# Food Recommendation System for Diabetic Patient

Mr. Sameer Maitre, Mr. Ajay Waghmare, Ms. Bhagyashree Metre, Ms. Sakshi Deshmukh, Ms. Shruti Gokhale,

Prof. L.K.Gautam

Sipna College of Engineering & Technology, Amravati.

Abstract—Blood glucose levels that are too high can cause diabetes. Our primary energy source is blood glucose, which is obtained from the food you eat. The pancreas produces the hormone insulin, which facilitates the uptake of glucose from food into our cells for use as fuel. Type-1 and Type-2 diabetes are conditions in which the body produces little to no insulin or uses it poorly. After that, glucose remains in our blood and does not enter our cells. Over time, having too much glucose in our blood can result in health issues. Patients frequently experience heart attacks or strokes, eye issues that can cause vision loss or blindness, discomfort, tingling, or numbness in their hands and feet (also known as nerve damage), renal issues that could result in our kidneys ceasing to function, teeth and gum issues, and kidney problems. A person with diabetes needs to maintain a healthy balance in their everyday activities to keep their diabetes under control and retain good health[1].

Our proposed system, uses their health records in the form of general parameters and recommends them a perfect combination of diet to maintain good health.

### Introduction

A study by WHO reports that inadequate and imbalanced intake of food causes around 9% of heart attack deaths, about 11% of ischemic heart disease deaths, and 14% of gastrointestinal cancer deaths worldwide. Moreover, around 0.25 billion children are suffering from Vitamin-A deficiency, 0.2 billion people are suffering from iron deficiency (anemia), and 0.7 billion people are suffering from iodine deficiency. The main objective of this work to recommend a diet to different individual[5].

Our proposed system, uses their health records and recommends them a perfect combination of diet to maintain good health.

This work is made easily accessible and available to the users in the form of a website, so that they can get their plans wherever and whenever they want to. Nowadays, a human being is suffering from various health problems such as fitness problem, inappropriate diet, mental problems etc. Various studies depict that

inappropriate and inadequate intake of diet is the major reasons of various health issues and diseases.

### Literature review

The fact that Software Based Smartphone Applications for

Promoting Healthy Diet and Nutrition by Steven S. Coughlin, Mary S. Whitehead, et al. [1] were put into place highlights how quickly technology has advanced, encouraging the use of smartphones in health promotion research and practice. The most popular approaches include giving criticism, setting healthy eating goals, cooking healthy meals, making grocery list decisions while keeping track of one's weight, and many others. The project also emphasizes how cellphones have the ability to enhance the precision and thoroughness of self- monitored nutritional intake in weight-control interventions and how they typically lighten the workload associated with such monitoring using conventional paper-based records. Accuracy of diet and nutrition measurements obtained using mobile devices has generally been found to be good. Participants prefer quick and easy to administer applications and those that increase awareness of food intake and weight management.

Barry M. Popkin, distinguished professor at W. R. Kenan, Jr.

University, wrote a study titled "Food Consumption and its Impact on Cardiovascular Disease: Importance of Solutions" that focused on the globalized food system. In [2], we learn about:

1. The evolution of the contemporary, globalized food system and its effects on the food supply.

2. A general agreement on the evidence linking different diets and macronutrients to CVD and its associated comorbidities.

3. A description of how modifications to the world's food system can both solve the nation's present dietary-related public health issues and counteract climate change. This paper provided a state-of-the-art review of the link

between specific macronutrients and foods and cardiovascular diseases (CVD) and summarized how the global food system contributes to dietary patterns that greatly increase the risks for



the population to experience ill health. In the article by Yue You, Svetlana Vladislavovna Doubova,

Diana Margarita Pinto Masis, Ricardo Pérez Cuevas, Victor Hugo Borja-Aburto, and Alan Hubbard, machine learning methodology is applied to evaluate the effectiveness of the diabetic's programme for patients with type 2 diabetes in family medicine clinics in Mexico. [3] Clinic and laboratory databases are where the data is gathered.

The machine learning algorithms utilized were precisely a regression tree, and the main examination of diabetes' impact on glucose management was the outcome, which was the predicted treatment impact. The glycemic control of people with type 2 diabetes is generally discussed in this paper. Counting the number of T2D individuals utilizing databases and algorithms.

Developing a decision support system to assess diabetic patients' carbohydrate consumption by Levente Kovacs, Lorand Vajda, and Patricia Pinter [4] makes use of telemedicine. The technology was created to provide home monitoring, which is an illustration of telemedicine. A form with text fields and drop-down menus can be used to enter personal and medical information to build new patient profiles.

Vital sign changes can be tracked using automatically generated graphs. The ability to prescribe medications is supported by notifications. There are instruments to help remind the patient to take their medication or to perform other tasks, such checking their blood pressure or exercising (each patient has daily programmes set up by their expert that trigger notifications every day). The clustering technique called Improved K-means which falls under the category of point assignment algorithms which is of unsupervised the initial partitions (centroids) have been calculated in a more significant way rather than random selection. This results in reducing the number of iterations.

# **Proposed Methodology**

The goal of this project is to develop a Diabetic Food Recommendation System that can provide personalized and relevant food recommendations for individuals with diabetes. The system will leverage machine learning techniques to analyze various factors such as the nutritional content, glycemic index, portion size, and personal preferences of users to generate tailored food recommendations.

The Diabetic Food Recommendation System will be designed to address the following challenges:

**Nutritional Content:** Identifying foods that are rich in essential nutrients, low in carbohydrates, and have a

balanced macronutrient profile is crucial for individuals with diabetes[3]. The recommendation system should be able to evaluate the nutritional content of different foods and provide recommendations based on the specific dietary requirements of the user.

**Portion Size:** Controlling portion sizes is important for managing blood sugar levels and overall calorie intake. The recommendation system should provide guidance on appropriate portion sizes for different foods, taking into consideration the user's individual needs and goals.

**Personal Preferences:** Taste preferences, cultural dietary habits, and individual food restrictions are important factors that influence food choices. The recommendation system should be able to account for these personal preferences and provide food recommendations that are aligned with the user's taste preferences and dietary restrictions.

The Diabetic Food Recommendation System aims to provide practical and actionable recommendations to help individuals with diabetes make informed food choices and manage their condition effectively[8]. By leveraging machine learning and techniques, the system can provide personalized, relevant, and adaptable food recommendations that align with the user's dietary requirements, taste preferences, and lifestyle, ultimately contributing to improved diabetes management outcomes.

# Design

### PROCESS

Food recommender systems are gaining popularity due to their importance in maintaining a healthy lifestyle. Most existing studies in the food domain focus on recommendations that suggest appropriate food items for individual users based on their preferences or health problems. In addition, recommendation features are extremely useful in the food industry. Such scenarios present numerous challenges for food recommender systems because the system must recommend food based on the height and weight of the users, which must be considered appropriately. These systems also include features for tracking nutritional consumption and persuading users to change their eating habits for the better. We present an overview of recommendation techniques for individuals and groups in the healthy food domain in this paper. Furthermore, we examine the current state-of-the-art in food recommender systems and discuss research challenges associated with the development of future food recommendation technologies.



#### **COLLECTION OF DATA**

The data has specific food related information which is related to the changes in the nutrients for the human being. The data used is observed from the lifestyle and daily records. The data collection process is responsible for the framework to do an inspection of the data collected and the data that can be compared to previously obtained data. This task further leads to the next task which is preprocessing which is applied to normalize the dataset using normalizing techniques. This task helps in dealing with data collecting and balancing to make a dataset that will be based on Machine learning.

We collect our data from Kaggle website in the form of rows and column. In our dataset there are multiple columns which contain the more information about the ingredient, prep time min, total time min, servings, course, is veg, is sugar free, is High protein, is gluten free, is sattvic, is vegan, diet, Allergic, Total Calories, website etc.

We have use Indian Food Dataset from Kaggle.com

A A	8	C	D	5		G H	1	1
1 foodID	name		prepTimeInWins co				course	isVeg
2	1 Masala Karela	6 Karela (Bitler Gourd' Pavakkai) - deseeder		30	45	6 Indian	Side Dish	FALSE
3	2 Spicy Tomato Rice	2-1 / 2 cups rice - cooked, 3 tornatoes, 3 tea	5	10	15	3 South Indian	Main Course	TRUE
4	3 Ragi Serriya Upma - Ragi Millet Verr	nice 1-1/2 cups Rice Vermicelli Noodles (Thin), 1	20	30	50	4 South Indian	South Indian Bra	n TRUE
5	4 Gongura Chicken Curry - Andhra St	vie (500 grams Chicken,2 Onion - chopped,1 To	15	30	45	4 Archra	Eurch	FALSE
6	5 Andhra Style Alam Pachadi - Adrak I	Chul 1 tablespoon chara dal, 1 tablespoon white u	10	20	30	4 Atchra	South Indian Bri	n TRUE
7	6 Pudina Khara Porgal (Rice and Lent	ils C 1 cup Rice - soaked for 20 minutes, 1/2 cup '	10	20	30	4 South Indian	South Indian Bra	n TRJE
8	7 Udupi Shyle Ash Gourd Coconut Curr	500 grams Vellai Poosanikai (Ash gourd Wh	i 10	30	40	4 Udupi	Lurch	TRIE
9	8 Mexican Style Black Bean Burnto	4 Tortillas, 1/4 cup Black bears - soaked over	r 10	30	40	4 Mexican	Lunch	TRUE
10	9 Spicy Crunchy Masala Idli	10 Idli - cut into strips,1 cup Green Bell Pepp	i 10	20	30	3 South Indian	Snack	TRJE
11	10 Cauliflower Leaves Chutney	1 cup cabbage leaves, 3/4 cup tomaloes, 18	5	20	25	3 South Indian	Side Dish	TRUE
12	11 Homemade Baked Beans (Wholeson	e & 250 grams Dry beans - (such as carnelini o	60	60	120	4 Fusion	High Protein Ver	TRJE
13	12 Veg Chili Cheese Burgers	2 Burger burs 5 Pickled Jalapenos - sliced 2	10	45	55	4 Continental	Nain Course	TRUE
14	13 Andhra Steel Ingavu Charu - Aslottd	a Fk 2 cup Tamarind Waler, 1/2 cup Arhar dal (Sp	0	30	30	4 Andrea	Lunch	TRUE
15	14 And fish soup - Bergali style fish in t	ona 600 grans Aar Maach (fish) - mhu' kalla fisl	5	15	20	2 Bengali	Dimer	FALSE
16	15 Saunf Aloo (Fennel Potato Curry)	5 Potatoes (Alco) - halved with skin,2 teaspo	10	15	25	6 Punjabi	Lurch	TRUE
17	16 South Indian Orion Chuthey - South	Indi 2 onions, 1 teaspoon curriin seeds, 2 tablesp	20	20	40	4 South Indian	Side Dish	TRUE
18	17 Hariyali Egg Curry In Coriander and	Min 4 Whole Eggs - boiled, 1 Cinnamon Stick (Da	i 15	20	35	4 Indian	Lunch	FALSE
19	18 Gourd Raita - Grated Bottle Gourd Yo	igur 2 cups curd, 1 cup gourd - peel and bytten,	15	15	30	4 Indian	Side Dish	TRUE
20	19 Homemade Tater Tots	10 Potatoes (Aloo),1 teaspoon Garlic powder	15	45	60	8 Continental	Appetizer	TRUE
21	20 Chettinad Vegetable Casserole - Che	tin; 1 cup rice - wash well in water, 12 small orio	15	45	60	4 Chetinad	Dimer	TRUE
22	21 Garlic Anla Chuhey - Letsun Anla	Chu 8 ania, 6 small crions, 10 bud garlic, 2 gree	5	10	15	3 Tamil Nadu	Side Dish	TRUE
23	22 Maharashtrian Kalodi Thalipeeth - Spi	ced 1/2 cup Gram four (besan), 1/2 cup Whole V	30	30	60	4 Maharashtrian	Indian Breakfast	FALSE
24	23 Homemade Healthy Subway Sandwic	h V/2 Submarine Bread (Subway Bread) - (flat b	15	15	30	4 Fusion	Orner	TRUE
25	24 South Indian Style Murungakkai Pony	al -5 Drumstick - cut into 3 inch pieces, 1/2 cup	30	15	45	4 South Indian	Side Dish	TRUE
26	25 Keser Mango Lessi - Seffron Mango	Las 34 cup Margo Pulp (Puree), 1 cup Curd (De	15	10	25	2 Indian	Snack	TR,E
27	26 Chettinad Sweet Paniyaram - Chettin	ad 11 cup idi dosa mixture, 4 tablespoons rice ra	5	20	25	10 Chetinad	Snack	TRUE
88	27 Mini Dal Samosa Curry / Sabzi	3 Cardamon (Elaichi) Pods/Seeds,6 Whole	5	30	35	3 North Indian	Lunch	TRUE
29	28 Spinach Garlic Bruschetta With Red	Bell 1 Baguette - sáced, 1 1/2 tablespoon Butter, 1	10	20	30	4 Italian	Appetizer	TRUE
10	59 Mini Dal Mature Chat - That Partiert S	Parti 1 run Whole Wheel Finur 1 run Al Purnose	50	15	65	30 Sinthi	Annetizer	TRIE

Figure 4.1 – Dataset Screenshot 1

sSugarFree	isHighProtein	isGluttenFree	isSattvic	isVegan	diet
TRUE	FALSE	FALSE	FALSE	FALSE	Diabetic Friendly
FALSE	FALSE	FALSE	FALSE	FALSE	Vegetarian
FALSE	TRUE	FALSE	FALSE	FALSE	High Protein Vegetarian
FALSE	FALSE	FALSE	FALSE	FALSE	Non Vegeterian
FALSE	FALSE	FALSE	FALSE	FALSE	Vegetarian
FALSE	TRUE	FALSE	FALSE	FALSE	High Protein Vegetarian
FALSE	FALSE	FALSE	FALSE	FALSE	Vegetarian
FALSE	FALSE	FALSE	FALSE	FALSE	Vegetarian
FALSE	FALSE	FALSE	FALSE	FALSE	Vegetarian
FALSE	FALSE	FALSE	FALSE	FALSE	Vegetarian
FALSE	FALSE	FALSE	FALSE	FALSE	Vegetarian
FALSE	FALSE	FALSE	FALSE	FALSE	Vegetarian
FALSE	FALSE	FALSE	FALSE	FALSE	Vegetarian
FALSE	TRUE	FALSE	FALSE	FALSE	High Protein Non Vegetaria
FALSE	FALSE	FALSE	FALSE	FALSE	Vegetarian
FALSE	FALSE	FALSE	FALSE	FALSE	Vegetarian
FALSE	FALSE	FALSE	FALSE	FALSE	Eggetarian
FALSE	FALSE	FALSE	FALSE	FALSE	Vegetarian
FALSE	FALSE	FALSE	FALSE	FALSE	Vegetarian
FALSE	FALSE	FALSE	FALSE	FALSE	Vegetarian
FALSE	FALSE	FALSE	FALSE	FALSE	Vegetarian
TRUE	FALSE	FALSE	FALSE	FALSE	Diabetic Friendly
FALSE	TRUE	FALSE	FALSE	FALSE	High Protein Vegetarian
FALSE	FALSE	FALSE	FALSE	FALSE	Vegetarian
FALSE	FALSE	FALSE	FALSE	FALSE	Vegetarian
FALSE	FALSE	FALSE	FALSE	FALSE	Vegetarian
FALSE	FALSE	FALSE	FALSE	FALSE	Vegetarian
FALSE	FALSE	FALSE	FALSE	FALSE	Vegetarian

Figure 4.2 – Dataset Screenshot 2

diet	allergies	totalCaloriesInCa instructions	URL	imgURL
Diabetic Friendly	None	106 To begin making the Ma	https://www.archa	https://www.archanaskitchen.co
Vegetarian	peanuts	212 To make tomato pulioge	r http://www.archa	https://www.archanaskitchen.co
High Protein Vegetarian	None	216 To begin making the Ra	http://www.archar	https://www.archanaskitchen.co
Non Vegeterian	None	1020 To begin making Gongu	http://www.archa	https://www.archanaskitchen.co
Vegetarian	None	41 To make Andhra Style A	https://www.archa	https://www.archanaskitchen.co
High Protein Vegetarian	None	212 To begin making Pudina	http://www.archar	https://www.archanaskitchen.co
Vegetarian	None	101 To begin making the Ud	http://www.archar	https://www.archanaskitchen.co
Vegetarian	lactose	522 To begin making the Bla	https://www.archa	https://www.archanaskitchen.co
Vegetarian	None	150 To prepare Spicy Crun	http://www.archar	https://www.archanaskitchen.co
Vegetarian	None	67 To make cauliflower lea	f http://www.archar	https://www.archanaskitchen.co
Vegetarian	None	155 To begin making the ho	https://www.archa	https://www.archanaskitchen.co
Vegetarian	lactose	160 To begin making the Ve	https://www.archa	https://www.archanaskitchen.co
Vegetarian	None	63 To begin making the An	https://www.archa	https://www.archanaskitchen.co
High Protein Non Vegetarian	None	232 To begin making the Aa	r https://www.archi	https://www.archanaskitchen.co
Vegetarian	lactose	112 To begin with Saunf Alo	https://www.archa	https://www.archanaskitchen.co
Vegetarian	None	90 To make South Indian C	http://www.archar	https://www.archanaskitchen.co
Eggetarian	lactose	156 To begin making the Ha	r https://www.archa	https://www.archanaskitchen.co
Vegetarian	None	24 To prepare gourd raita,	http://www.archar	https://www.archanaskitchen.co
Vegetarian	None	18 To begin making the Ho	http://www.archar	https://www.archanaskitchen.co
Vegetarian	None	359 To make Chettinad vege	https://www.archa	https://www.archanaskitchen.co
Vegetarian	None	70 To make Garlic Amla Cl	http://www.archa	https://www.archanaskitchen.co
Diabetic Friendly	None	100 To begin making the Ma	http://www.archar	https://www.archanaskitchen.co
High Protein Vegetarian	lactose	400 To begin making Subwa	http://www.archa	https://www.archanaskitchen.co
Vegetarian	None	176 To begin making Murun	chttp://www.archa	https://www.archanaskitchen.co
Vegetarian	lactose	360 To begin making Kesar	https://www.archa	https://www.archanaskitchen.co
Vegetarian	None	1960 To make the Chettinad	http://www.archar	https://www.archanaskitchen.co
Vegetarian	lactose	103 To begin making the Mi	http://www.archar	https://www.archanaskitchen.co
Vegetarian	lactose	89 To prepare Spinach Ga	r http://www.archar	https://www.archanaskitchen.co

Figure 4.3 - Dataset Screenshot 3

#### DATA PREPARATION

In this study, the data preparation stage is data preprocessing that prepares raw data before the following process (clustering and memory-based processes) to obtain clean data. One way to data pre-processing is to reduce irrelevant attributes.



6	A	В	С	D		E	F	G	Н	1	1	K	L	М	N	0	P	Q	R	S	T	U	1
foot	dD	пате	ingredient	prepTin	nel coo	okTimel I	totalTimel	servings	cuisine	course	isVeg	isSugarFn	e isHighProt	isGluttenF	i isSattvic	isVegan	diet	allergies	totalCal	or instructio	ri URL	imgURL	
	1	L Masala Ka	6 Karela (8		15	30	45		6 Indian	Side Dish	FALSE	TRUE	FALSE	FALSE	FALSE	FALSE	Diabetic	FrNone	1	06 To begin	n https://	wy https://	www.ar
	1	2 Spicy Tom	2-1/2 cup		5	10	15		3 South Inc	li Main Cour	TRUE	FALSE	FALSE	FALSE	FALSE	FALSE	Vegetari	ar peanuts	2	12 To make	ti http://v	w https://	www.ar
	1	4 Gongura C	500 grams		15	30	45		4 Andhra	Lunch	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	Non Veg	et None	10	20 To begin	n http://v	w https://	www.ar
	- 31	6 Pudina Kha	1 cup Rice		10	20	30		4 South Inc	fi: South Indi	TRUE	FALSE	TRUE	FALSE	FALSE	FALSE	High Pro	te None	2	12 To begin	r http://v	w https://	www.ar
	1	1 Homemad	250 grams		60	60	120		4 Fusion	High Prote	TRUE	FALSE	FALSE	FALSE	FALSE	FALSE	Vegetari	ar None	1	55 To begin	n https://	wv https://	www.ar
	1	3 Andhra Ste	2 cup Tam		0	30	30		4 Andhra	Lunch	TRUE	FALSE	FALSE	FALSE	FALSE	FALSE	Vegetari	ar None	1	63 To begin	n https://	wv https://	www.ar
	1	And fish so	600 grams		5	15	20		2 Bengali	Dinner	FALSE	FALSE	TRUE	FALSE	FALSE	FALSE	High Pro	te None	2	32 To begin	n https://	wy https://	www.ar
	1	5 Saunf Aloc	5 Potatoer		10	15	25		6 Punjabi	Lunch	TRUE	FALSE	FALSE	FALSE	FALSE	FALSE	Vegetari	ar lactose	1	12 To begin	whttps://	wittps://	www.ar
5	1	5 South Indi	2 onions, 1		20	20	40		4 South Inc	fi Side Dish	TRUE	FALSE	FALSE	FALSE	FALSE	FALSE	Vegetari	ar None	1	90 To make	S http://v	w https://	www.ar
1	1	7 Hariyali Eg	4 Whole E		15	20	35		4 Indian	Lunch	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	Eggetari	an lactose	1	56 To begin	n https://	wv https://	www.ar
2	2	L Garlic Aml	8 amía, 6 s		5	10	15		3 Tamil Na	d Side Dish	TRUE	FALSE	FALSE	FALSE	FALSE	FALSE	Vegetari	ar None		70 To make	Ghttp://v	w https://	www.ar
3	2	5 Chettinad	1 cup idí d		5	20	25	1	0 Chettina	d Snack	TRUE	FALSE	FALSE	FALSE	FALSE	FALSE	Vegetari	ar None	19	50 To make	tihttp://v	w https://	www.ar
4	2	7 Mini Dal S	3 Cardamo		5	30	35		3 North In	ś Lunch	TRUE	FALSE	FALSE	FALSE	FALSE	FALSE	Vegetari	ar lactose	1	03 To begin	n http://v	w https://	www.ar
5	3	D Asian Style	300 grams		10	15	25		4 Thai	Side Dish	TRUE	FALSE	TRUE	FALSE	FALSE	FALSE	High Pro	te peanuts	1	58 To begin	n http://v	w https://	www.ar
5	3	Chinese Da	1 tablespo		15	25	40		4 Chinese	Dessert	TRUE	FALSE	FALSE	FALSE	FALSE	FALSE	Vegetari	ar None	2	00 To begin	n http://v	w https://	www.ar
7	3	2 Sundakkai	1/2 cup Su		30	35	65		4 South Inc	fi Lunch	FALSE	TRUE	FALSE	FALSE	FALSE	FALSE	Diabetic	<b>Fr None</b>	1	40 To make	t/https://	wv https://	www.ar
8	3	5 Kerala palv	1 cup Ada,		15	75	90		4 kerala	Dessert	TRUE	FALSE	FALSE	FALSE	FALSE	FALSE	Vegetari	ar lactose	1	65 To begin	n https://	wv https://	www.ar
Э	3	7 Tindona Sa	1 cup Tird		20	15	35		1 Gujarati	Side Dish	TRUE	FALSE	FALSE	FALSE	FALSE	FALSE	Vegetari	ar None	1	02 To begin	n https://	wv https://	www.ar
3	3	8 Chettinad	1 Chicken		15	20	35		3 Chettina	d Appetizer	FALSE	FALSE	TRUE	FALSE	FALSE	FALSE	High Pro	te lactose	18	00 To begin	n https://	wv https://	www.ar
1	3	9 Palakottai	1 cup Jack		5	20	25		4 Tamil Na	d Side Dish	TRUE	FALSE	FALSE	FALSE	FALSE	FALSE	Vegetari	ar None		92 To begin	n http://v	w https://	www.ar
2		D Baingan Bl			25	35	60		4 Punjabi	Side Dish	TRUE	FALSE	FALSE	FALSE	FALSE	FALSE	Vegetari	ar lactose		12 To start t			
3	4	2 Pacha Mar	1 raw man		5	30	35		4 Tamil Na	d Side Dish	TRUE	FALSE	FALSE	FALSE	FALSE	FALSE	Vegetari	ar None	10	70 To make	≥ http://v	w https://	www.ar
4		3 Short Knot			10	45	55		4 Coorg	Lunch	TRUE	FALSE	FALSE	FALSE	FALSE	FALSE	Vegetari	ar None		20 To begin			
5	4	Whole Wh	1 cup Who		10	45	55		4 Coorg	Lunch	TRUE	FALSE	FALSE	FALSE	FALSE	FALSE	Vegetari	ar lactose	2	65 To begin	n http://v	w https://	www.ar
5	4	6 Cabbage a	3 Carrots (		10	20	30		4 Gujarati	? Lunch	TRUE	FALSE	FALSE	FALSE	FALSE	FALSE	Vegetari	ar None	2	42 To begin	n http://v	w https://	www.ar
7	4	8 Andhra Sto	1 kg Toma		15	60	75		5 Andhra	Side Dish	TRUE	FALSE	FALSE	FALSE	FALSE	FALSE	Vegetari	ar None	14	56 To begin	n http://v	w https://	www.ar
8	5	D Aamras - I	3 cups Ma		10	5	15		4 North In	ii Side Dish	TRUE	FALSE	FALSE	FALSE	FALSE	TRUE	Vegan	None	1	57 To begin	n http://v	w https://	www.an
9	5	Peerkanga	2 cups Rid		15	10	25		4 South Inc	is Side Dish	TRUE	FALSE	FALSE	FALSE	FALSE	FALSE	Vegetari	ar None	2	15 To begin	white://w	w https://	www.ac

Figure 4.4 – Modified Dataset 1 chances of receiving identical recommendations exist because a user's preferences might not alter frequently. The aforementioned approaches for recommendingailments or have diets do dealwith balancing dietspecifically with programs. certain

The proposed system recommends substituted foods according to nutrition and food parameters.

Based on the user's choices and the system's recommendations for personalized recipes. The algorithm of the suggested system made use of matrix factorization and latent feature vectors. By using tags that closely match the recommendations with users' tastes, prediction accuracy is attained. To balance the user's diet in accordance with his needs, the writers do not take nutrition into account. It is suggested to employ a content-based food recommender system that makes recipe recommendations based on the user's pre-established preferences[6]. Additionally,

#### **Decision Tree Classifier**

4	A 8	5	0	D	ε	F	G	н	1	1	K	L	М	N	0	р	Q	8	5	T	U	٧	W
31	55 Pan F	ried F2	tablespo	20	45	65		4 Continer	tz Appetizer	TRUE	FALSE	FALSE	FALSE	EALSE	FALSE	Vegetaria	rlactose	683	To begin	n.http://w	v https://w	ww.archarv	askitchen.com
32	58 Charv	ar Da 21	00 grans	15	60	75		4 Bengali	Lusch	TRUE	FALSE	FALSE	EALSE	FALSE	FALSE	Vegetaria	rlactose	222	Tobegin	n https://v	v https://w	ww.archan	askitchen.com
33	60 Bergi	alirai3	Mocil/ N	15	20	35		4 Bengali	Lunch	TRUE	FALSE	FALSE	FALSE	FALSE	FALSE	Vegetaria	rNone	170	To begin	n https://v	https://w	ww.archan	askitchen.com
34	.61 Beetr	00151	chakund	10	45	55		3 South In	ditunch	TRUE	FALSE	TRUE	FALSE	FALSE	FALSE	<b>High Prot</b>	e None	90	To make	chttp://w	v https://w	ww.archan	askitchen.com
35	63 Irania	in Bal 4	Whole E	10	30	40		4 Middle E	arlunch	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	Egyetaria	nNone	83	To begin	n https://v	titos//w	ww.archars	askitchen.com
36	64 Vegat	n Chit 1	cup Grat	20	30	50		4 North In	di North Indi	TRUE	FALSE	TRUE	FALSE	FALSE	FALSE	High Prot	elactose	480	Tobegin	n http://w	v https://w	ww.archan	askitchen.com
37	69 Toma	eto Rr 2	cups Tor	0	30	30		4 Continer	n: Side Dish	TRUE	FALSE	FALSE	FALSE	FALSE	FALSE	Vegetaria	r None	90	To begin	n http://w	v https://w	ww.archan	askitchen.com
38	70 Dubui	k Vat 2	Onions -	10	35	45		4 Maharas	hilunch	TRUE	FALSE	FALSE	FALSE	FALSE	FALSE	Vegetaria	r None	285	To begin	n http://w	v https://w	ww.archan	askitchen.com
39	71 Radis	h Sox 8	Mooli/ N	10	30	40		4 Indian	Side Dish	TRUE	FALSE	FALSE	FALSE	FALSE	FALSE	Vegetaria	None	203	To begin	n https://v	tetes//w	ww.archan	eskitchen.com
40	73 Mata	r Pan 1	CUP AS P	50	30	80		4 North in	di Lunch	TRUE	FALSE	FALSE	FALSE	FALSE	FALSE	Vegetaria	rlactose	330	To begin	n http://w	v https://w	ww.archark	askitchen.com
41	76 Karwi	ar 513 5	Hog Plun	25	25	50		4 Coastal	Kalusch	TRUE	FALSE	FALSE	EALSE	FALSE	FALSE	Vegetaria	r None	159	Tobegin	n http://w	v https://w	ww.archan	askitchen.com
42	77 Roast	ted C2	CUPS Cas	15	25	40		4 South In	di Dinner	TRUE	FALSE	FALSE	FALSE	FALSE	FALSE	Vegetaria	r None	25	To begin	n http://w	v https://w	ww.archan	askitchen.com
43	83 White	Pun 1-	1/2 cups	10	25	35		2 kerala	<b>Eurch</b>	TRUE	FALSE	FALSE	FALSE	FALSE	FALSE	Vegetaria	rlactose	87	To begin	n http://w	v https://w	ww.archan	askitchen.com
44	89 Murg	h Ma Si	00 gram I	150	30	180		4 North In	di Lunch	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	Non Vega	tlactose	264	Tobegin	n http://w	v https://w	ww.archan	askitchen.com
45	95 Kober	n Che 2-	1/2 cup	10	40	50		4 South In	dilutch	TRUE	FALSE	TRUE	FALSE	FALSE	FALSE	High Prot	e lactose	230	To make	ti https://v	v https://w	ww.archan	askitchen.com
46	98 Tamil	Ned 2	Whole E	5	30	35		4 Tami Na	d Appetiter	FALSE	FALSE	FALSE	EALSE.	FALSE	FALSE	Eggetaria	n None	207	To begin	ntmp://w	v https://w	ww.archen	askitchen.com
47	99 Chire	r Dud 2	cups Free	20	20	40		4 Bengal	Dessert	TRUE	FALSE	FALSE	FALSE	FALSE	FALSE	Vegetaria	actose	1840	To begin	n https://v	v https://w	ww.archan	askitchen.com
48	102 North	hind 1	cup Butt	10	30	-40		2 North In	di Lunch	TRUE	FALSE	TRUE	FALSE	FALSE	FALSE	High Prot	e mushroom	152	Tobegin	n https://v	teps//w	ww.archan	askitchen.com
49	105 Dhani	ie Ch 1	12 cup Cc	480	60	540		4 North In	di Main Cour	TRUE	FALSE	TRUE	FALSE	FALSE	FALSE	High Prot	e None	223	To begin	n https://v	vhttps://w	ww.archan	eskitchen.com
50	104 Capsi	cum 1	teaspoor	38	25	63		4 South In	dilusch	TRUE	FALSE	FALSE	FALSE	FALSE	FALSE	Vegetaria	rpeanuts	116	i To make	c.http://w	v https://w	ww.archan	eskitchen.com
51	105 Keral	a Styl 1	12 cup ra	360	20	380		4 kerala	Dinner	TRUE	FALSE	FALSE	FALSE	FALSE	FALSE	Vegetaria	r None	180	To make	K http://w	v https://w	ww.archan	askitchen.com
52	107 Dahi I	Wale 4	00 grans	65	15	80		6 North In	di Lunch	TRUE	FALSE	TRUE	FALSE	FALSE	FALSE	High Prot	e lactose	314	To begin	n https://v	v https://w	ww.archan	askitchen.com
\$3	109 Chick	en Til 4	Hard tac	20	0	20		4 Mexican	Dirner	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	Non Vege	tNone	410	To begin	n https://v	x https://w	ww.archan	askitchen.com
54	110 Masa	leua 2	trumpet	20	30	50		4 North in	di Side Dish	TRUE	FALSE	FALSE	FALSE	FALSE	FALSE	Vegetaria	r None	174	To make	ti https://v	v https://w	ww.archan	askitchen.com
55	113 Mutte	alkos SI	00 grams	15	30	45		3 South In	di Side Dish	FALSE	FALSE	FALSE	FALSE	TRUE	FALSE	No Onion	None	167	Tobegin	nhttp://w	v https://w	ww.archan	askitchen.com
56	114 Bengi	aluru 1	British (B	50	40	90		4 Kamatai	a Side Dish	TRUE	FALSE	FALSE	FALSE	FALSE	FALSE	Vegetaria	None	13	To begin	n http://w	v https://w	ww.archan	eskitchen.com
57	116 Lemo	n Frc 1	4 cup Le	600	0	600		3 Continer	nti Snack	TRUE	FALSE	FALSE	FALSE	FALSE	FALSE	Vegetaria	rlactose	114	To begin	n https://v	teps //w	ww.archan	askitchen.com
58	118 Dry St	uran 25	50 grams	15	20	35		4 North In	di Side Dish	TRUE	FALSE	FALSE	FALSE	FALSE	FALSE	Vegetaria	r None	187	To begin	n https://v	v https://w	ww.archan	askitchen.com
59	119 One F	ot Pr1	150 0000	15	25	40		4 South In	dilutch	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	Non Veze	tNone	403	To make	Chras/he	v https://w	ww.archara	askitchen.com

A supervised learning approach called a decision tree can be applied to classification and regression problems.

The method generates rules that can be represented by a tree structure using training data. It includes a root node, internal nodes, and leaf nodes like any other tree representation would. The internal node represents an attribute condition, the branches the condition's outcomes, and the leaf node the class Figure 4.5 - Modified Dataset 2 label.

Following the if-else style rules, you start at the root node at the top and work your way down to the leaf node to arrive at the classification. Your classification problem's class label is the leaf node where you end up [9].



Volume: 07 Issue: 05 | May - 2023

SJIF 2023: 8.176

ISSN: 2582-3930

1	121 Togari Bell 1 cup Arha	10	25	35	4 South India Lur	nch	TRUE	FALSE	TRUE	FALSE	FALSE	FALSE	High Prote None	140 To beain # https://ww/https://www.archanaskitchen.com
2	124 Palak Char 1 cup Char	20	30	50	4 North Indi Lui	nch	TRUE	FALSE	TRUE	FALSE	FALSE	FALSE	High Prote None	221 To bean mitto://wwwittos://www.archanaskitchen.com
53	125 Pasi Parup 1/2 cup Ye	15	30	45	4 South India Lur	nch	TRUE	FALSE	FALSE	FALSE	FALSE	FALSE	Vegetariar None	160 We begin rhttps://w/https://www.archanaskitchen.com
64	126 Maharashi 1 cup Sooj	60	25	85	4 Maharashi De	essert	TRUE	FALSE	FALSE	FALSE	FALSE	FALSE	Vegetariar lactose	192 To begin mittps://wwittps://www.archanaskitchen.com
65	127 Ragi Masa 2 cups Rag	20	20	40	4 South India Lur	nch I	FALSE	TRUE	FALSE	FALSE	FALSE	FALSE	Diabetic Fr None	56 To begin # https://ww https://www.archanaskitchen.com
66	128 Dry Sweet 2 Sweet Pr	10	10	20	4 kerala Sid	de Dish	TRUE	FALSE	FALSE	FALSE	FALSE	FALSE	Vegetariar None	212 To begin # https://www.archanaskitchen.com
67	129 Oreo Milks® Oreo bis	10	15	25	2 Continent: Sn	ack	TRUE	FALSE	FALSE	FALSE	FALSE	FALSE	Vegetariar lactose	262 To begin # http://ww https://www.archanaskitchen.com
68	130 Mexican G-4 Sweet cc	5	20	25	4 Mexican Sna	ack	TRUE	FALSE	FALSE	FALSE	FALSE	FALSE	Vegetariar lactose	265 To begin # http://ww https://www.archanaskitchen.com
69	131 Paneer Ma 1-1/2 cups	15	45	60	4 North Indi Dir	nner	TRUE	FALSE	TRUE	FALSE	FALSE	FALSE	High Prote lactose	451 To begin # https://ww https://www.archanaskitchen.com
70	132 Dhugare B 1 eggplant	5	20	25	4 Luckrowi Sid	de Dish	TRUE	FALSE	FALSE	FALSE	FALSE	FALSE	Vegetariar lactose	221 To make Dhttp://www.tttps://www.archanaskitchen.com
71	133 Rajasthani 1 cup Gran	10	30	40	4 Rajasthani Sid	de Dish I	FALSE	FALSE	FALSE	FALSE	TRUE	FALSE	No Onion lactose	385 To begin # https://ww https://www.archanaskitchen.com
72	134 Indo Chine 250 grams	10	60	70	3 Fusion Ap	ipetizer I	FALSE	FALSE	TRUE	FALSE	FALSE	FALSE	High Prote None	320 To begin # https://ww https://www.archanaskitchen.com
73	135 And fish sc 1 Aar Maa	10	40	50	2 Bengali Lur	nch I	FALSE	FALSE	TRUE	FALSE	FALSE	FALSE	High Prote None	315 To begin # https://ww https://www.archanaskitchen.com
74	136 Stem Letti, 1 cup Gree	10	20	30	4 South India Lur	nch I	FALSE	TRUE	FALSE	FALSE	FALSE	FALSE	Diabetic Fr None	29 To begin # https://www.archanaskitchen.com
75	137 Crispy Ma: 1 cup Raw	10	Ъ	35	3 Gujarati ? Sn	ack	TRUE	FALSE	TRUE	FALSE	FALSE	FALSE	High Prote peanuts	613 To begin # http://ww https://www.archanaskitchen.com
76	138 Maa Ki Dai 1 cup blacl	10	150	160	4 Punjabi Dir	nner	TRUE	FALSE	TRUE	FALSE	FALSE	FALSE	High Prote None	139 To make Mittp://www.tittps://www.archanaskitchen.com
77	139 Palak Musi 200 grams	10	30	40	4 North Indi Lur	nch	TRUE	FALSE	TRUE	FALSE	FALSE	FALSE	High Prote lactose	180 To begin # http://ww https://www.archanaskitchen.com
78	140 Dahi Bhall 4 Curd Vac	10	30	40	4 North Indi Lur	nch	TRUE	FALSE	TRUE	FALSE	FALSE	FALSE	High Prote peanuts	592 To begin # http://ww https://www.archanaskitchen.com
79	141 Wheat Gra50 grams V	10	20	30	3 Indian Sn	ack	TRUE	FALSE	FALSE	FALSE	FALSE	FALSE	Vegetariar None	25 To make til https://www.archanaskitchen.com
80	142 Chole Mas 1 cup Kabi	10	30	40	2 Punjabi Lur	nch	TRUE	FALSE	TRUE	FALSE	FALSE	FALSE	High Prote None	223 To make Chttps://wvhttps://www.archanaskitchen.com
81	143 Rajasthani 6 kachris, :	10	8	18	2 Rajasthani Sid	de Dish	TRUE	FALSE	FALSE	FALSE	FALSE	FALSE	Vegetariar None	157 To make R http://ww https://www.archanaskitchen.com
82	144 Fasting Por 7 Potatoes	20	60	80	4 North Indi Lur	nch	TRUE	FALSE	FALSE	FALSE	FALSE	FALSE	Vegetariar lactose	206 To make Ahttps://www.archanaskitchen.com
83	145 Katachi Ar2 cups lent	10	15	25	4 Maharashi Lui	nch	TRUE	FALSE	FALSE	FALSE	FALSE	FALSE	Vegetariar None	287 To make k https://www.archanaskitchen.com
84	146 Khara Obb 1 cup Who	30	45	75	4 Kamataka Lu	nch	TRUE	FALSE	FALSE	FALSE	FALSE	FALSE	Vegetariar lactose	175 We begin Hittps://wwihttps://www.archanaskitchen.com
85	147 Caulflowe 1 cabbage	10	35	45	3 Indian Dir	nner	TRUE	FALSE	FALSE	FALSE	FALSE	FALSE	Vegetariar None	164 To make c http://ww https://www.archanaskitchen.com
86	148 Kashmiri S 350 grams	5	30	35	4 Kashmiri Lur	nch	TRUE	FALSE	FALSE	FALSE	FALSE	FALSE	Vegetariar lactose	124 To begin # http://ww https://www.archanaskitchen.com
87	149 Aromatic \ 2 cups Rice	20	30	50	4 Indian Lur	nch	TRUE	FALSE	FALSE	FALSE	FALSE	FALSE	Vegetariar None	250 To begin # https://ww https://www.archanaskitchen.com
88	150 Spicy Curd Water - as	10	20	30	2 Indian Sid	de Dish	TRUE	FALSE	FALSE	FALSE	FALSE	FALSE	Vegetariar None	90 To make T https://ww/https://www.archanaskitchen.com
89	151 Mustard V 2 teasoool	20	30	50	4 Indian Sid	de Dish	TRUE	FALSE	FALSE	FALSE	FALSE	FALSE	Vegetariar lactose	200 To make til http://ww/https://www.archanaskitchen.com

- 4. The System will then recommend diet to the users into three categories (breakfast, lunch, dinner) based on input
- 5. The Users can choose from multiple recommended items and make their diet plan.
- 6. After selecting food items the system will calculate selected food calories and show user's comparison between how much calories they chosen against how much they need to consume daily. 7. Accordingly then the User's will make its diet plan.

Enter Name:	Raju	
Enter Gender	Male	-
Enter Address:	napur	
Enter Contact(Patient):	9654653132	
Enter Age(Patient):	55	
Enter Height	135	
Enter Weight:	n	E MA
Activity Level	lightly_active	
N. In		Pedat
		A YEAL )

Screenshot 1 : Data of User

# **Experiments and Results**

1. User's will enter the necessary information like their age, gender, weight etc. on the website.

2. The information will then go through the ML model in following manner:

2.1 Decision Tree Classifier is used to classify the food items and predict the food items based on the input.

Figure 4.6 - Modified Dataset 3 3. After analyzing all the data the system will respond by showing user's BMI and their

A Food Recommendation System (FRS) current state (Overweight, Underweight,

#### Healthy)



**Screenshot 2: Prediction of User** 

Based on the parameters the food is recommended. The value is matched in the dataset and prediction is shown on web page. The various factors like BMI, BMR, calories needed are calculated.

Table No.1 - Calorie Evaluation with personal parameters

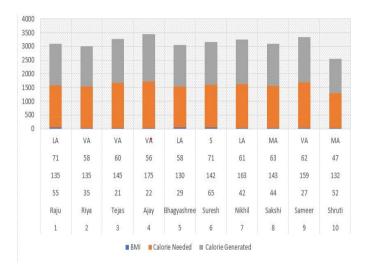


Volume: 07 Issue: 05 | May - 2023

SJIF 2023: 8.176

ISSN: 2582-3930

Sr. No.	Name of Person	Age	Height	Weight	Activity Level	BMI	Calorie Needed	Calorie Generated
1	Raju	55	135	71	LA	38.96	1541.05	1511
2	Riya	35	135	58	VA	31.82	1500.72	1469
3	Tejas	21	145	60	VA	28.54	1632.83	1612
4	Ajay	22	175	56	VA	18.25	1694.78	1726
5	Bhagyashree	29	130	58	LA	34.32	1513.31	1500
6	Suresh	65	142	71	S	37.2	1559.5	1563
7	Nikhil	42	163	61	LA	22.96	1601.73	1617
8	Sakshi	44	143	63	MA	30.81	1539.18	1534
9	Sameer	27	159	62	VA	24.52	1675.89	1641
10	Shruti	52	132	47	MA	26.97	1279.17	1250





# **Conclusion & Future Scope**

By the end of our project, we concluded that if Diet Recommendation system based on diabetics was properly designed, implemented, and evaluated, it could be used as an effective tool to improve nutrition and promote a healthy lifestyle.

An accuracy of 78.488 % for food recommendation was achieved for this system.

Thesaid accuracies will provide recommendations to the users. Accuracy can be improved by adding more data into the system. The increase in the number of data values will help in the formation for classifier more precisely, thereby improving the accuracy of the system.

In the future, we plan to expand our project to include an Android application that will recommend foods for specific diseases as well as a basic exercise plan and its maintenance.

### References

- Coughlin, Steven & Whitehead, Mary & Sheats, Joyce & Mastromonico, Jeff & Hardy, Dale & Smith, Selina. (2016). Smartphone Applications for Promoting Healthy Diet and Nutrition: A Literature Review. Jacobs journal of food and nutrition. 2. 021.
- Gonzalez13, Dariush Mozafarrian14, Salim Yusuf, Walter C. Willett12, and Barry M. Popkin15. "Food Consumption and its impact on Cardiovascular Disease: Importance of Solutions focused on the globalized food system." J Am Coll Cardiol 66, no. 14
- You, Yue & Doubova, Svetlana & Pinto Masis, Diana & Perez-Cuevas, Ricardo & BorjaAburto, Víctor & Hubbard, Alan. (2019). Application of machine learning methodology to assess the performance of DIABETIMSS program for patients with type 2 diabetes in family medicine clinics in Mexico. BMC Medical Informatics and Decision Making. 19.
- P. Pintér, L. Vajda and L. Kovács, "Developing a decision support system to determine carbohydrate intake of diabetic patients," 2012 IEEE 10th International Symposium on Applied Machine Intelligence and Informatics (SAMI), Herl'any, Slovakia, 2012, pp. 427-430
- 5. <u>https://apps.who.int/iris/bitstream/10665/20487</u> <u>1/1/9789241565257\_eng.pdf</u>
- M. Phanich, P. Pholkul and S. Phimoltares, "Food Recommendation System Using Clustering Analysis for Diabetic Patients," 2010 International Conference on Information Science and Applications, Seoul, Korea (South), 2010, pp. 1-8, doi:10.1109/ICISA.2010.5480416.
- Janakiraman, Bhavithra, and Saradha Arumugam. "Personalized Nutrition Recommendation for Diabetic Patients Using Optimization Techniques." INTELLIGENT AUTOMATION AND SOFT COMPUTING 26, no. 2 (2020): 269-280
- Dehais, Joachim, Marios Anthimopoulos, and Stavroula Mougiakakou. "Food image segmentation for dietary assessment." In Proceedings of the 2nd International Workshop on Multimedia Assisted Dietary Management, pp. 23-28. 2016.

9. <u>https://scikit</u>-

learn.org/stable/modules/generated/sklearn.tree. DecisionTreeClassifier.html