

Enhancing AI based evaluation for smart cultivation and crop testing using agro-datasets

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How to cite this paper: Shibin David, R.S. Anand, Martin Sagayam (2020). Enhancing AI based evaluation for smart cultivation and crop testing using agro-datasets. Journal of Artificial Intelligence and Systems, 2, 149–167.
<https://doi.org/10.33969/AIS.2020.21010>.

Received: March 26, 2020

Accepted: April 20, 2020

Published: April 21, 2020

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Abstract

In India, Agriculture is considered as the backbone of the country where 70 percent of the economy relies upon it. It requires the involvement of many natural resources including land, water, and energy. An astonishing factor is that 60 percent of the water diverted or pumped for irrigation is wasted via runoff into waterways or evapotranspiration. Although chemical fertilizers improve the growth of plants and increase the yields of fruits and vegetables in a relatively short period of time, there are certain shortcomings of using chemical fertilizers as an opponent to the use of organic fertilizers obtained from natural wealth. The persistent use of chemical fertilizers causes the pollution of ground water sources, or leaching. Since the chemicals present in the chemical fertilizers spoil the soil scraps, the inference of this will be a high impact on the soil with reduced drainage and in the air circulation. The synthesizers that are used for farming will adversely affect the nature and the pH of the soil. Nowadays water scarcity is considered to be major concern in the cultivation process. Alongside, yet another major problem faced in cultivation is the usage of lots of fertilizers which makes the land infertile.. In this paper, the nature of the work to sense various factors such as atmospheric temperature, soil moisture, rain, and pH value through a GSM module. Using the aforementioned factors, the farmers and landlords can able to predict how much water is required for the land and how much nutrients required for land. The technological advancement in artificial intelligence paves a way to detect the sensed values and predict whether a crop could be planted on the soil present in a region.

Keywords

Arduino UNO, Soil moisture sensor, pH sensor, Temperature sensor, Rain drop sensor, GSM module, Agriculture land

1. Introduction

Agriculture is one of the most important areas of human activity. Agriculture needs the perseverance of various natural resources, including land, water, and energy [1]. Agriculture is the biggest water user, with irrigation accounting for 70% of global water withdrawals. As the report from the UN food and water organization says, an astounding 60% of the water side-tracked or impelled for irrigation is wasted through runoff into waterways or evapotranspiration [2,3]. Although chemical fertilizers improve the growth of plants and increase the yields of fruits and vegetables in a relatively short period of time, there are certain disadvantages of using chemical fertilizers as critical organic fertilizers derived from natural sources [1,2].

- The persistent use of chemical fertilizers causes the pollution of ground water sources, or leaching. As the chemicals within the chemical fertilizers destroy soil crumbs; the results are extremely compacted soil with reduced evacuation and air circulation.
- The artificial chemicals within the chemical fertilizers adversely have an effect on the health of naturally found soil micro-organisms by poignant the soil pH scale [1].
- The scope of this paper is to reduce water scarcity, to reduce the use of inorganic fertilizers, to sense various factors of the soil such as temperature, moisture, rain, pH content and to predict the amount of water and nutrients to be required for the soil [4].
- Water scarcity and infertile lands are the primary concern in this work where smart cultivation will prove to assist the farmers despite of the use of existing preventive measures.

2. Related works

2.1. Existing system

Traditional way of farming is otherwise termed to be the primogenital food production system and the origin of agriculture where it is being practiced for too many years. It also acts as the key source for refining the farming in terms of conventional, modern and organic farming which promotes genetic diversity also. Increasing environmental awareness farmers [2-3,5] following modern and scientific understanding of ecology, soil science, irrigation, using traditional method like crop rotations, compost / dung fertilizer, natural weed, biological pest control to protect the ecosystem. One major problem of traditional farming is farmers have to be present in the field all the time [5-7]. Another major problem is over pumping of

water to the field results in wastage of water [3]. The existing system contains only two or three sensors in the sensing part like soil moisture sensor and temperature sensor, humidity sensor and temperature sensor, rain drop sensor, humidity sensor and soil moisture sensor and so on.

2.2. Motivation

The motivation behind this work is by sensing various factors such as atmospheric temperature, soil moisture, rain, pH value. Using these factors we can able to predict how much water is required for the land and how much nutrients required for land. So this can be narrowed down to a specific crop.

3. Experimental setup

The system is developed by two different ways for smart cultivation system a) System Module b) Implementation.

3.1. System Module

The system module is explained in this section. Fig.1. shows the flow of the architecture for smart cultivation system. Here we are having various sensors such as temperature, soil moisture, raindrop, pH and GSM. This hardware can be deployed in any type of crop to sense the value of the soil.

A Sequence diagram is an interaction diagram that shows how processes operate with one another and in what order. It is a construct of a Message Sequence Chart. A sequence diagram shows object interactions arranged in time sequence. It depicts the objects and classes involved in the scenario and the sequence of messages exchanged between the objects needed to carry out the functionality of the scenario. Sequence diagrams are typically associated with use case realizations in the Logical View of the system under development. Sequence diagrams are sometimes called event diagrams or event scenarios.

LM35 temperature sensor as shown in Fig 2 is used to measure hotness or coldness of an object. The output of LM35 temperature sensor is in °C. The temperature sensor operates in the range from -55°C to 150°C [8].

The factors such as temperature value, pH value and moisture value measures the strength of the crop to be cultivated in a region. The nominal range of these factors is tabulated in Table.1. Based on the constant metrics, the soil moisture, rain intensity and pH value is tested using the arduino program and the same is communicated to the agriculturists.

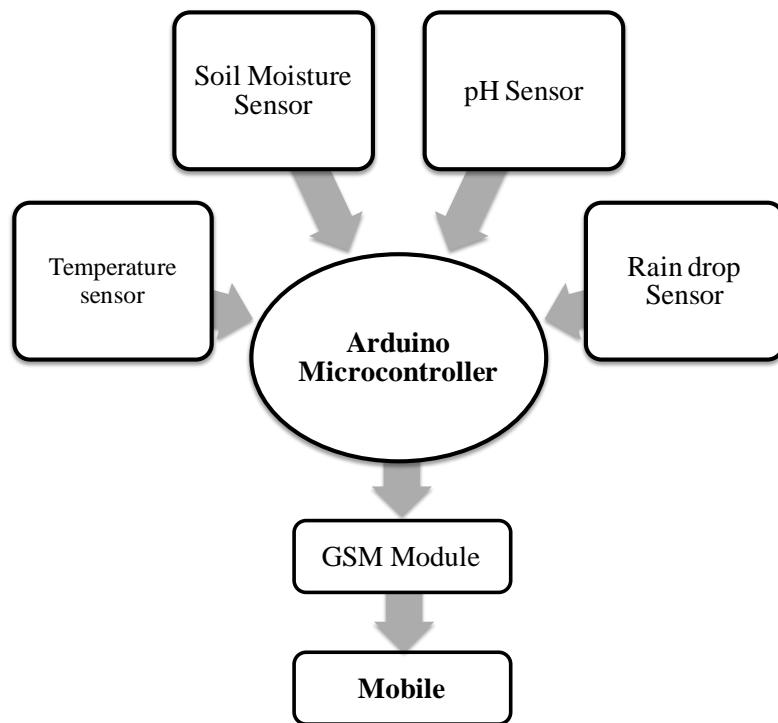


Figure 1. Flow diagram for smart cultivation



Figure 2. LM35 temperature sensor

Soil moisture sensor is used to estimate the soil volumetric content of the soil. It monitors the soil content to control the irrigation system. The soil moisture sensor

is depicted in Fig 3. Rain drop sensor as shown in Fig 4 is a rain sensing module which is used to measure the rain fall intensity [9,11]. It is used to calculate the amount of rain over a particular period.

pH sensor is used to sense the pH value of the soil. If the pH value is less than 7, then it is acidic. If it is equal to 7, then it is neutral. If it is greater than 7, then it is basic or alkaline. It is displayed in Fig 5.



Figure 3. Soil Moisture sensor



Figure 4. Rain drop sensor

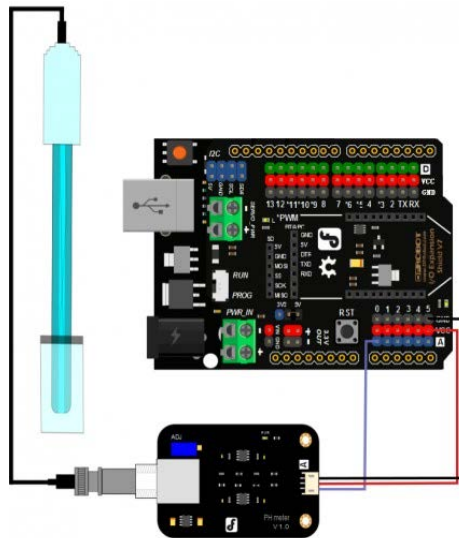


Figure 5. pH sensor

3.2. Implementation

The values of temperature sensor, soil moisture sensor, pH sensor and rain drop sensor are given as input to Arduino [4]. The Output of Arduino is given as input to GSM module. The output of GSM module is sent as notification to mobile user as SMS.

The values from the sensors are sensed and stored in Arduino and corresponding action is done [11]. For temperature sensor the temperature value ranges from 16°C to 27°C [10,13-14]. If the value is less than or greater than the required temperature it intimates the user via SMS else it display data to the user via SMS.

For soil moisture sensor the moisture value for dry, medium, wet soil is 14%, 18% and 20% respectively. If the value is less than or greater than the required moisture content it intimates the user via SMS else it display data to the user via SMS. For pH sensor the pH value ranges from 6.5 to 8. If the value is less than or greater than the required pH value it intimates the user via SMS else it display data to the user via SMS.

The values of the sensor are sensed and given to Arduino [18]. The Arduino detects the data and intimate the user via GSM. GSM is global system for mobile communication that is used for transmitting data and voice services [4,10]. The mobile user gets the information as SMS via GSM. Sensing of various factors can be narrowed down to a specific crop.

3.3. Steps to implement the module

There are certain functional requirements which may sense and provide the statistical analysis report for implementing the smart cultivation system into the agricultural lands.

Procedure for arduino setup:

- Install arduino 1.6.10
- Open the sketch book
- Connect the sensors, arduino and GSM (Fig.6)
- Type the code in the sketch book
- Verify the execution of the code (Fig.7, Fig.8, Fig.9, Fig.10, Fig.11 depicts it)
- Upload the code
- If the channel connectivity between the user's mobile and sensors are fine, then the output message will reach the user

A minimal arduino C/C++ sketch which is used for implementation, as seen by the Arduino IDE programmer, consists of two functions including setup and loop functions [12] [17].

- Setup: This function is called once when a sketch starts after power-up or reset. It is used to initialize variables, input and output pin modes, and other libraries needed in the sketch.
- Loop: After setup has been called, function loop is executed repeatedly in the main program. It controls the board until the board is powered off or is reset. Most Arduino boards contain a light-emitting diode (LED) and a load resistor connected between pin 13 and ground, which is a convenient feature for many tests and program functions.

The same connectivity is made along with the soil moisture sensor, raindrop sensor, temperature sensor, pH sensors [16].

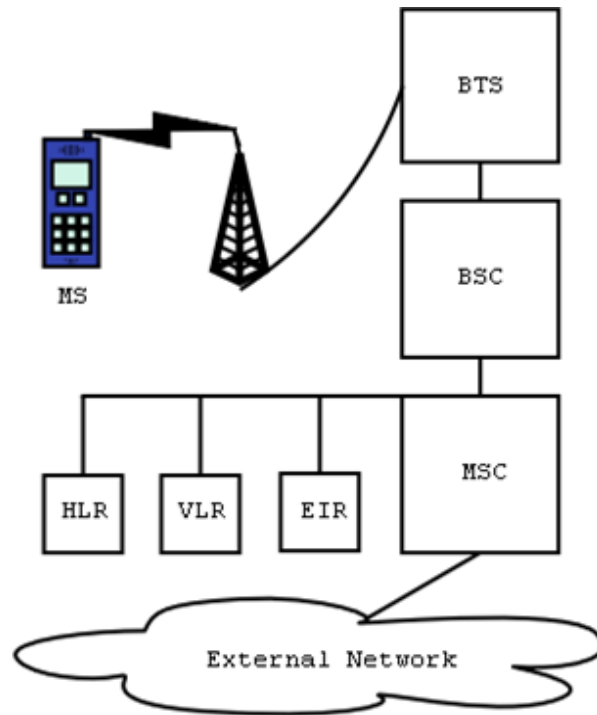


Figure 6. GSM Module

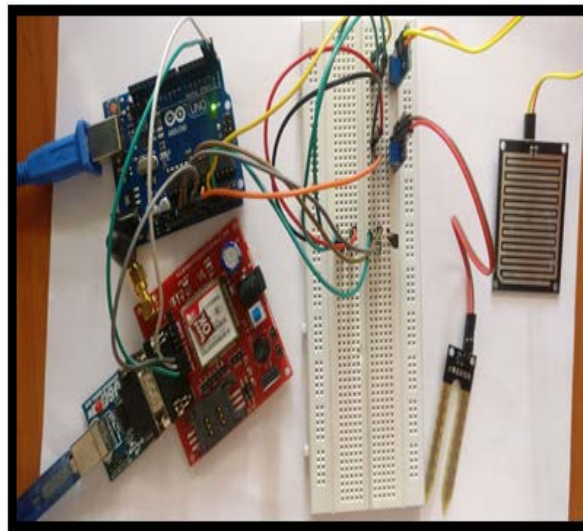


Figure 7. Connectivity between the sensors, arduino and GSM



Figure 8. Testing using soil moisture sensor

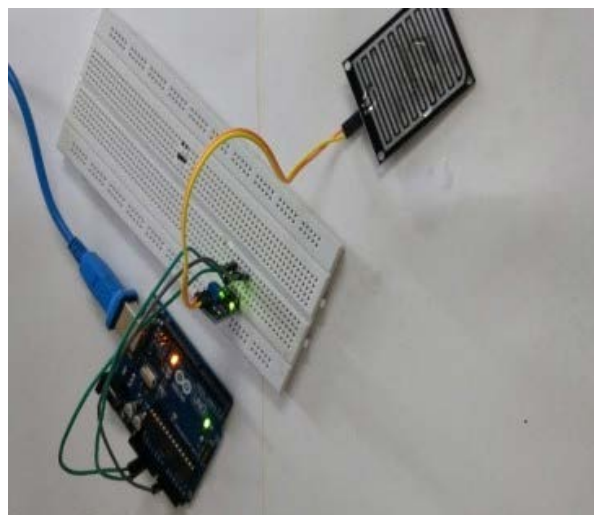


Figure 9. Raindrop sensor

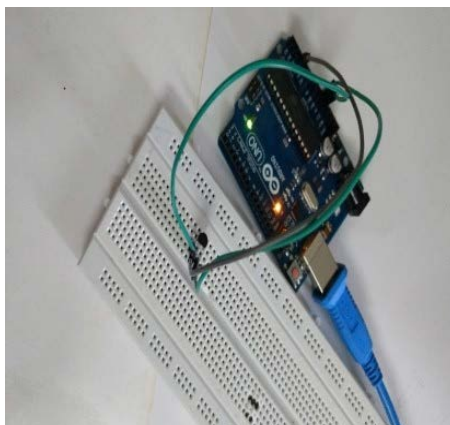


Figure 10. Temperature sensor

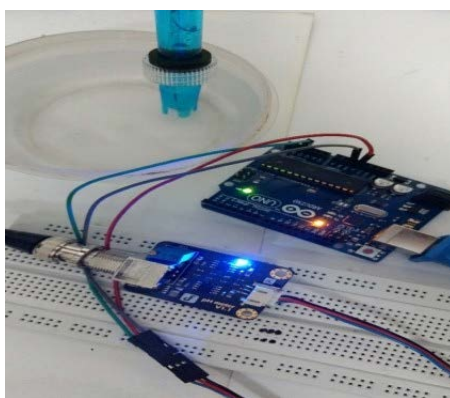


Figure 11. pH sensor

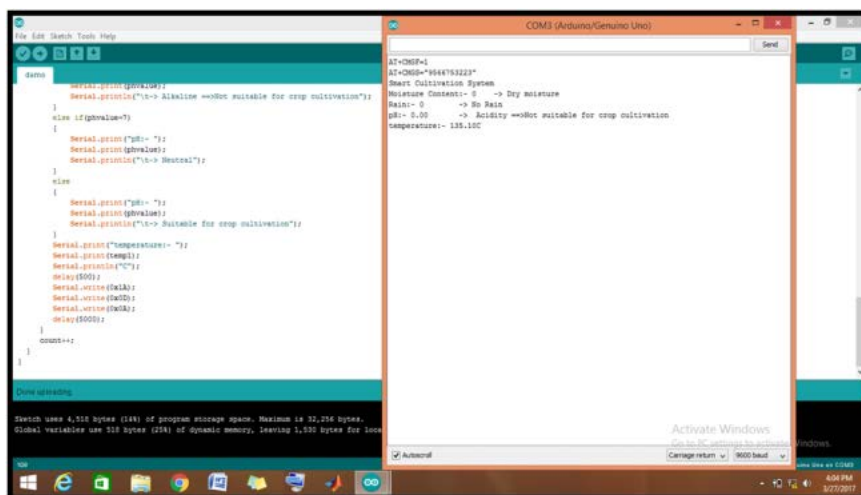


Figure 12. Outcome of the smart cultivation



Figure 13. Sensed output from smart cultivation

The outcome including the moisture, pH, temperature and acidity will be sent to the user as depicted in the Fig 12

4. Results and Discussions

4.1. Experimental analysis

There are certain functional requirements which may sense and provide the statistical analysis based on the range of factors prevailing for the crops presented in Table. 1.

This statistics will help the agriculturists to sense the pH value, determine the metrics to know the strength of the soil, create evidence against the strength using Bayes algorithm and provide the results.

The functional requirements include

- Sensing the pH value
- Sensing the atmospheric temperature
- Sending SMS via GSM

- Sensing the moisture content

Table 1. Range of factors prevailing the crops.

Crop	Temperature value	pH value	Moisture value
Paddy	16-27°C	6.5-8.5	Dry-14% Medium-18% High-20%
Peppers	14-16°C (night) 22-30°C (day)	5.5-6.5	60-65%
Egg plant	15-18°C (night) 22-26°C (day)	5.5-7.0	60-70%
Cucumber	22-28°C (day) 18-20°C (night)	5.5 – 6.5	70-90%
Potato	7-27°C	5.0 – 5.5	63–83%

The values of temperature sensor, soil moisture sensor, pH sensor and rain drop sensor are given as input to Arduino [19]. The Output of Arduino is given as input to GSM module. The output of GSM module is sent as notification to mobile user as SMS. The values from the sensors are sensed and stored in Arduino and corresponding action is done. For temperature sensor the temperature value ranges from 16°C to 27°C. If the value is less than or greater than the required temperature it intimates the user via SMS else it display data to the user via SMS.

For soil moisture sensor the moisture value for dry, medium, wet soil is 14%, 18% and 20% respectively. If the value is less than or greater than the required moisture content it intimates the user via SMS else it display data to the user via SMS.

For pH sensor the pH value ranges from 6.5 to 8, If the value is less than or greater than the required pH value it intimates the user via SMS else it display data to the user via SMS. The values of the sensor are sensed and given to Arduino. The Arduino detects the data and intimate the user via GSM. GSM is global system for mobile communication that is used for transmitting data and voice services. The mobile user gets the information as SMS via GSM. Sensing of various factors can be narrowed down to a specific crop. Our project deals with paddy.

Table 2. Test case using soil sensors.

Test case Name	Test case procedure	Expected Result	Obtained result	Status Pass/fail
	Detect the heavy moisture of the soil	Moisture > 80	Moisture = 81	Pass
Sense soil moisture	Detect the medium moisture of the soil	Moisture > 25 <80	Moisture = 58.90	Pass
	Detect the dry moisture of the soil	Moisture < 25	Moisture = 23	Pass

Table 3. Test case using rain sensor.

Test case Name	Test case procedure	Expected Result	Obtained result	Status Pass/fail
	Detect heavy rain	Rain intensity > 300	Rain intensity = 1020	Pass
Sense rain intensity	Detect moderate rain	Rain intensity > 300 < 500	Rain intensity = 363	Pass
	Detect no rain	Rain intensity > 500	Rain intensity = 237	Pass

Table 2 deals with the test case of deploying soil moisture sensor, Table 3 indicates the use of rain sensors, Table 4 denotes that temperature sensor could be used to test the temperature and Table 5 indicates the deployment of pH sensors.

Table 2 depicts the status of sensing the soil moisture upon heavy ratio, moderate ratio and dry ratio and infers whether cultivation can be made on the soil or not. The moisture level obtained when heavy ratio of moisture is found to be 81 whereas the moderate ratio and dry ratio is found to be 58.90 and 23 respectively. Hence, it is

concluded that the smart cultivation can be performed on all the levels of moisture available in the soil.

Table 4. Atmospheric temperature testing.

Test Name	case	Test case procedure	Expected Result	Obtained result	Status Pass/fail
Sense atmospheric temperature		Detect temperature	Temperature >15 <20	Temperature = 19.35	Pass

Table 3 shows the sensing of rain intensity upon various climatic conditions. The soil sensor is used to sense the rain intensity and it outlays 1020, 363, 237 for heavy intensity, moderate intensity and low intensity respectively.

The atmospheric temperature range on an optimum should lie between 15°C and 20°C. As tabulated in Table 4, while the atmospheric temperature was sensed using the sensor, it was detected and found that it lies in the range of 19.35° C.

The soil nature can be classified into three categories which includes acidic, neutral and alkaline. When the test case is fixed for acidic nature of the soil, the sensor senses the soil and produces the pH result as 4.09 which are favorable over the expected value. The test is done using the soil sensor which is kept and tested at a pH range less than 5.0. This indicates that the soil can used for smart cultivation in terms of acidic nature of the soil. When the test case is fixed for neutral nature of the soil, the sensor senses the soil and produces the pH result as 7 which are favorable over the expected value. The test is done using the soil sensor which is kept and tested at a pH range equal to 7.0. This indicates that the soil can used for smart cultivation in terms of neutral nature of the soil. When the test case is fixed for alkaline nature of the soil, the sensor senses the soil and produces the pH result as 7.5 which are favorable over the expected value. The test is done using the soil sensor which is kept and tested at a pH range greater than 7.0. This indicates that the soil can used for smart cultivation in terms of alkaline nature of the soil.

Table 5. pH value testing.

Test Name	case	Test procedure	case	Expected Result	Obtained result	Status Pass/fail
		Detect nature of the soil	acidic	pH <5	pH = 4.09	Pass
Sense pH value		Detect nature of the soil	neutral	pH =7	pH = 7	Pass
		Detect nature of the soil	alkaline	pH >7	pH = 7.5	Pass

4.2. Theoretical analysis and discussions

Let 'A' be the chance of deploying the paddy crop in the fields, 'B' be the chance of growth in the crop deployed in the field.

Assume $\neg A \rightarrow \neg B$

This implies, $B \rightarrow A$ sensor,

The chance of growth in the field is possible.

Table 6. Paddy data set.

PADDY	Temperature (°C)	pH	Moisture (%)
	3	2.3	29
	11	4.2	25
	19	7.3	20
	29	9.2	13
	36	10.6	7

The dataset represented in Table.6 depicts the various factors to determine the strength of soil, to cultivate paddy crop in the local area based on the dataset randomized variable [20-24] containing the set of values which includes its temperature, pH value and moisture level. Here the set $S = \{19, 7.3, 20\}$ consisting the test variables to be incorporated into Naive Bayes algorithm [20, 24-26] in order to test the accuracy a possibility of the growth of paddy crop in our local area.

Let the chance of positive growth be calculated as $P(B|A) = A \rightarrow B$, and the prior probability of planting the crop in the field can be denoted as $\alpha(A)$.

The probability of growth in the suitable environment is denoted as $P(A|B) = 1$ which implies $B \rightarrow A$.

Based on the Naïve Bayes theorem,

$$P(A|B) = \frac{P(A).P(B/A)}{P(B)} \tag{1}$$

Similarly,

$$P(A|B) = \frac{P(B/A). \alpha(A)}{P(B/A). \alpha(A) + P(B|\neg A)P(B).\alpha(\neg A)} \tag{2}$$

$$P(A) = \sum_{i=1}^{i=n} \text{Decisions on } \{P_1, P_2, P_3 \dots P_n\} \tag{3}$$

$$P(B) = \sum_{j=1}^{j=n} \text{Growth of } \{P_1, P_2, P_3 \dots P_n\} \tag{4}$$

$$P(B|A) = \frac{\text{Decisions upon evidences}}{\text{Growth}} \tag{5}$$

$$P(B|A) = \sum_{i=1}^{i=n} \frac{D \{P_1, P_2, P_3 \dots P_n\}}{G \{P_1, P_2, P_3 \dots P_n\}} \tag{6}$$

$$P(A|B) = \frac{P(A). P(B|A)}{P(B)} \tag{7}$$

$$P(A|B) = \frac{P(B|A). \alpha(A)}{P(B|A). \alpha(A) + P(B|\neg A).\alpha(\neg A)} \tag{8}$$

Propositional logic,

$$(\neg A \rightarrow \neg B) \rightarrow (B \rightarrow A)$$

$$P(B|A) = (A \rightarrow B)$$

$\alpha(A)$ = Prior Probability

$$P(A|B) = 1 = B \rightarrow A$$

$$P(A|B) = \frac{P(B/A). \text{Prior probability}}{P(B/A). \text{Prior probability} + P(B/\text{Propositional logic}).\alpha(\text{Propositional logic})} \tag{9}$$

Let the propositional logic applied over A be $\alpha(\neg A)$

$$P(A|B) = \frac{P(B|A). \alpha(A)}{P(B|A). \alpha(A) + P(B|\neg A).\alpha(\neg A)} \tag{10}$$

Evaluating the variables present in Table.6 based on Bayes theorem,

$$P(A) = 0.526, P(B) = 0.2618, P(B/A) = 0.326$$

$$P(A/B) = 0.655 = 65.5\%$$

The decision indicates that there is a possibility of 65.5% growth of paddy crop in our locality is possible and it is determined that the factor supporting such growth is available in our locality.

5. Conclusions

Factors such as temperature, soil moisture, pH value and rain are therefore sensed to predict how much water is required and the fertility of soil for a particular crop. Thus this way makes the farmers to do cultivation in a smarter way. As of this work, the KR representation and Bayes algorithm has been deployed to check whether a particular crop can be grown on the soil present in a region. This work can be further improved by using wireless module technology with addition of NPK (Nitrogen, Phosphorus and Potassium) sensor which decides how much extra nutrient contents are to be added in the soil to increase crop fertility and also by using automatic water supply combining with IOT concept.

Acknowledgements

The study on the cultivation of paddy and its practices has been made from the web portal TNAU.

Conflicts of Interest

There is no conflict of interest.

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