

**Project Design Phase-II**  
**Technology Stack (Architecture & Stack)**

|               |   |
|---------------|---|
| Date          | 20 November 2023                                      |
| Team ID       | 591569  |
| Project Name  | Machine Learning Approach for Predicting the Rainfall |
| Maximum Marks | 4 Marks   |

**Technical Architecture:**

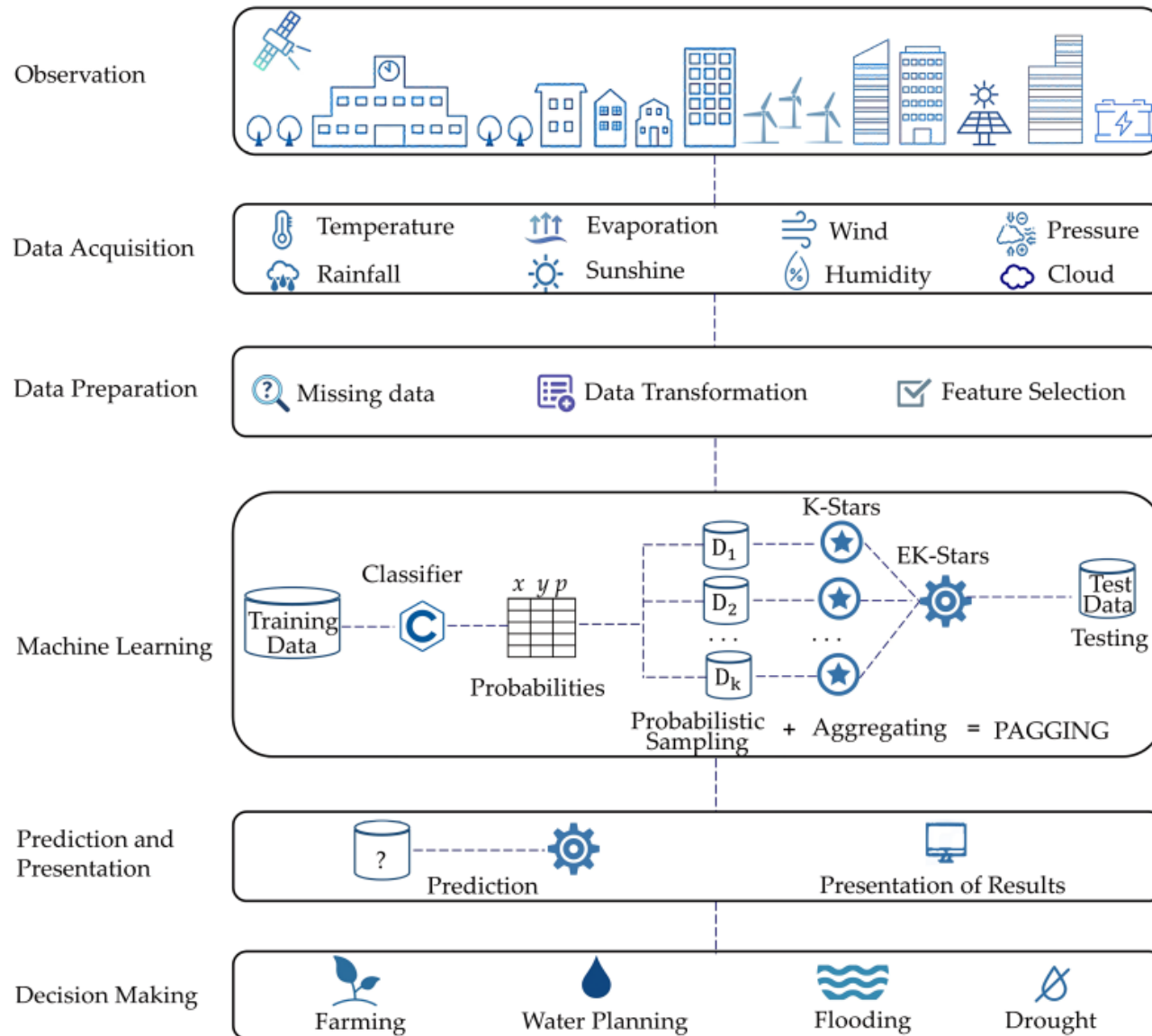
The Deliverable shall include the architectural diagram as below and the information as per the table1 & table 2

**Example: Rainfall Prediction using an Ensemble Machine Learning Model**

**Reference:** <https://www.mdpi.com/2071-1050/15/7/5889>

**Guidelines:**

1. Include all the processes (As an application logic / Technology Block)
2. Provide infrastructural demarcation (Local / Cloud)
3. Indicate external interfaces (third party API's etc.)
4. Indicate Data Storage components / services
5. Indicate interface to machine learning models (if applicable)



**Table-1: Components & Technologies:**

| S.No | Component                                  | Description  | Technology   |
|------|--|--|--|
| 1.   | Data Collection                            | Responsible for acquiring historical rainfall data from various sources.                             | Web scraping tools, APIs for meteorological data, data connectors.   |
| 2.   | Data Preprocessing and Feature Engineering | Cleans and processes the data, extracts relevant features, and engineers new ones.                   | Python libraries such as Pandas, NumPy for preprocessing, and feature engineering.                               |
| 3.   | Machine Learning Model Development         | Develops and trains machine learning models on historical data.                                      | Python-based machine learning libraries like scikit-learn, TensorFlow, or PyTorch.                               |
| 4.   | Real-time Prediction Module                | Handles real-time predictions using the trained machine learning model.                              | Streaming platforms (e.g., Apache Kafka), microservices architecture with containers (e.g., Docker, Kubernetes). |
| 5.   | Scalable Infrastructure                    | Provides a scalable and flexible computing environment for model training and real-time predictions. | Cloud platforms such as AWS, Azure, or Google Cloud.   |
| 6.   | Cloud Database                             | Database Service on Cloud  | IBM DB2, IBM Cloudant etc.   |
| 7.   | API Integration                            | Exposes APIs for external systems to query real-time and forecasted rainfall data.                   | RESTful APIs, JSON for data exchange, API Gateway for managing API access.                                       |
| 8.   | External API-1                             | Purpose of External API used in the application  | IBM Weather API, etc.  |
| 9.   | User Interface/Dashboard                   | Presents historical data and real-time predictions in an interactive and user-friendly format.       | Web-based dashboard using frontend frameworks like React or Angular.   |

|     |                        |  |   |
|-----|------------------------|--|---|
| 10. | Monitoring and Logging | Monitors system health, logs events, and performance metrics.      | Logging frameworks (e.g., ELK Stack - Elasticsearch, Logstash, Kibana), monitoring tools.                             |
| 11. | Security Measures      | Implements security measures to protect data and system integrity. | Encryption algorithms for data at rest (e.g., AES), HTTPS for secure communication, role-based access control (RBAC). |

**Table-2: Application Characteristics:**

| S.No | Characteristics                      | Description  | Technology   |
|------|--------------------------------------|--|--|
| 1.   | Real-time Prediction and Forecasting | The system provides real-time rainfall predictions and forecasts for future time periods, allowing users to make informed decisions based on the latest weather information. | Streaming platforms (e.g., Apache Kafka) for handling live data streams, machine learning models for forecasting.                                      |
| 2.   | Historical Data Visualization        | Users can explore and visualize historical rainfall patterns through an intuitive and interactive dashboard, gaining insights into past weather trends.                      | Web-based dashboard using frontend frameworks (e.g., React, Angular, Vue.js), backend APIs for data retrieval.   |
| 3.   | Scalability and Flexibility          | The system is designed to scale with increasing data volumes and user demands, providing flexibility to adapt to changing requirements.                                      | Cloud platforms (e.g., AWS, Azure) for scalable and flexible computing resources, containerization (e.g., Docker, Kubernetes) for resource management. |
| 4.   | User-Friendly Interface              | The application features an intuitive and user-friendly interface, allowing users to   | Responsive web design, frontend frameworks (e.g., React, Angular,  |

|    |                   |   |   |
|----|-------------------|---|---|
|    |                   | easily navigate and interpret rainfall predictions and historical data.   | Vue.js), interactive data visualization libraries.  |
| 5. | Security Measures | The application prioritizes data security, implementing encryption for data at rest and in transit, secure APIs, and access controls. | Encryption algorithms (e.g., AES), HTTPS for secure communication, API security mechanisms, role-based access control (RBAC). |

#### References:

<https://www.mdpi.com/2071-1050/15/7/5889>

<https://journalofbigdata.springeropen.com/articles/10.1186/s40537-021-00545-4>

<https://www.ncbi.nlm.nih.gov/pmc/articles/PMC9099780/>