

KABARAK UNIVERSITY

## SCHOOL OF COMPUTER SCIENCE AND BIOFORMATICS

## DEPARTMENT OF COMPUTER SCIENCE AND IT

## PROJECT PROPOSAL/RESEARCH 11

# **E-DAIRY PREDICTION AI SYSTEM**

A Project Report Documentation Submitted in the Department of Computer Science and IT in partial fulfillment of the degree of Computer Science

## MAURINE CHEPCHIRCHIR

BBIT/MG/3068/09/20

SUPERVISOR: MR. CLEOPHAS MOCHOGE

**BACHELOR OF BUSINESS INFORMATION TECHNOLOGY** 

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# **DECLARATION AND APPROVAL**

Project documentation submitted to the department of Computer Science in the school of Science Engineering and Technology partial fulfillment of the requirements for the award of the Bachelor's degree in Business Information Technology at Kabarak University. This project is my original work and has not been presented anywhere as a pre-requisite for the award of bachelor's degree in any other University.

| Name:                |                    |  |  |
|----------------------|--------------------|--|--|
| MAURINE CHEPCHIRCHIR | BBIT/MG/3068/09/20 |  |  |
|                      |                    |  |  |
|                      |                    |  |  |
| Signature            | Date               |  |  |

# APPROVAL BY THE SUPERVISOR

| his project has been submitted for examination with my approval as University |      |  |
|---|------|--|
| Supervisor:   |      |  |
| Signature   | Date |  |
|   |      |  |

Name: MR. CLEOPHAS MOCHOGE

#### **DEDICATION**

I dedicate this project to the Almighty God for His divine protection and guidance throughout the development of the E-Dairy AI Prediction System; without His mercy and grace, this project would not have been possible. To the countless individuals whose collective efforts and unwavering support have brought the E-Dairy AI Prediction System from conceptualization to fruition. My special gratitude goes to the dairy farmers, veterinarians, and system administrators who generously shared their experiences, insights, and time, forming the bedrock of this system. The commitment and collaboration of these stakeholders have been indispensable in crafting a solution that addresses real-world challenges in dairy farm management.

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# LIST OF ACRONYMNS

**RBAC**- Role Based Access Control

**MFA-** Multi-Factor Authenticator

UI- User-Interface

**UML-** Unified Modeling Language

**DFD-** Data Flow Diagram

#### **ABSTRACT**

In the dynamic landscape of dairy farming, the need for efficiency, accuracy, and predictive capabilities has never been more crucial. Traditional dairy management processes often suffer from delays, lack of data-driven insights, and manual operational errors. Current dairy systems are plagued by inefficiencies, leading to inconsistent milk production, opaque health monitoring, and increased susceptibility to errors. Manual handling of data and the lack of real-time tracking contribute to a lack of trust and effectiveness in the dairy management of many farms across the nation. These challenges highlight the urgent need for a comprehensive, technology-driven solution.

The existing challenges in dairy management are multifaceted; cumbersome paperwork, time-consuming communication, and a lack of standardized processes contribute to inefficiencies and ineffectiveness in the related sectors. Additionally, the absence of a centralized tracking mechanism makes it difficult to monitor the progress and health of dairy cattle in real-time, hindering timely decision-making.

Therefore, we developed the E-Dairy AI Prediction System, which aims to revolutionize the dairy farming landscape by introducing a seamless, transparent, and efficient solution for emerging and existing farms aiming for tangible daily improvements towards meeting their production and health objectives. By leveraging advanced technology, we seek to eliminate manual bottlenecks, enhance communication, and instill confidence in stakeholders through real-time tracking and predictive analytics.

The developed system is built on a robust and scalable technological framework, utilizing the latest advancements in artificial intelligence, cloud computing, and data analytics. A user-friendly interface facilitates easy navigation for both farmers and dairy managers. Features such as automated health monitoring, real-time status updates, and data-driven insights are integrated to ensure a comprehensive and efficient dairy management ecosystem.

#### **CHAPTER ONE**

#### **INRODUCTION**

### 1.1 Background of the study

The dairy industry is a vital part of the agricultural sector, significantly impacting the economy and serving as a primary source of income for millions of farmers worldwide. Despite its importance, the industry faces several challenges that impede its efficiency and productivity. Traditional dairy farming methods, which rely on manual processes and lack real-time data, are increasingly inadequate in meeting the market's growing demands. Therefore, there is an urgent need for a more sophisticated and predictive approach.

# Challenges in Traditional Dairy Farming

Conventional dairy farming depends heavily on manual record-keeping and observation, which are both labor-intensive and prone to human error. These practices often result in inconsistencies in milk production, inadequate health monitoring of cattle, and delayed responses to issues like disease outbreaks or nutritional deficiencies. The absence of standardized processes and real-time data further exacerbates these problems, making it difficult for farmers to make timely and informed decisions.

Additionally, the dairy sector struggles with inefficiencies in feeding, breeding, and overall herd management. Farmers face challenges in optimizing feed rations, monitoring reproductive cycles, and predicting milk yields, all of which are crucial for maintaining a profitable and sustainable operation. The lack of a centralized system for tracking and analyzing data across these various aspects of dairy farming contributes to suboptimal performance and increased operational costs.

### The Role of Technology in Modernizing Dairy Farming

Technological advancements, particularly in artificial intelligence (AI) and data analytics, offer promising solutions to these challenges. AI can process large amounts of data quickly, providing actionable insights that enhance decision-making and operational efficiency. Integrating AI with dairy farming enables the prediction of milk production, real-time monitoring of cattle health, and optimization of feeding and breeding programs.

Cloud computing and Internet of Things (IoT) devices further enhance AI's capabilities by enabling continuous data collection and analysis. Sensors and wearable devices can monitor cattle's vital signs, track their movements, and even detect early signs of illness. This real-time data can be analyzed to predict health issues before they become severe, allowing for timely interventions and reducing overall morbidity and mortality rates in dairy herds.

### The E-Dairy AI Prediction System

To address the urgent need for modernized dairy farming, we developed the E-Dairy AI Prediction System. This system aims to revolutionize the dairy industry by providing a comprehensive, technology-driven solution that addresses the inefficiencies of traditional methods. The E-Dairy AI Prediction System leverages advanced AI algorithms, cloud computing, and IoT technologies to create a robust and scalable platform for dairy management.

The system features a user-friendly interface that facilitates easy navigation for farmers and dairy managers. Key features include automated health monitoring, real-time status updates, predictive analytics for milk production, and data-driven insights into feeding and breeding programs. By centralizing data and providing real-time tracking, the E-Dairy AI Prediction System enhances transparency, accountability, and efficiency in dairy farming operations.

# **Benefits and Implications**

The adoption of the E-Dairy AI Prediction System has the potential to significantly transform the dairy industry. Farmers can benefit from increased milk yields, improved cattle health, and reduced operational costs. The system's predictive capabilities enable proactive management, reducing the risk of disease outbreaks and enhancing overall herd productivity. Additionally, the real-time data and insights provided by the system can help farmers make informed decisions, leading to better resource allocation and more sustainable farming practices.

In summary, the E-Dairy AI Prediction System represents a major advancement in the dairy industry, addressing the longstanding challenges of traditional farming methods. By harnessing the power of AI and modern technology, this system offers a promising solution for enhancing efficiency, transparency, and productivity in dairy farming, ultimately contributing to a more sustainable and profitable industry.

#### 1.2 Problem statement

Traditional dairy farming methods suffer from inefficiency and errors due to reliance on manual processes and lack of real-time data, highlighting the need for a modern, technology-driven solution to improve productivity and decision-making.

## 1.3 Objectives

## 1.3.1 General Objective

The main objective of developing this system is to predict dairy outputs which will result in maximum farm production in the dairy industry

# 1.3.2 Specific Objectives

- To investigate the challenges faced in the dairy industry and identify the proposed solutions.
- To design a system that will be able to predict dairy outputs based on the inputs.
- iii. To implement a data-driven decision making system in the dairy sector.
- iv. To test the effectiveness of an E-Dairy AI Prediction System.

# 1.4 Research Questions

- i. What are the challenges of existing systems implemented in the dairy industry?
- ii. How can the E-Dairy AI Prediction System be designed?
- iii. How can the E-Dairy AI Prediction System be implemented?
- iv. How can the E-Dairy AI Prediction System be tested?

#### 1.5 Justification of the study

- Increased Efficiency: The E-Dairy AI Prediction System automates data collection and health monitoring, reducing manual labor and errors, which boosts overall farm productivity.
- ii. **Better Animal Health**: Real-time data and AI insights enable early detection of cattle health issues, allowing for timely interventions and improved herd welfare.

- iii. **Data-Driven Decisions:** Advanced analytics provide valuable insights into feeding, breeding, and milk production, supporting informed decision-making and resource optimization.
- iv. **Enhanced Transparency:** Real-time tracking and centralized data management improve transparency and accountability in dairy operations, ensuring effective monitoring.
- v. **Sustainability and Cost Savings:** Optimized processes and better herd management lead to more sustainable practices and reduced operational costs.

### 1.6 Significance of the study

- i. **Advancement in Dairy Farming:** The E-Dairy AI Prediction System introduces innovative technology to the dairy sector, demonstrating how AI and real-time data can transform traditional farming methods.
- ii. **Economic Benefits: By** improving efficiency and productivity, the system boosts profitability for dairy farmers, supporting the growth and sustainability of the dairy industry.
- iii. **Livestock Health and Welfare:** The system's real-time health monitoring and predictive features ensure better care for cattle, resulting in healthier herds and higher animal welfare standards.
- iv. **Efficient Resource Use:** The system provides data-driven insights that enable more efficient use of resources like feed and labor, promoting environmentally sustainable farming practices.

## 1.7 Scope and limitation of the study

- i. **Data Integration:** The system integrates data from sensors, wearables, and manual entries to track cattle health, milk production, and feeding.
- ii. **Predictive Analytics:** AI algorithms analyze this data to forecast milk yields, detect health issues, and optimize feeding and breeding.
- iii. **Real-Time Monitoring:** It offers real-time updates on cattle health and herd management, enabling prompt decisions.
- iv. **Data Quality:** The system's accuracy relies on the quality and completeness of the data collected.

v. **Adoption Challenges:** Farmers with limited tech skills or resources might face difficulties adopting the new system.

#### **CHAPTER TWO**

#### LITERATURE REVIEW

#### 2.1 Introduction

In recent years, the intersection of agriculture and dairy farming has become a focal point of scholarly inquiry. As advancements in technology continue to reshape the landscape, questions surrounding maximum outcomes persist, necessitating a comprehensive examination of the existing literature. This literature review aims to provide a critical synthesis of key findings, identify gaps and offer insights into the evolving study. Within this exploration, we will navigate through dairy advancements, drawing upon established theories to guide our analysis. As we embark on this, it is imperative to underscore the timeliness and significance of this review in the context of technology integration into farming. As the dairy industry grapples with the challenges of optimizing milk production, managing herd health and navigating through resource utilization, the application of artificial intelligence (AI) to predict and optimize outcomes has gained considerable attention.

### 2.2 Overview of the main topic

E-Dairy systems is continually evolving with ongoing developments and innovations in the field. Farmers often choose systems based on the specific needs of their operations and the features offered by different providers. In the past few years. There have been various existing E-Dairy Systems that typically incorporate a combination of hardware, software and data analytics to provide farmers with dairy management techniques. In the late 19th century milking machines were introduced to automate milking process and ease hand milking labour. In the mid-20<sup>th</sup> century, introduction of bulk milk cooling systems allowed dairy farmers to store and transport milk at lower temperatures, preserving its quality and reducing spoilage. Later on that century, computerized record-keeping systems were adopted in the dairy farms. Finally, the 21st century has witnessed technologies such as sensors, GPS and analytics which have allowed farmers to monitor and manage various aspects like environmental conditions. Clearly, the integration of technology into dairy family has become weighty in enhancing efficiency, sustainability

## 2.3 Review of specific objectives

# 2.3.1 Investigation of Challenges

**Nimbalkar and Singh (2021):** In their study "*Dairy farming innovations for productivity enhancement*," Nimbalkar and Singh identified key challenges in the dairy industry, such as milk production fluctuations, disease outbreaks, and feed inefficiencies. They proposed solutions including precision farming technologies, enhanced herd management practices, and the use of predictive analytics to anticipate production trends and health issues.

**Kaushik and Bhadauria** (2023): Their paper "Modeling the challenges of technology adoption in dairy farming," explored the impact of climate change on dairy production and the role of technology in mitigating these effects. They emphasized the importance of real-time monitoring systems for feed management and environmental control to enhance dairy productivity and sustainability.

#### 2.3.2 System Design

Neves & Miranda. (2022): In "). *Predictive Models of Dairy Cow Thermal State*," Neves and colleagues developed a predictive model using machine learning techniques to forecast milk yield based on variables such as feed type, barn conditions, and animal health. Their model demonstrated

significant accuracy in predicting daily milk outputs, highlighting AI's potential in dairy management.

Hernandez and Gonzalez (2024): Their study "Dairy farming in the era of artificial intelligence: "focused on designing an AI system that integrates various data inputs, including environmental conditions, feed composition, and milking frequency, to predict dairy production. Their findings highlighted the system's effectiveness in optimizing resource use and improving overall milk yield.

## 2.3.3 System Implementation

Sharma and Verma (2023): In their work "Implementation of AI Systems in Agriculture," Sharma and Verma discussed the practical aspects of deploying AI systems in agricultural settings, including dairy farms. They provided a detailed case study on the implementation of an AI-based monitoring system that improved herd health management and production efficiency.

Csaba and Márton (2020): Their article "Digitization and big data system of intelligence management in smart dairy farming," detailed the steps and considerations for implementing an AI prediction system. They emphasized the importance of stakeholder engagement, robust data infrastructure, and iterative testing to ensure successful deployment and adoption of the technology.

## 2.3.4 System Testing

**Fuentes and Gonzalez (2020)**: In "Evaluating AI Systems in Dairy Production," Fuentes and Gonzalez conducted a comprehensive evaluation of an AI prediction system's effectiveness in various dairy farms. Their study showed significant improvements in milk yield prediction accuracy and operational efficiency, demonstrating the system's practical benefits.

Salamone and Atashi (2022): Their research "Prediction of first test day milk yield using historical records in dairy cows," involved testing an AI system's performance across different dairy farm environments. They found that the system effectively adapted to varying conditions and provided reliable predictions, leading to better decision-making and productivity gains [These studies collectively highlight the potential and impact of AI-driven systems in addressing dairy industry challenges, optimizing production, and enhancing overall farm management. They provide a solid

foundation for developing and implementing the E-Dairy AI Prediction System, ensuring that it is both innovative and grounded in proven research methodologies.

#### 2.4 Lessons learnt from Previous Studies

- Addressing Industry Challenges: Studies stress the need to clearly identify dairy industry
  problems, such as manual inefficiencies and lack of real-time data, to develop targeted
  solutions.
- ii. **Design for Accuracy:** Accurate prediction relies on robust data and algorithms. Previous research highlights the importance of reliable data collection and predictive accuracy for effective system design.
- iii. **Implementation Issues:** Successful technology adoption requires careful planning, including phased rollout, user training, and technical support, as past studies show challenges in system integration and user adaptation.
- iv. **Effectiveness Testing:** Evaluating system performance in real-world settings is crucial. Previous research recommends assessing productivity, user satisfaction, and efficiency to validate and refine the system.

# 2.5 Methods of identifying feature selection techniques

- i. **Correlation Analysis:** Examine the correlation between features and the target variable to identify and retain relevant features.
- ii. **Recursive Feature Elimination (RFE):** Use RFE to iteratively remove the least important features based on model performance, highlighting the most significant ones.
- iii. **Feature Importance from Models:** Leverage models like decision trees that provide importance scores for each feature to prioritize the most relevant ones.
- iv. **Mutual Information:** Evaluate the mutual information between features and the target variable to identify features with high informational value.
- v. **Statistical Tests:** Apply tests like chi-square or ANOVA to assess the significance of features in relation to the target variable.

#### 2.6 Evaluation of the correlation between optimal features

i. **Data Collection**: Collect data from the E-Dairy AI Prediction System, including all potential features and target variables related to dairy outputs.

- ii. **Data Preprocessing**: Clean and preprocess the data to handle missing values, normalize numerical features, and encode categorical variables.
- iii. **Correlation Matrix**: Generate a correlation matrix to visualize relationships between features, identifying those with high correlations that might be redundant.
- iv. **Heatmap Visualization**: Use heatmaps to display the correlation matrix for easier identification of strong correlations between features and the target variable.
- v. **Feature Pair Analysis**: Analyze highly correlated feature pairs to determine if they can be reduced to a single feature without losing significant information.

## 2.7 User security awareness level

- i. **Training and Education**: Regularly train users on security best practices, recognizing threats, and safe system usage.
- ii. **Guidelines and Policies:** Provide clear guidelines on password management, data handling, and device security, ensuring users understand their responsibilities.
- iii. **Access Controls:** Enforce strict access controls, granting users only the necessary permissions, and regularly updating these controls to prevent unauthorized access.
- iv. **Multi-Factor Authentication (MFA):** Implement MFA for system access to add an extra layer of security, protecting against unauthorized entry even if credentials are compromised.
- v. **Security Audits:** Conduct regular audits to identify vulnerabilities and ensure compliance with security standards.
- vi. **Incident Response Plan:** Develop and share a plan outlining steps to take in case of a security breach, ensuring users know how to report incidents.
- vii. **Awareness Campaigns:** Run periodic campaigns to keep security practices and emerging threats top-of-mind for users.

### 2.8 Design framework and Prototype design

## 2.8.1 Design framework

The E-Dairy AI Prediction System is designed with a robust architecture to ensure comprehensive data integration, processing, and user interaction. The system collects data from various sources, including sensors, wearable devices, and manual inputs. These sources provide essential information such as temperature, humidity, motion, and health metrics of the cattle. The data

collection module consolidates this information through IoT devices that transmit data to a central server. This data is securely stored in scalable databases, ensuring encryption both at rest and in transit to protect sensitive information.

The processing module employs advanced AI algorithms and data analytics tools to analyze the collected data. This real-time processing enables the generation of accurate predictions for milk yield, cattle health, and optimal feeding schedules. The results are then displayed through user-friendly web and mobile applications, providing farmers with interactive dashboards, real-time monitoring, alerts, and detailed reports. The system's user interface is designed to be intuitive, offering easy navigation and clear visualization of key metrics. To maintain security, the system incorporates multi-factor authentication, role-based access control, and regular security audits.

Data flow within the system begins with continuous data ingestion from all sources. APIs and data pipelines facilitate seamless data flow, which is then securely stored in the central database. The machine learning models process this data to deliver real-time analytics and predictions. The processed data is outputted on the user interfaces, and critical alerts and notifications are sent to users as needed. This structured data flow ensures that farmers have access to timely and actionable insights.

Feature selection is critical for the system's predictive accuracy. Features such as milk yield, health indicators, and feeding patterns are identified and selected using statistical methods and domain expertise. Techniques like correlation analysis, Recursive Feature Elimination (RFE), and expert validation ensure that only the most relevant features are used in the prediction models. This careful selection process enhances the system's overall effectiveness.

#### 2.8.2 Prototype design

The prototype design phase involves creating initial wireframes to outline the layout and navigation flow. Detailed design mockups are then developed to focus on usability and aesthetics. An interactive prototype simulates the user experience, allowing for user feedback and iterative improvements. This ensures that the final product is user-friendly and meets the practical needs of dairy farmers.

Implementation of the system involves both backend and frontend development. Servers and databases are set up, AI algorithms are implemented, and data processing logic is developed on

the backend. The frontend involves creating responsive web and mobile interfaces compatible across various devices and browsers. Integration testing ensures seamless operation between backend and frontend components, verifying data consistency and system performance.

Extensive testing is conducted to ensure the system's reliability. Unit testing checks the functionality of individual components, while integration testing examines the entire system. User testing involves actual users to gather feedback and refine the system. This thorough testing process ensures that the system is robust and user-friendly.

Finally, the deployment phase begins with a pilot deployment in a real-world setting to monitor performance and gather feedback. Based on the results, the system is gradually scaled up for full deployment. Continuous monitoring, maintenance, and updates ensure that the system remains efficient and effective in enhancing dairy farm productivity and management.

This design framework provides a structured approach to developing, testing, and deploying the E-Dairy AI Prediction System, ensuring it meets the needs of dairy farmers and significantly improves operational efficiency.

#### 2.9 Summary

The literature review for the E-Dairy AI Prediction System underscores the dairy industry's key challenges, such as variable milk yields, cattle health problems, and inefficient feeding strategies. Studies suggest that AI and machine learning can address these issues by offering predictive insights and real-time monitoring. Research supports the use of IoT devices and wearables for enhanced data collection, leading to more precise predictions. Additionally, the review highlights the need for user-friendly interfaces and strong security protocols to ensure effective adoption and reliability. Overall, the review justifies the need for an integrated AI-based system to enhance dairy farm management, improve cattle health, and boost productivity.

#### **CHAPTER THREE**

#### RESEARCH METHODOLOGY

#### 3.1 Introduction

The research methodology outlines the approach and techniques used to investigate the development and effectiveness of the E-Dairy AI Prediction System. This study employs a mixed-methods approach, combining quantitative and qualitative research to gain a comprehensive understanding of the system's performance and impact. Quantitative methods include data collection from sensors and wearable devices, statistical analysis of predictive accuracy, and evaluation of system efficiency through metrics such as milk yield improvements and health monitoring outcomes. Qualitative methods involve user feedback and expert consultations to assess usability, system integration, and overall satisfaction.

To ensure robust findings, the research methodology includes a series of structured phases. Initial phases involve designing the system prototype, followed by pilot testing in real-world conditions to gather preliminary data. This is succeeded by a full-scale deployment and extensive evaluation of system performance. The methodology also incorporates iterative feedback loops, allowing for continuous refinement based on user experiences and emerging needs. This comprehensive approach aims to validate the system's effectiveness in optimizing dairy farm operations and enhancing productivity.

#### 3.2 Research design

The research design for the E-Dairy AI Prediction System utilizes a mixed-methods approach, integrating both qualitative and quantitative research methodologies to thoroughly understand the system's impact and user requirements.

Qualitative methods include stakeholder interviews, focus groups with dairy farmers, and analysis of operational workflows to gain detailed insights into user experiences, challenges, and expectations. These qualitative insights are complemented by quantitative data collected through surveys and questionnaires, which measure user satisfaction, preferences, and system usability. Additionally, iterative prototyping and usability testing are employed to refine the system based on user feedback, ensuring that the final product meets practical needs and enhances usability.

This mixed-methods approach provides a comprehensive view by triangulating data from multiple sources, thus improving the validity and reliability of the research outcomes. The synthesized results from both qualitative and quantitative analyses will form a solid foundation for the development and optimization of the E-Dairy AI Prediction System, ensuring it effectively addresses the needs of dairy farmers and improves dairy farm management.

### 3.3 Population of study

The population of study for the E-Dairy AI Prediction System includes a diverse range of stakeholders involved in dairy farm operations and management. This population is categorized into key groups such as dairy farmers who use the system for daily operations, veterinarians monitoring cattle health, feed suppliers managing nutritional inputs, IT staff responsible for system maintenance, and system administrators overseeing its functionality.

Additionally, the study extends to agricultural extension services, private sector firms engaged in dairy technology, and research institutions focused on agricultural innovation. This broad scope ensures a comprehensive exploration of perspectives and needs across different sectors within the dairy industry. The diverse demographic and organizational makeup of the population reflects the variety of entities involved in dairy farming. By engaging with representatives from these varied groups, the research aims to capture a thorough understanding of the challenges and requirements within the dairy sector, providing insights that will guide the development of an effective and inclusive E-Dairy AI Prediction System to address the needs of all stakeholders involved.

### 3.4 sampling technique

Choosing an appropriate sampling technique is crucial for collecting representative data during the analysis and design of the E-Dairy AI Prediction System. The following sampling techniques were employed to gather requirement elicitation for the developed system:

**Stratified Sampling**: The study population was divided into distinct strata based on relevant characteristics, such as farm size, roles within the dairy industry (e.g., farmers, veterinarians, feed suppliers), and geographical location.

Rationale: This approach ensures representation from different segments of the
population, allowing for a more accurate analysis of each stratum based on their specific
requirements.

**Cluster Sampling**: The population was further divided into clusters, such as specific regions or groups of dairy farms. Randomly selected clusters included all members within those clusters in the sample.

• **Rationale**: This method is practical when the population is naturally grouped, and it is more efficient to sample entire clusters rather than individual members.

These sampling techniques ensure that the diverse needs and perspectives of all stakeholders are accurately represented, providing a solid foundation for the design and development of the E-Dairy AI Prediction System.

#### 3.5 Methods of Data Collection

In the research for the E-Dairy AI Prediction System, a combination of questionnaires, surveys, and interviews played a pivotal role in gathering comprehensive information from stakeholders.

**Surveys**: Surveys were designed to collect quantitative data, using structured questions to assess user satisfaction levels, preferences, and perceptions related to current dairy farming practices. Respondents, including dairy farmers, veterinarians, and feed suppliers, provided numerical ratings or selected predefined options to quantify their experiences. The survey reached at least 150 respondents, from which the analysis was drawn.

**Interviews**: Interviews were conducted to delve deeper into the qualitative aspects of the research. One-on-one interviews with key stakeholders, such as dairy farmers and veterinarians, were designed to elicit detailed narratives regarding their experiences with existing farming systems, challenges faced, and expectations for improvement. Open-ended questions allowed for rich, qualitative insights, providing a nuanced understanding of user needs and system requirements. At least 100 interviewees were eligible to participate.

Questionnaires: Open-ended and closed-ended questionnaires were used to gather opinions and preferences about the system. Open-ended questionnaires, administered to a sample population of around 30 dairy farmers and 50 veterinarians, allowed respondents to provide detailed feedback on their experiences and needs. Closed-ended questionnaires, with predetermined answer options, were given to a total of 200 respondents, including farmers, veterinarians, and IT personnel, to

quantify specific aspects of their experiences and requirements. The analysis of these questionnaires provided the basis for concluding the need and design requirements for the system.

By combining these data collection methods, the research ensured a comprehensive understanding of stakeholder needs and preferences, guiding the development of an effective and user-centric E-Dairy AI Prediction System.

### 3.6 System analysis and Design

The system analysis and design phase for the E-Dairy AI Prediction System involves a detailed examination and structured planning to ensure the system aligns with the needs of various stakeholders in the dairy farming sector. This phase focuses on capturing both functional and non-functional requirements from dairy farmers, veterinarians, and feed suppliers, and translating these into a well-defined system architecture. Key activities include designing data models, developing user-friendly interfaces, and planning the integration of IoT devices and AI algorithms. The objective is to create a scalable, efficient, and user-centric system that enhances dairy farm management through precise predictions and actionable insights.

#### 3.6.1 Waterfall model

The development of the E-Dairy AI Prediction System follows the Waterfall Model, a linear and sequential approach ensuring each phase is completed before moving on to the next. This model provides a structured framework, making it suitable for the systematic development of the system.

**Requirements Analysis**: In the initial phase, extensive consultations with stakeholders, including dairy farmers, veterinarians, and feed suppliers, are conducted to gather detailed requirements. This involves understanding their needs, challenges, and expectations. The collected requirements are then documented in a comprehensive specification report, which serves as the foundation for the subsequent stages.

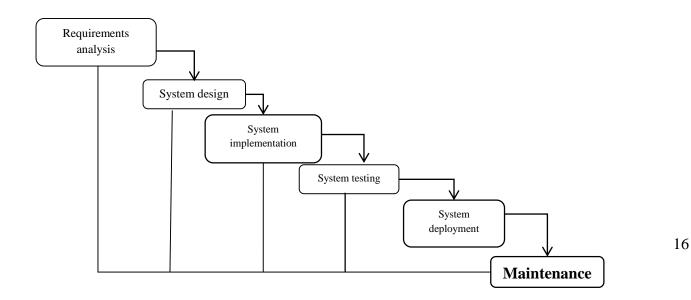
**System Design**: Based on the documented requirements, the system design phase involves creating the architectural blueprint of the E-Dairy AI Prediction System. This includes designing data models, defining system architecture, and planning the integration of IoT devices and AI algorithms. The goal is to ensure that the design meets all specified requirements and provides a clear guide for the development team.

**Implementation**: In this phase, the actual coding and development of the system take place. The design specifications are translated into functional code, with developers focusing on building the system's various modules and components. This includes developing the AI algorithms for predictions, creating user interfaces, and setting up data integration pipelines.

Integration and Testing: Once the implementation is complete, the system undergoes rigorous integration and testing. This stage ensures that all components work together seamlessly and that the system functions as intended. Various testing methods, such as unit testing, integration testing, and system testing, are employed to identify and fix any issues or bugs. **Deployment**: After successful testing, the system is deployed in a real-world environment. This involves installing and configuring the system on the users' premises, training stakeholders on how to use it, and ensuring that it operates smoothly in the production environment. Pilot deployments may be conducted initially to gather feedback and make any necessary adjustments before full-scale deployment.

**Maintenance**: The final phase involves ongoing maintenance and support to ensure the system continues to function effectively over time. This includes regular updates, bug fixes, performance enhancements, and providing user support. Continuous monitoring and feedback collection help in making iterative improvements to the system, ensuring it remains aligned with user needs and technological advancements.

By following the Waterfall Model, the development of the E-Dairy AI Prediction System is carried out in a structured and systematic manner, ensuring each phase is thoroughly completed before proceeding to the next, thus minimizing risks and ensuring a high-quality outcome.



# Fig 3.1: Waterfall Model

# 3.6.2 Context Diagram

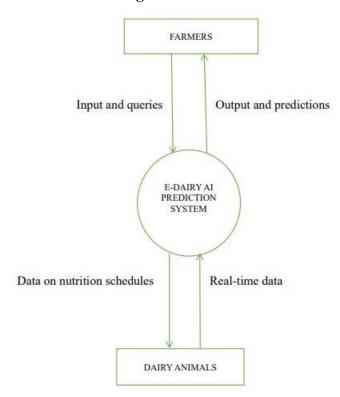


Fig 3.2 Context Diagram

# 3.6.3 Level 1 Data Flow diagram

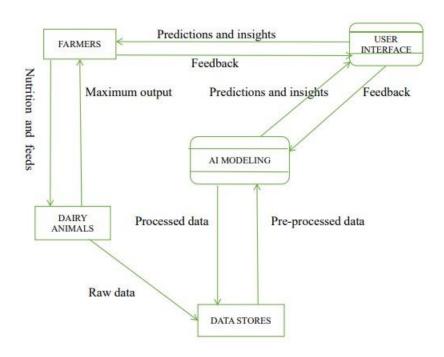


Fig 3.3 Level 1 Data Flow diagram

## 3.6.4 Use Case Diagram

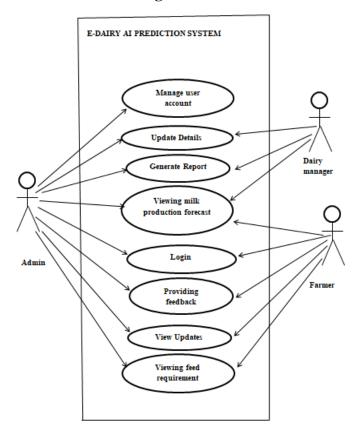


Fig 3.4 Use Case Diagram

## 3.7 Implementation

During the implementation phase, the coding and development process takes place, adhering to established standards and best practices. The designed database schema is implemented, populated with relevant data, and fortified with security measures. The user interface is constructed based on the finalized design, ensuring an intuitive and accessible experience for all users. The integration of system components is meticulously conducted to guarantee seamless communication and functionality. A pilot implementation is initiated to gather user feedback and identify any issues, which are then addressed before full-scale deployment.

#### 3.8 Research Ethics

To uphold ethical standards, participant confidentiality and anonymity were strictly maintained throughout the research process. Informed consent was obtained from all participants before data collection, clearly outlining the study's purpose, potential risks, and their rights. Any personal

information collected was securely stored, and only aggregate data was reported to ensure anonymity. Ethical approval was sought from the relevant institutional review board to ensure compliance with ethical guidelines in researching human subjects. This methodology ensures a comprehensive and ethical approach to the development and evaluation of the E-Dairy AI Prediction System.

#### **CHAPTER 4**

#### SYSTEM IMPLEMENTATION AND DEPLOYMENT

#### 4.1 Introduction

The System Implementation and Deployment phase is a crucial milestone in the software development life cycle (SDLC). It involves translating design and development efforts into a fully operational system that meets the needs of dairy farmers, veterinarians, and feed suppliers. This section outlines the specific tasks, considerations, and best practices related to implementing and deploying the E-Dairy AI Prediction System.

### 4.2 System Architecture

System architecture defines the structure and behavior of the E-Dairy AI Prediction System. It outlines the components of the system, their interactions, and the rules governing these interactions. The system is a web-based application that relies on internet connectivity. It comprises two main user types: the system administrators, who manage and maintain the backend data and functionalities, and the end-users, including dairy farmers and veterinarians, who interact with the system through a user-friendly dashboard to access predictive insights and manage their data.

#### 4.3 User Interface Design

User interface design involves creating the visual and interactive aspects of the E-Dairy AI Prediction System. The goal is to design an interface that is both intuitive and engaging, ensuring ease of use for all users. The design focuses on creating a seamless experience, with clear navigation, accessible features, and a visually appealing layout that enhances user satisfaction and efficiency in accessing predictive data and system functionalities.

## 4.3.1 Components

 Data Collection Module: IoT devices such as temperature sensors, humidity sensors, RFID tags, and milk meters are used to collect data on breed, age, lactation period, health status, feed type, feed amount, temperature, humidity, barn condition, milking frequency, and veterinary visits.

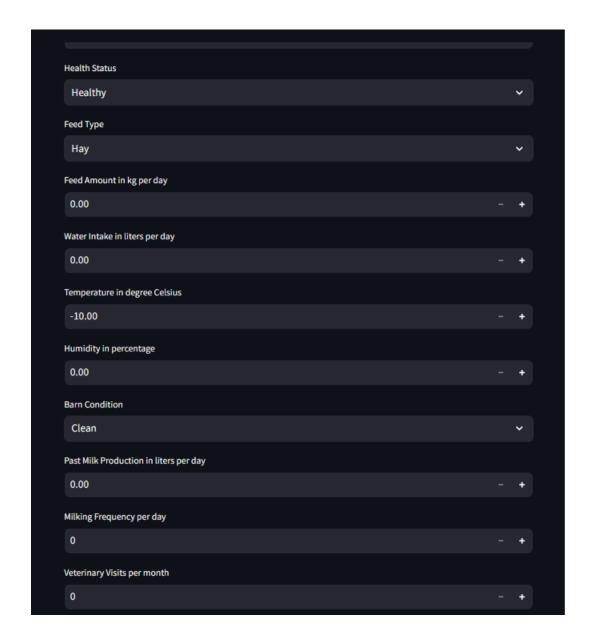


Fig 4.1 Data Collection Module

- **Data Processing Module**: This module preprocesses the collected data, including cleaning, normalization, and transformation, to make it suitable for analysis.
- Prediction Module: Machine learning models are developed to predict milk yield, milk production distribution, and milk production per breed.

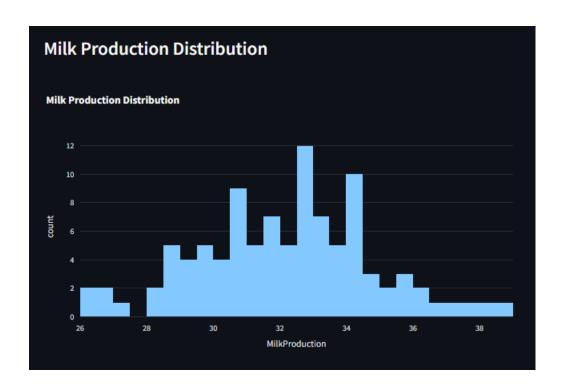


Fig 4.2 milk production distribution

• **Data Visualization Module**: Dashboards and visual tools are provided for farmers to visualize data and predictions.

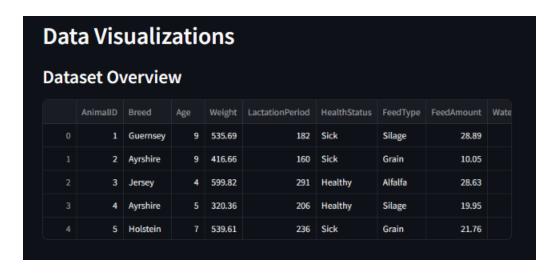


Fig 4.3 Data Visualization Module

• **User Interface**: A web-based dashboard and a mobile application are provided for farmers to interact with the system, view predictions, and receive alerts.



Fig 4.4 User Interface

# **4.4 Development Environment**

The development environment includes both hardware and software components necessary for building and testing the system.

#### 4.4.1 Hardware

- **Servers**: Sreamlit instances are used for hosting the system.
- **IoT Devices**: Temperature sensors, humidity sensors, RFID tags, and milk meters are deployed on the farm.

## 4.4.2 Software

- **Programming Languages**: Python is used for data processing and machine learning, while JavaScript is used for the user interface.
- **Development Frameworks**: TensorFlow for machine learning, MongoDB for the backend, and React for the frontend.
- Databases: MySQL is used for storing processed data.
- **Version Control**: Git is used for version control.

## 4.5 Implementation

The implementation process involves setting up the data collection infrastructure, developing the data processing and prediction modules, and creating the user interface.

#### 4.5.1 Data Collection

- **Setup**: IoT devices are installed on the farm and configured to transmit data to the central server.
- Configuration: Data transmission protocols are established to ensure reliable data flow.

## 4.5.2 Data Processing

- **Cleaning**: Scripts are developed to clean the raw data by removing outliers and handling missing values.
- **Normalization**: Data is normalized to ensure consistency.
- **Storage**: Processed data is stored in the MySQL database.

# **4.5.3 Model Development**

• **Training**: Historical data is used to train machine learning models for predicting milk yield, milk production distribution, and milk production per breed.



Fig 4.5 Model Development Training

- Evaluation: Model performance is evaluated using metrics such as accuracy and F1 score.
- **Tuning**: Hyper parameters are tuned to optimize model performance.

# 4.6 Back-End Development

# 4.6.1 Purpose and Role

Backend development involves creating the server-side logic and infrastructure that powers the entire system. In the context of our procurement system, the backend plays a crucial role in

handling data storage, business logic, security, and communication between the frontend (user interface) and the database.

- **API Development**: RESTful APIs are developed using Flask to handle data requests and responses between the frontend and backend.
- **Database Management**: MySQL is used to manage and store data. SQL Alchemy is used as an ORM (Object-Relational Mapping) tool to interact with the database.



Fig 4.6 Database Management

- Authentication and Authorization: Implement user authentication and authorization using JWT (JSON Web Tokens) to secure the system.
- **Data Processing Pipelines**: Data pipelines are developed to automate the process of data collection, cleaning, and storage.
- **Scalability**: The backend is scalable to handle increasing amounts of data and user requests by using load balancers and auto-scaling groups on AWS.

### 4.7 Deployment

The deployment process involves setting up the production environment and deploying the system components.

### 4.7.1 Deployment Environment

- **Production Setup**: Streamlit.app is configured for the production environment.
- Configuration: Servers and cloud services are configured to host the system.

## 4.7.2 Deployment Steps

- Data Collection Module: IoT devices are deployed and connected to the central server.
- Data Processing and Prediction Modules: These modules are deployed on the server.
- Data Visualization Module: Visualization tools are deployed and integrated with the user interface.
- **User Interface**: The web-based dashboard and mobile application are deployed and made accessible to users.

### 4.8 Testing and Validation

Testing and validation ensure the system functions correctly and meets user requirements.

#### 4.8.1 Unit Testing

• Components: Individual components are tested for functionality using unit tests.

### 4.8.2 Integration Testing

• **System Integration**: All components are tested together to ensure seamless integration.

#### 4.8.3 User Acceptance Testing

Feedback: Farmers test the system and provide feedback, which is used to make necessary
adjustments.

### 4.9 Maintenance and Updates

A maintenance plan is established to ensure the system remains functional and up-to-date.

### 4.9.1 Regular Maintenance

- **Monitoring**: System performance is monitored continuously.
- **Issue Resolution**: Any issues that arise are addressed promptly.

### **4.9.2 Updates**

- **Feature Implementation**: New features are implemented based on user feedback.
- **Model Updates**: Machine learning models are updated with new data to improve accuracy.

## 4.9.3 Conclusion

This chapter has outlined the implementation and deployment of the eDairy AI prediction system, highlighting the importance of a well-structured approach to ensure the system's success.

#### **CHAPTER FIVE**

#### **CONCLUSION**

The E-Dairy AI Prediction System marks a major breakthrough in dairy farm management, offering a cutting-edge solution to boost productivity and improve decision-making. Utilizing advanced AI algorithms and integrating real-time data from IoT sensors, the system delivers precise predictions and valuable insights tailored to the needs of dairy farmers, veterinarians, and feed suppliers. Its intuitive interface and robust infrastructure facilitate smooth interactions and operational efficiency, enhancing overall management practices in the dairy industry.

The effective implementation and deployment of the E-Dairy AI Prediction System represent a significant modernization of dairy operations. Through thorough testing and user feedback, the system has been optimized to address the specific needs of its users, ensuring accuracy and dependability. This comprehensive solution not only fills existing gaps in dairy management but also establishes a new benchmark for future technological advancements, promising substantial improvements in productivity and sustainability for the dairy industry.

### **CHAPTER SIX**

#### RECOMMENDATION

To fully realize the advantages of the E-Dairy AI Prediction System, it is recommended to implement the system in phases. Start with a pilot phase involving a small group of users to collect feedback and resolve any issues before proceeding with a broader rollout. This approach will facilitate smoother system adoption and refinement. Additionally, providing continuous training and support is crucial to help users effectively navigate and utilize the system's features.

It is also important to regularly review and update the system in response to user feedback and technological progress. Keeping the system up-to-date will ensure it remains effective and relevant to the needs of the dairy industry. Investing in strong cybersecurity measures is essential to safeguard data and maintain system integrity. Following these recommendations will help the E-Dairy AI Prediction System deliver long-term value and enhance dairy farm management practices.

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# APPENDIX 1

## Schedule

| Month/phase         | May | June | July |
|---------------------|-----|------|------|
| Requirement         |     |      |      |
| analysis and        |     |      |      |
| specification       |     |      |      |
| Research            |     |      |      |
| Methodology         |     |      |      |
| System Analysis and |     |      |      |
| Design,             |     |      |      |
| System              |     |      |      |
| Implementation,     |     |      |      |
| Testing and         |     |      |      |
| Deployment          |     |      |      |

# APPENDIX 2

# Expenditure

| ITEM                    | QUANTITY | PRICE (Ksh) |
|-------------------------|----------|-------------|
|                         |          |             |
| Laptop                  | 1        | 28,000      |
|                         |          |             |
| Mobile phone            | 1        | 15,000      |
|                         |          |             |
| Questionnaires printing | 20       | 100         |
|                         |          |             |
| Total                   |          | 43,100      |
|                         |          |             |

## **APPENDIX 3**

# Sample Questionnaire Used to collect Stakeholders Views

## **Section 1: General Information**

| 1.                          | Name (Optional):  |                         |
|-----------------------------|---|-------------------------|
| 2.                          | Role:   |                         |
|                             | 0   | Dairy Farmer            |
|                             | 0   | Veterinarian            |
|                             | 0   | Feed Supplier           |
|                             | 0   | System Administrator    |
|                             | 0   | Other (Please specify): |
| 3.                          | 3. Organization Name:   |                         |
| 4.                          | 4. How long have you been using the E-Dairy AI Prediction System? |                         |
|                             | 0   | Less than 1 month       |
|                             | 0   | 1-3 months              |
|                             | 0   | 3-6 months              |
|                             | 0   | More than 6 months      |
| Section 2: System Usability |   |                         |

- 5. How would you rate the overall user interface of the system?
  - Excellent
  - $\circ$  Good

| 0             | Average   |
|---------------|---|
| 0             | Poor  |
| 0             | Very difficult  |
| 6. Are th     | ne system's predictive insights and recommendations relevant to your needs? |
| 0             | Highly relevant   |
| 0             | Somewhat relevant   |
| 0             | Neutral   |
| 0             | Not very relevant   |
| 0             | Not relevant at all   |
| Section 3: Sy | estem Performance   |
| 7. How s      | satisfied are you with the system's accuracy in predicting dairy outputs?   |
| 0             | Very satisfied  |
| 0             | Satisfied   |
| 0             | Neutral   |
| 0             | Unsatisfied   |
| 0             | Very unsatisfied  |
| 8. How        | would you rate the system's response time and performance?                  |
| 0             | Excellent   |
| 0             | Good  |
| 0             | Average   |
| 0             | Poor  |

# **Section 4: Training and Support**

9. Was the training provided sufficient to use the system effectively?

| 0   | Yes   |  |
|---|---|--|
| 0   | No (Please specify what additional training would be helpful):              |  |
| Section 5: Overall Satisfaction and Suggestions                                   |   |  |
| 10. Overall, how satisfied are you with the E-Dairy AI Prediction System?         |   |  |
| 0   | Very satisfied  |  |
| 0   | Satisfied   |  |
| 0   | Neutral   |  |
| 0   | Unsatisfied   |  |
| 0   | Very unsatisfied  |  |
| 11. What i  | features do you find most useful in the system? (Please list):              |  |
| 12. What improvements would you suggest for the system? (Please provide details): |   |  |
| 13. Would   | you recommend the E-Dairy AI Prediction System to others in your field?     |  |
| 0   | Yes   |  |
| 0   | No  |  |
| 0   | Maybe   |  |
| Section 6: Additional Comments  |   |  |
| 17. Please  | provide any additional comments or feedback about the E-Dairy AI Prediction |  |
| System:   |   |  |