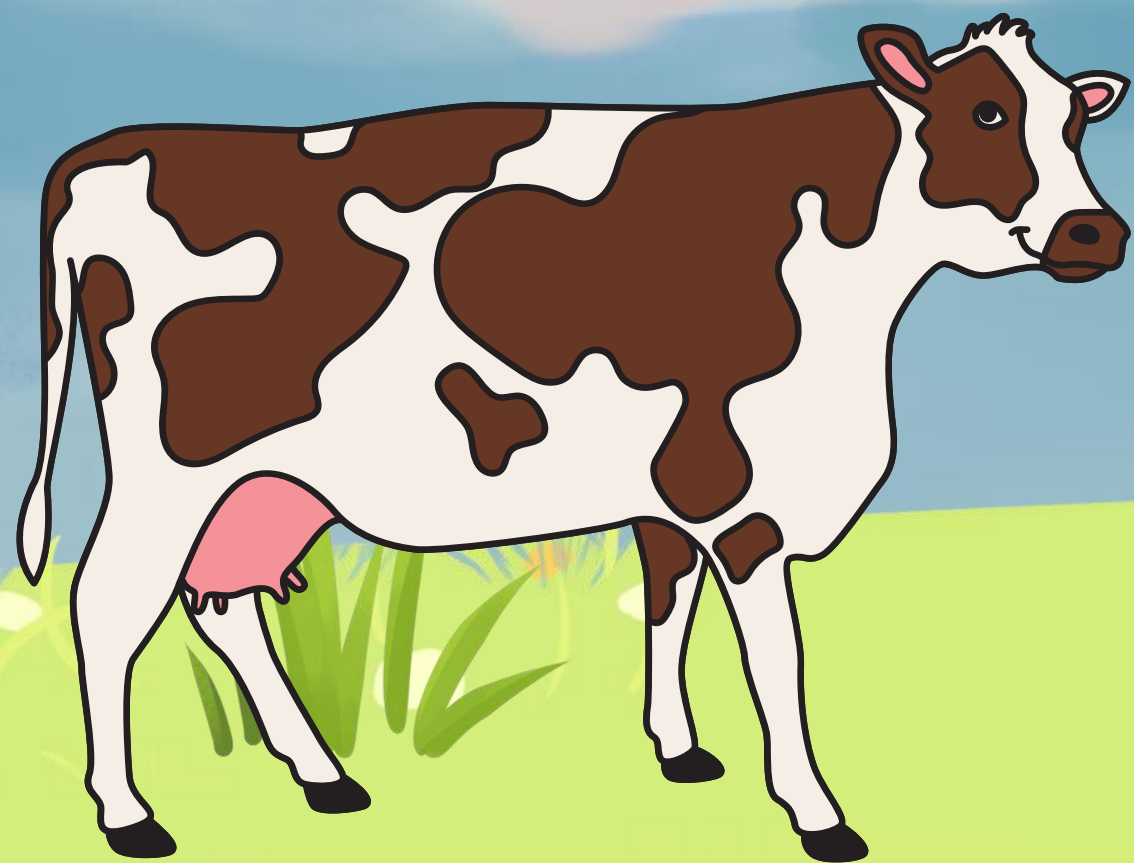


Cattle Disease Detection Project



INTRODUCTION

- **Context:** Importance of livestock health and early disease detection.
- **Challenge:** Difficulty in diagnosing cattle diseases with traditional methods.
- **Solution:** AI-powered disease detection tool based on symptoms.



PROBLEM STATEMENT

The cattle farming industry faces significant challenges due to the lack of early disease detection systems. Diseases in cattle can cause severe financial losses, reduced productivity, and even the spread of infections.

- Problem: Farmers often struggle to diagnose diseases in cattle on time, especially in rural areas with limited access to veterinary services.
- Impact: Delayed diagnosis leads to increased mortality rates, reduced milk production, and higher medical costs.



OBJECTIVES

- ✓ Early and accurate disease detection
- ✓ Integrating machine learning algorithms for better diagnoses.
- ✓ Provide a simple and effective tool for farmers and veterinarians.



PROPOSED METHODOLOGY

✓ Our methodology involves collecting data on symptoms and diseases, preprocessing the data, and implementing multiple machine learning models to predict diseases in cattle.

- Data Collection: Gather data on symptoms and diseases in cattle from veterinary and agricultural databases.
- Data Preprocessing: Clean and format the dataset to remove inconsistencies and ensure it's ready for machine learning training.
- Model Development: Implement and train the following algorithms:

K-Nearest Neighbors

Random Forest

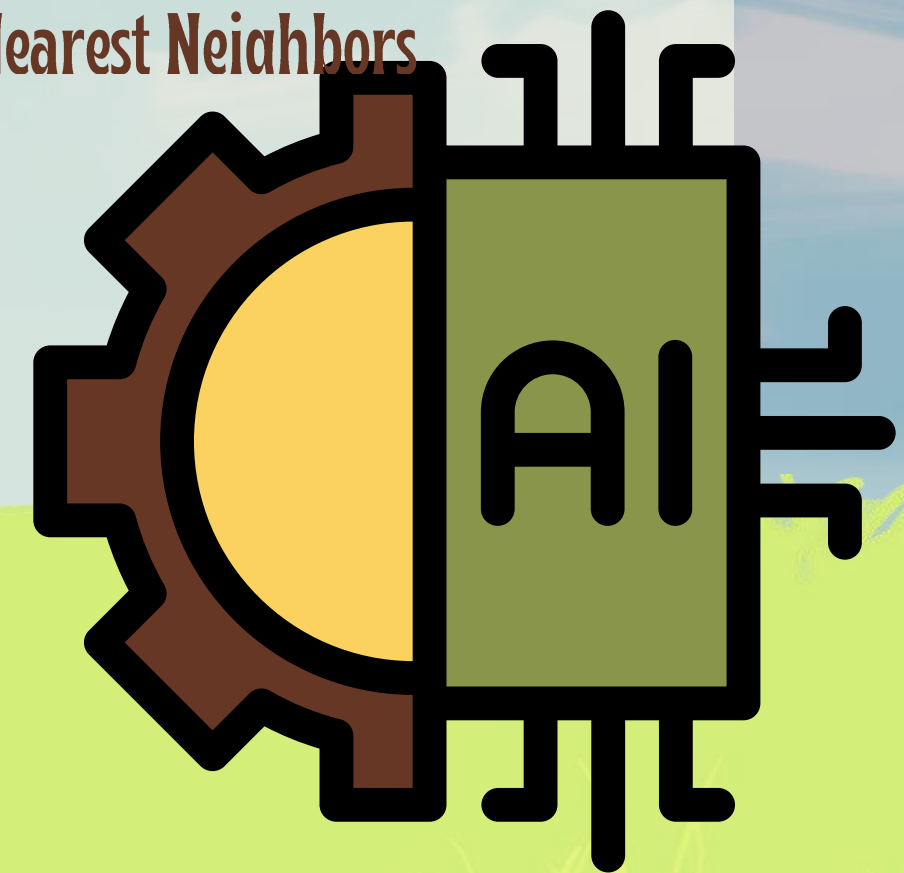
Naive Bayes

Decision Tree

Logistic Regression

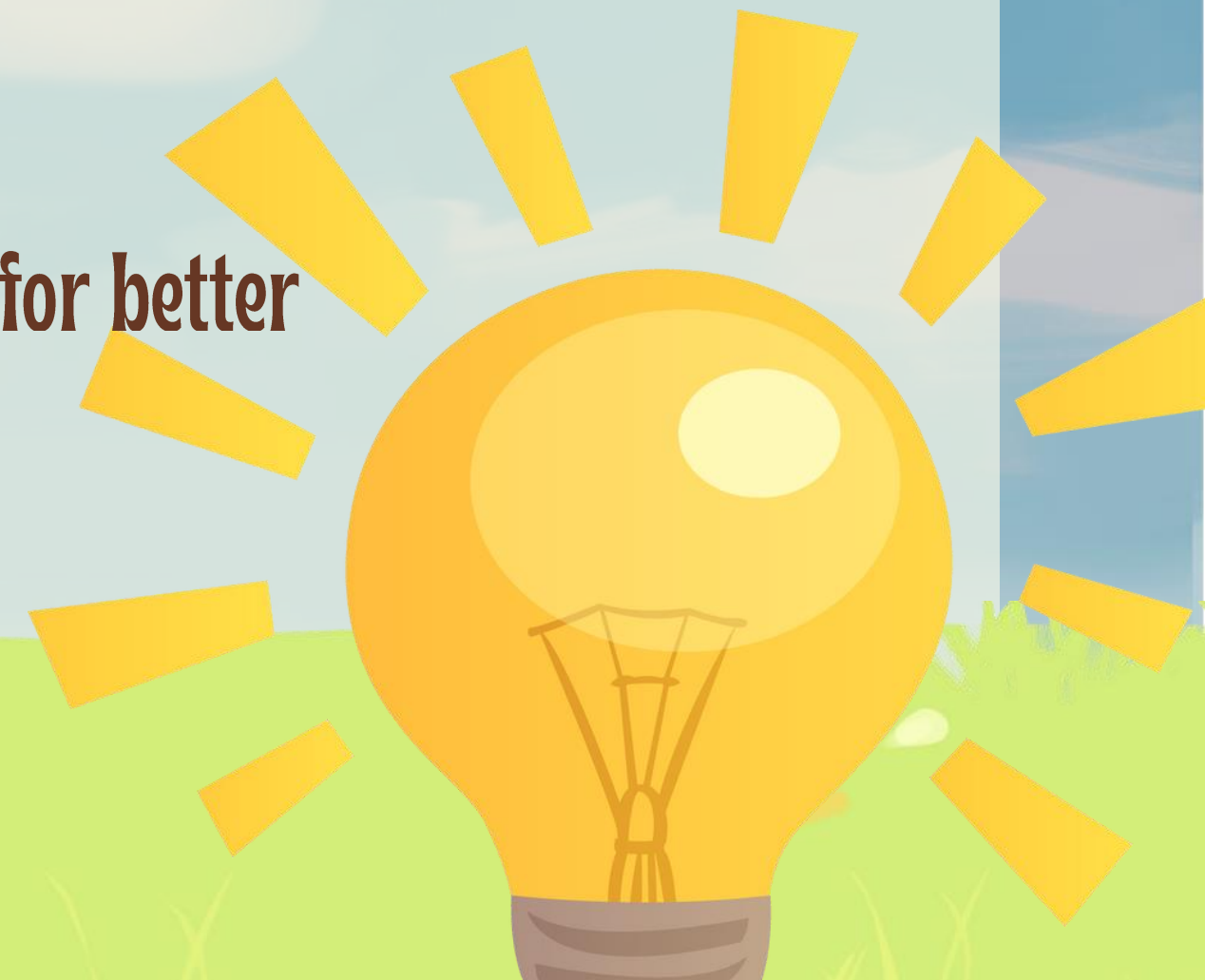
K-Nearest Neighbors

- Model Evaluation: Evaluate the models using accuracy, precision, recall, and F1 score metrics.
- User Interface: Develop a desktop interface where users input symptoms and receive prediction



SYSTEM FEATURES

- **Symptom Input:** Simple interface to input symptoms observed in cattle.
- **AI Diagnosis:** AI models (Decision Tree, Random Forest, KNN, etc.) analyze symptoms to provide disease predictions.
- **Learning Over Time:** Continuous improvement of AI models for better accuracy.



TECHNOLOGY STACK

- Frontend: Tkinter (Python)
- AI Algorithms: Decision Tree. Random Forest. Naive Bayes. KNN. Logistic Regression
- Backend: Python (Machine Learning Integration)
- Database: Storing symptoms and results for improved prediction accuracy (using SQL)



PROJECT PHASES

1. Data Collection
2. Algorithm Training
3. System Development & Integration
4. Testing & Validation
5. Deployment



RESULTS

The expected results include:

1. **Prediction Accuracy:** An effective machine learning model capable of accurately predicting diseases based on symptoms.
2. **User Feedback:** A simple and interactive interface for farmers to input symptoms and receive diagnosis results.
3. **Model Comparison:** Evaluation of different models to determine which one provides the highest accuracy and reliability in disease detection.

RELATED WORK

Several attempts have been made to use machine learning for disease prediction in animals, but these methods are not widely accessible or tailored to cattle farming needs.

- **Study 1:** A similar system for poultry disease prediction using decision trees and random forests showed promising results, but it was not adapted for cattle.
- **Study 2:** An application for livestock disease management focused on general veterinary care but lacked machine learning capabilities for specific disease prediction.
- **Gaps:** No comprehensive solution exists that combines machine learning models with a user-friendly interface specifically for cattle disease detection.

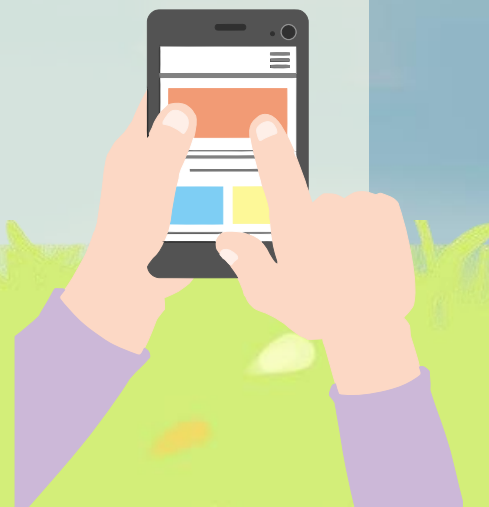
CONCLUSION

Key Takeaways:

- Early detection of cattle diseases can prevent economic losses and ensure better health management for livestock.
- Machine learning provides a powerful tool to predict diseases based on symptoms, making it accessible even in remote areas.
- Impact: The project can greatly benefit farmers by helping them manage cattle health more efficiently, improving productivity and preventing disease outbreaks.

Future Work:

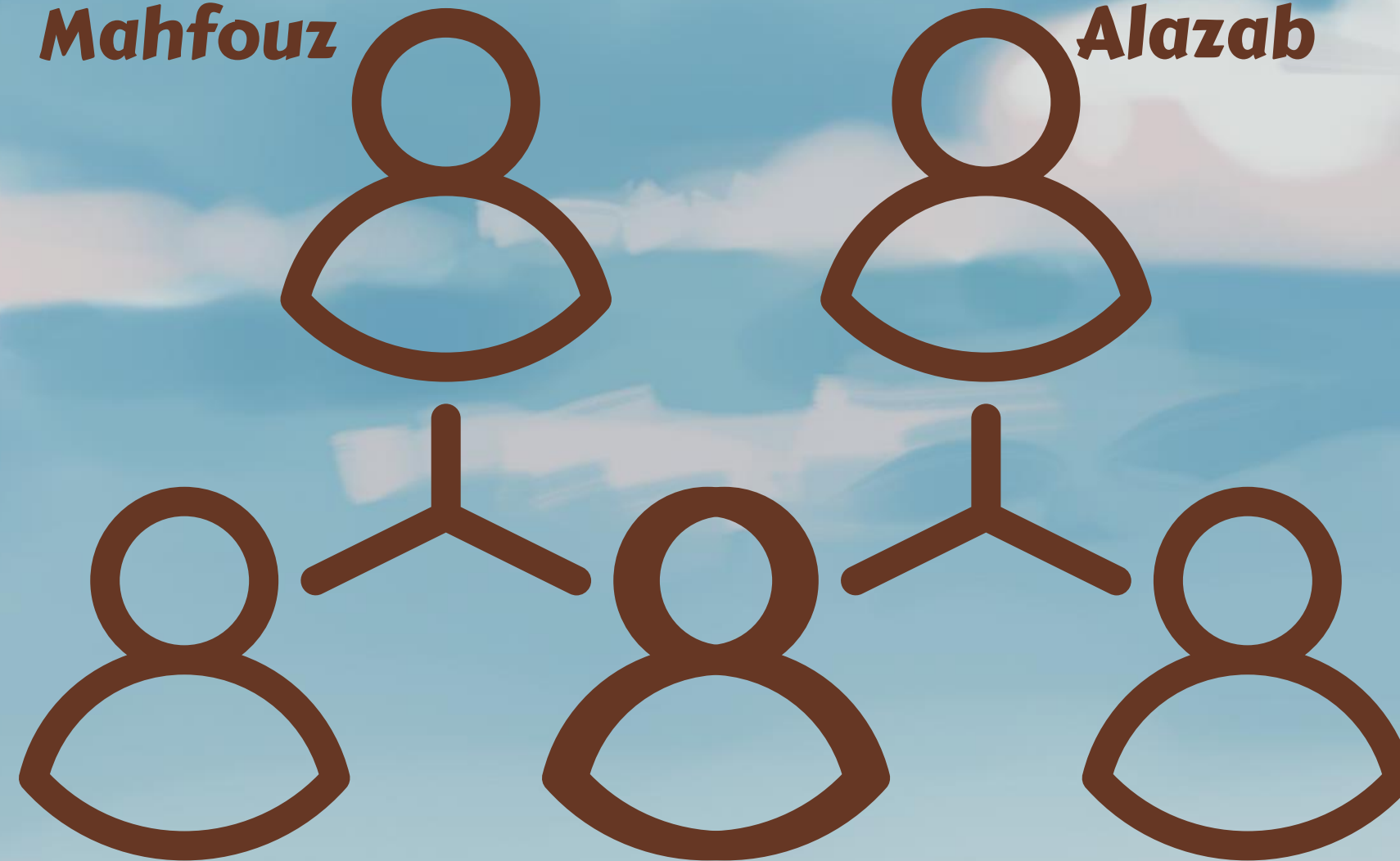
- Expand the dataset with more diseases and symptoms for better model accuracy.
- Develop a mobile version for wider accessibility.
- Implement real-time monitoring features using IoT for continuous disease detection.



THE TEAM

**Ahmed Hassan
Mahfouz**

**Ahmed Mustafa
Alazab**



**Aboelftouh abdelsamad
Aboelftouh**

Mohamed Eldesoky Fadl

Ebrahim Mahmoud AbdelGawad