorie Needed

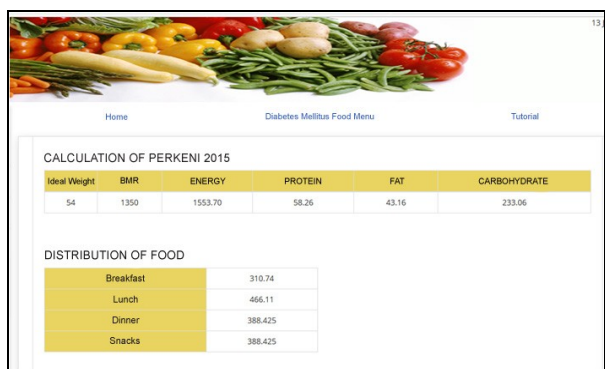


Fig 7. Pages of Partition Food Calorie

No	Population Initialization	Total Calories	Objective Function Value	Fitness Value (F)	Fitness Relative (P)	Cumulative Fitness (Q)	Random Value (R)
0	Nasi tim, Tepung teri, Tempe gembus, Botok lamtoro	485.3	174.56	0.0057	0.1776	0.1776	0.9162
1	Nasi gurih, Ayam goreng sukabumi paha, Keripik tempe, Pelecing kangkung	522.7	211.96	0.0047	0.1464	0.324	0.5696
2	Nasi gurih, Kepting, Tempe goreng, Sayur bunga papaya	470	159.26	0.0063	0.1963	0.5203	0.6397
3	Nasi beras merah, Bandeng, Tempe goreng, Toge goreng	225.5	85.24	0.0117	0.3645	0.8848	0.0757
4	Nasi gurih, Ikan asin kering, Tempe gembus, Buncis rebus	584	273.26	0.0037	0.1153	1.0001	0.2709

Fig 8. Pages of Initial Population and Fitness Value

Random 1 to 20	Random Mutation
Bilangan acak: 2 Kromosom ke-0 index ke-1	Bandeng->Ayam goreng mbok berek dada
Bilangan acak: 6 Kromosom ke-1 index ke-1	Ikan asin kering->Ayam goreng kentucky dada
Bilangan acak: 12 Kromosom ke-2 index ke-3	Toge goreng->Buncis rebus
Bilangan acak: 20 Kromosom ke-4 index ke-3	Pelecing kangkung->Tumis bayam bersantan

Fig 12. Page of Mutation Stage

No	The Value of R Approaching Q	Selected New Population
0	Kromosom ke-4 dari populasi sebelumnya	Nasi gurih, Ikan asin kering, Tempe gembus, Buncis rebus
1	Kromosom ke-3 dari populasi sebelumnya	Nasi beras merah, Bandeng, Tempe goreng, Toge goreng
2	Kromosom ke-3 dari populasi sebelumnya	Nasi beras merah, Bandeng, Tempe goreng, Toge goreng
3	Kromosom ke-0 dari populasi sebelumnya	Nasi tim, Tepung teri, Tempe gembus, Botok lamtoro
4	Kromosom ke-1 dari populasi sebelumnya	Nasi gurih, Ayam goreng sukabumi paha, Keripik tempe, Pelecing kangkung

Fig 9. Pages of Selection Result

No	New Population of Mutation Result	Calorie	Objective Function	Fitness
0	Nasi gurih, Ayam goreng mbok berek dada, Tempe goreng, Toge goreng	522	211.26	0.0047
1	Nasi beras merah, Ayam goreng kentucky dada, Tempe gembus, Buncis rebus	318	7.26	0.1359
2	Nasi beras merah, Bandeng, Tempe goreng, Buncis rebus	211.5	99.24	0.0101
3	Nasi tim, Tepung teri, Tempe gembus, Botok lamtoro	485.3	174.56	0.0057
4	Nasi gurih, Ayam goreng sukabumi paha, Keripik tempe, Tumis bayam bersantan	495.7	184.96	0.0054

Fig 13. Page of Mutation Result

No	Random Value (R)
1	0.0803
2	0.0081
3	0.9362
4	0.6894
5	0.0736

Fig 10. Pages of Random Value Crossover Determination

Menu 1	Menu 2	Menu 3	Menu 4	Menu 5
Breakfast • Nasi beras merah 50 gram • Tongkol 100 gram • Tempe gembus 100 gram • Semur jengkol 30 gram	Breakfast • Nasi beras merah 50 gram • Tongkol 100 gram • Tempe gembus 100 gram • Semur jengkol 30 gram	Breakfast • Nasi beras merah 50 gram • Tongkol 100 gram • Tempe gembus 100 gram • Semur jengkol 30 gram	Breakfast • Nasi beras merah 50 gram • Tongkol 100 gram • Tempe gembus 100 gram • Semur jengkol 30 gram	Breakfast • Nasi beras merah 50 gram • Tongkol 100 gram • Tempe gembus 100 gram • Semur jengkol 30 gram
Lunch • Nasi gurih 75 gram	Lunch • Nasi gurih 75 gram	Lunch • Nasi gurih 75 gram	Lunch • Nasi gurih 75 gram	Lunch • Nasi gurih 75 gram

Fig 14. Page of Food Menu Determination Result

No	New Population of Crossover Result	Calorie
0	Nasi gurih, Bandeng, Tempe goreng, Toge goreng	436
1	Nasi beras merah, Ikan asin kering, Tempe gembus, Buncis rebus	373.5
2	Nasi beras merah, Bandeng, Tempe goreng, Toge goreng	225.5
3	Nasi tim, Tepung teri, Tempe gembus, Botok lamtoro	485.3
4	Nasi gurih, Ayam goreng sukabumi paha, Keripik tempe, Pelecing kangkung	522.7

Fig 11. Pages of Crossover Result

Table 15. Testing System

Number	Total of Genes in 1 Chromosomes	Total of Chromosomes	Maximum of Generation	Probability of Crossover (Pc)	Probability of Mutation (Pm)	Attempts to -										Time
						1	2	3	4	5	6	7	8	9	10	
1	4	5	5	0.1	0.1	3	0	1	0	5	1	2	4	0	5	1 second
2	4	15	5	0.3	0.1	1	3	3	5	2	5	3	2	0	3	2 second
3	4	25	5	0.5	0.1	2	5	1	5	1	5	4	4	4	5	3 second

Number	Total of Genes in 1 Chromosomes	Total of Chromosomes	Maximum of Generation	Probability of Crossover (Pc)	Probability of Mutation (Pm)	Attempts to -										Time
						1	2	3	4	5	6	7	8	9	10	
4	4	5	10	0.1	0.3	10	10	7	0	7	6	7	9	5	5	2 second
5	4	15	10	0.3	0.3	9	6	7	8	9	7	7	8	7	7	4 second
6	4	25	10	0.5	0.3	6	9	7	6	9	7	5	6	5	4	7 second
7	4	5	15	0.1	0.5	9	7	14	2	10	11	11	5	9	3	3 second
8	4	15	15	0.3	0.5	13	11	10	12	13	15	14	12	10	12	7 second
9	4	25	15	0.5	0.5	13	12	14	15	14	10	14	13	13	11	11 second
10	4	5	20	0.1	0.1	9	9	8	11	7	8	2	10	7	11	3 second
11	4	15	20	0.3	0.1	8	10	9	13	8	8	10	2	4	14	8 second
12	4	25	20	0.5	0.1	13	16	5	9	9	6	14	7	7	6	12 second
13	4	5	25	0.1	0.3	21	18	16	15	18	15	17	9	11	1	4 second
14	4	15	25	0.3	0.3	10	17	16	16	9	14	10	14	10	8	10 second
15	4	25	25	0.5	0.3	11	6	8	9	13	18	5	11	19	11	15 second
16	4	5	30	0.1	0.5	23	18	19	20	22	15	16	23	20	22	5 second
17	4	15	30	0.3	0.5	25	28	29	27	29	25	29	28	23	24	12 second
18	4	25	30	0.5	0.5	26	28	29	30	28	26	25	27	27	26	21 second

5. Conclusion

4.4. Analysis result

The results of the analysis that can be taken from the results of system testing performed:

1. In the determination of diabetes mellitus food made comparison using genetic algorithm parameters namely Chromosome Number, Maximum Generation, Crossover Probability, and Probability of Mutation.
2. By entering the number of chromosomes a little for example entering the number of chromosomes 5 will use a fast time during the process. Whereas if entering chromosome more than 15 will use a lot of time.
3. Output results will be more optimal if the number of chromosomes are used slightly and the greater the value of Maximum Generation, Probability Crossover and Probability Mutations will be more varied food that will come out.
4. Testing the system performed to produce benefits in terms of users is that if the user is a nutritionist should be use total chromosome = 5, Maxium of Generation = 30, Probability of Crossover = 0,1 and Probability of Mutation = 0,5. Because, with the total of parameter will produce a stable variation and use shorted time, that is 4 second. If the user is a layman and want to choose a lot of dietary variations it is better to use the total chromosomes = 15, Maximum of Generation = 30, Probability of Crossover = 0,3, and Probability of Mutation = 0,5. Because, with the number of parameter will produce a lot of dietary variations and use shorted time to compared with other, that is 12 second.

Genetic Algorithm Method can be applied to decision support system of food menu selection for people with diabetes mellitus. In the results of research conducted by comparing the parameters of genetic algorithm. By entering a small number of chromosomes will use a fast time during the process. And the greater the value of Maximum Generation, Crossover Probability and Probability Mutation will be more variation of food that will come out. Using Number of Chromosomes = 5, Maximum Generation = 30, Crossover Probability = 0.1 and Probability Mutation = 0.5 using a not too long time of 5 seconds with little food variation. While using the Number of Chromosomes = 15, Maximum of 50, Probability Crossover = 0.3 and Probability Mutation = 0.5 using a slightly longer time of 12 seconds with many variations of food and by using Number of Chromosomes = 25, Maximum Generation = 30, Probability Crossover = 0.5 and Probability of Mutation = 0.5 using a very long time of 21 seconds with many food variations and relatively stable.

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