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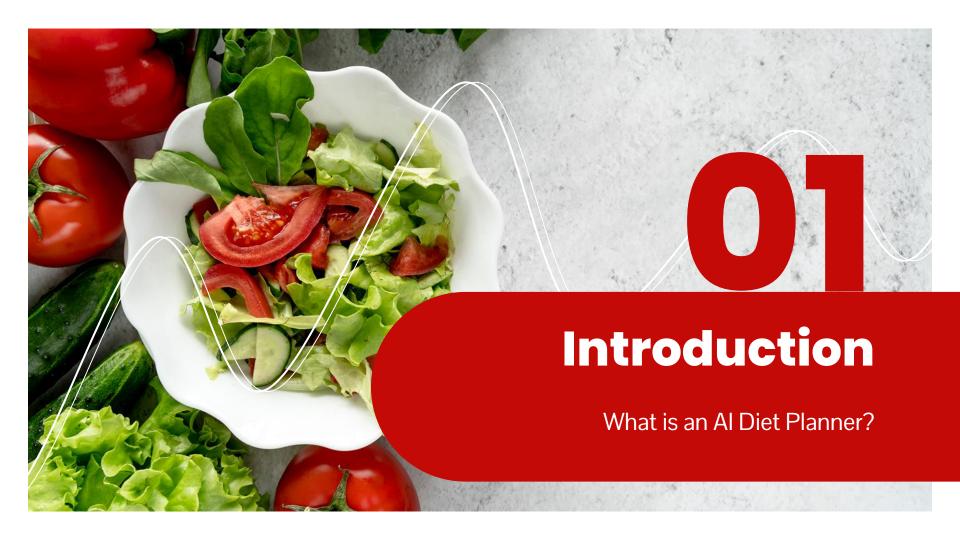
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Agent Function & Program

Goal-Based Al Agent Using Supervised Learning for Diet Planning 80

Project Goal

Al-Personalized Diet Planner for Weight Loss



Introduction

Al Diet Planner Agent is an intelligent system designed to help users plan personalized diets and activity schedules based on their health goals.

Tailors recommendations based on:

- Current and target weight
- Time to achieve the goal
- Age and gender
- O Activity level (e.g., high, medium, low)
- O Diet type (e.g., veg, non-veg, mixed)
- Food allergies





Problem Area:



Unrealistic Weight Goal Planning

Many users set goals without understanding what is achievable or healthy.



Allergy and Dietary Restriction Challenges

Manual plans often ignore personal allergies or food intolerances.



Generalized and Manual Planning

Traditional plans are not personalized and often do not meet individual needs.



Lack of Calorie Awareness

People often struggle to calculate or track their required daily calorie intake.



Inconsistent Meal and Activity Scheduling

Users fail to maintain regular and effective routines for diet and exercise.





- Personalized Meal Planning
- Activity Recommendations
- Probability Estimation Using Al
- Allergy Detection for Safer Meal Plans
- Structured Data Logging for Progress Tracking





Supervised Learning in Al Diet Planner

- Model is trained on input-output pairs (features + outcomes).
- Labeled data is used to train the model.
- The model learns to predict outcomes based on labeled data.

In our project:

Inputs: current weight, target weight, age, height, gender, activity level, etc.

Output Label: something like success probability (High, Moderate, Low)





Why Gaussian Naive Bayes Classifier?

- Naive Bayes works on Bayes Theorem, assuming that all features are independent.
- It handles both categorical (e.g., gender, diet type)
 and numerical (e.g., days, weight) inputs.
- Fast, lightweight, and works well with small datasets.
- Gaussian Naive Bayes is specifically designed for continuous data (weight, height)



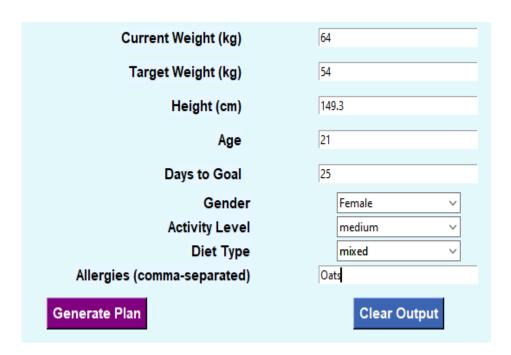
Component	Description
Performance Measure	 User goal achievement (e.g., reaching target weight) Accuracy of weight-loss success prediction Completeness and practicality of diet and activity plan User satisfaction and feedback based on results and warnings

Component	Description
Environment	 User-specific inputs: Current and target weight, Time to achieve the goal, Age and gender, Activity level, Diet type, Food allergies Predefined meal plans and calorie logic based on nutritional rules

Component	Description
Actuators	 Al model output (prediction of success) Meal plan generator tailored to input Calorie recommendation system Warning system for allergies

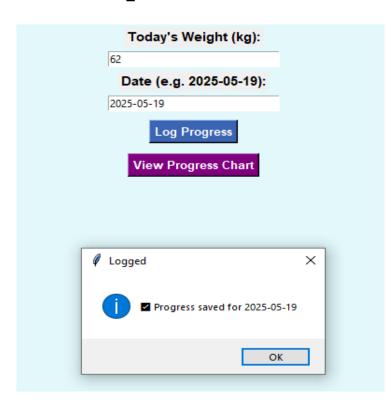
Description Component GUI Interface (Tkinter): Collects user data (weight, activity level, diet type, allergies). Sensors (User inputs) User Input Validation (Label Encoder): Preprocesses and encodes categorical data.

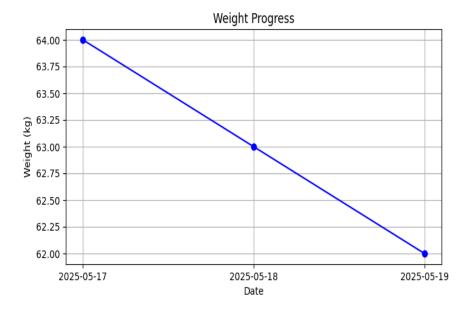
Output





Output







Proposed Solution

01

Predicts weight loss success probability using Naïve Bayes classification. 02

Calculates calorie deficit based on user's weight goals and activity level. 03

Generates adaptive meal plans tailored to dietary preferences.

04

Detects food allergies and issues warnings for safer meal choices.



Logs user history for tracking progress and refining future recommendations.



Agent Function

The agent processes user inputs, predicts success likelihood, and generates a structured meal and activity plan. It follows a goal-based approach, ensuring recommendations align with the user's weight loss objectives. It calculates, predicts, and then generates a goal-aligned plan for the user.



Step-by-Step Agent Function:

- Takes user input (weight, target, age, gender, height, activity, diet, allergies) via GUI
- 2. Encodes data using LabelEncoder for categorical inputs
- 3. Predicts success probability using Gaussian Naive Bayes model
- 4. Calculates BMI & BMR to understand current body state



Step-by-Step Agent Function:

- Generates meal plan based on diet type (veg, non-veg, mixed)
- 6. Adjusts activity time based on selected activity level
- 7. Checks allergies and gives warning messages
- 8. **Recommends calorie intake** based on weight gain or loss goal
- Logs results into a CSV file for future tracking f(cw,tw,d,at,dt) -> success_prediction



Agent Program

- 1) Model Training: Gaussian Naive Bayes is trained using user data (weight, diet, activity, etc.).
- 2) Label Encoding: Categorical data like diet, activity are encoded using LabelEncoder for model compatibility.
- 3) Prediction Logic: The model predicts success probability (High, Moderate, Low).
- 4) Plan Generator: Functions like generate_plan() calculate BMI, BMR, calorie intake, and activity minutes.



Agent Program

- 4) Allergy Detection: Checks meals against allergy inputs and raises warnings.
- 5) CSV Logging: Saves each user input and prediction result in diet_logs.csv
- 6) **GUI (Tkinter):** Interface to collect inputs and show output in a friendly layout.
- 7) Progress Report: User logs in his/ her progress daily





Statistical Entailment (AI Model Predictions)

- The Gaussian Naive Bayes classifier is used.
- It statistically predicts:

"What is the probability that this user will reach their weight goal?"

- Based on features like: current_weight, target_weight, days, activity, diet
- Example:

Input: 70kg → 65kg, 10 days, diet: mixed, activity: medium

Output: Moderate chance of success (via Naive Bayes probability)



Logical Entailment (Rule-Based Calculations)

Rule-based calculations like:

- BMI = weight / height²
- BMR using gender, weight, age, height
- Calorie intake = BMR * activity multiplier

Example: If BMI $< 18.5 \rightarrow$ Underweight

If target = current → "You are already at your goal"



Conditional Entailment (Adaptive Planning)

Meal suggestions change based on:

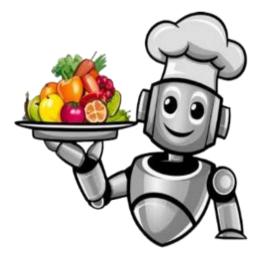
- Diet type
- Activity level
- Allergies

Custom warnings shown:

If allergy == "egg" and breakfast == "boiled egg" → show warning

Example: If user chooses low activity, activity time = 30 min

If allergies include "milk", and meal = "oats + milk" $\rightarrow \triangle$ warning





Project Goal:

- Develop an Al-based system for personalized diet and activity planning
- Help users achieve their target weight in a healthy timeframe
- Use user inputs like weight, days, activity level, diet type, and allergies
- Predict success likelihood using the Naive Bayes classifier
- Generate meal plans and calorie recommendations based on input
- Promote informed and healthy lifestyle decisions through automation



Phase 1: Research & Design:

- Define user requirements (weight management goals, dietary preferences, activity levels).
- Establish PEAS framework for structured Al agent design.
- Select Naïve Bayes classification for success probability estimation.



Phase 2: Al Model Development:

- Collect and preprocess training data (weight targets, activities, diets).
- Encode categorical variables using Label Encoder.
- Train Naïve Bayes model for probability estimation.
- Validate model accuracy with test data.



Phase 3: GUI & User Interaction (Tkinter):

- Build Tkinter interface for user input collection.
- Implement input validation for error handling.
- Design interactive result display with meal and activity plan output.



Phase 4: Meal Plan & Allergy Detection:

- Map diet preferences to structured meal recommendations.
- Implement allergy detection logic for safe meal choices.



Phase 5: Data Logging & Tracking

- Develop CSV logging system for tracking user inputs and predictions.
- Enable historical analysis for future refinements.



Phase 6: Testing & Refinements:

Conduct functional testing (edge cases, erroneous inputs).

Optimize prediction accuracy and meal plan recommendations.

Improve UI aesthetics for better user experience.



Phase 7: Deployment & Future Enhancements:

- Package the application for end-user usability.
- Explore alternative machine learning models for better predictions.
- Expand features (more diet options, deeper customization).



Thank You!

Do you have any questions?