



An Advanced Flood Detection System Through Integrated Environmental Sensors and Real-Time Data Analysis

EL-SADAT STEM SCHOOL – GRADE 12 – FIRST SEMESTER – 2024/2025 – GROUP NO.25309

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Abstract

The project focused on two Grand Challenges in Egypt: Improving science and technology for everyone and reducing the effects of climate change. Climate change has made these floods worse, threatening communities and infrastructure. A solution was chosen: a flood early warning system. The system uses sensors to measure water level, flow rate, Temperature, humidity, air pressure, and water speed. The goal was to detect floods early and warn people in time to take action. The system's design included certain needs. Sensors had to be accurate, strong, and able to work in tough environments. The system also needed to show real-time data and send alerts if a flood was likely. A prototype was built using a transparent tank to simulate a river or reservoir. Sensors were connected to a microcontroller (ESP32-WROOM-32U) to collect and process data. Tests were done by a stimulation for flood conditions. The system successfully measured changes in water levels and sent warnings when danger levels were reached. The results showed that the system worked well and met its goals. In conclusion, the project proved that an early warning system for floods can help reduce damage caused by floods. This system can save lives, protect buildings, and reduce economic losses. It is a simple yet effective solution that can be used in other flood-prone areas, helping communities prepare for the impacts of climate change.

Introduction

Egypt faces some challenges in mitigating the effects of climate change, especially with increased flood risks within vulnerable areas, as shown in Figure 1. These challenges require innovative strategies in disaster prevention, in relation to the major challenges related to resilient infrastructure. This challenge forms the core of the current Capstone project in the semester regarding leveraging technology to enhance Public safety. Previous solutions have been researched, studied and analyzed in terms of their relative strengths and weaknesses. Improved flood prediction and community safety through a self-sustaining sensor network with local risk mapping and replication was improved in Honduras. Due to its complexity and resource requirements, it is not scalable further. In Jakarta, meanwhile, an IoT-based system was implemented using Raspberry Pi functions as a microcomputer for central control, collecting data from a water level sensor to measure water height, an ultrasonic sensor for further monitoring, a DHT11 sensor to monitor temperature and humidity, and an alarm as an audible warning device. Although the solution was cost-effective and automated, its narrow deployment and limited scope reduced its effectiveness. The solution will be selected by the team to work on the above-mentioned limit based on scalability and real-time processing. The proposed design incorporates IoT, advanced data analytics and enhanced communication channels to provide the required reliable and adaptive flood warning system. Design requirements: multi-sensor integration, signal processing, real-time monitoring, data visualization, energy efficiency and data storage. A prototype was designed integrating DHT11 sensors for both temperature and humidity, water level sensors for monitoring water level, BMP 280 atmospheric pressure, YF-S201C transparent water flow sensor required to calculate water velocity according to the flow rate law: $V=Q/A$ and ESP32 controllers to actuate and transmit such data. The chosen prototype solution is cost-effective and energy-efficient to address the weaknesses identified in the previous solutions. In this context, the team set out to develop an inexpensive flood detection system that ensures efficient data collection, cost-effective hardware, and predictive analytics to meet local and national needs. The design process consistently emphasizes user accessibility, community engagement, and integration with existing infrastructure.

Materials



The materials utilized in conducting the solution.

Methods & Test Plan

Connection

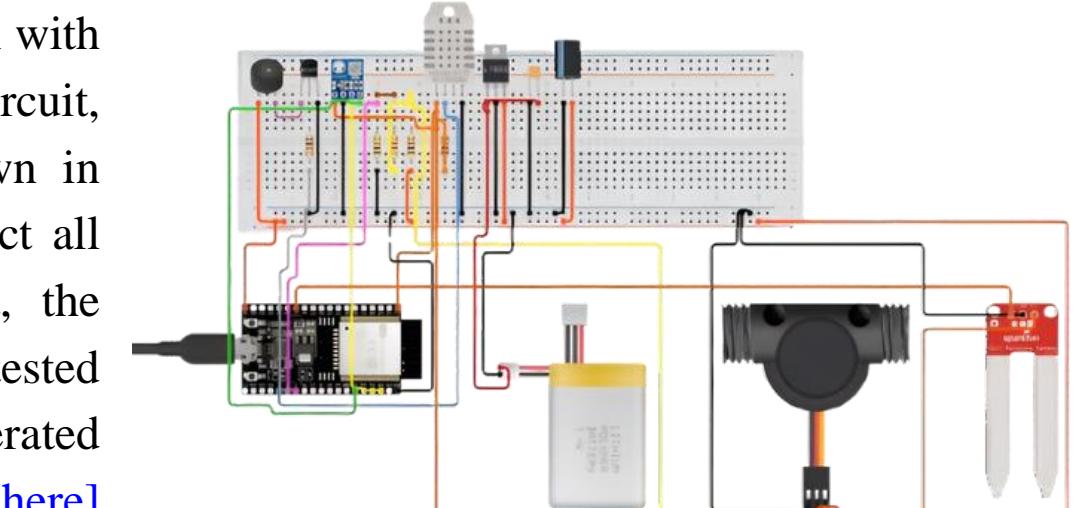


Fig. 2: illustrates the circuit design

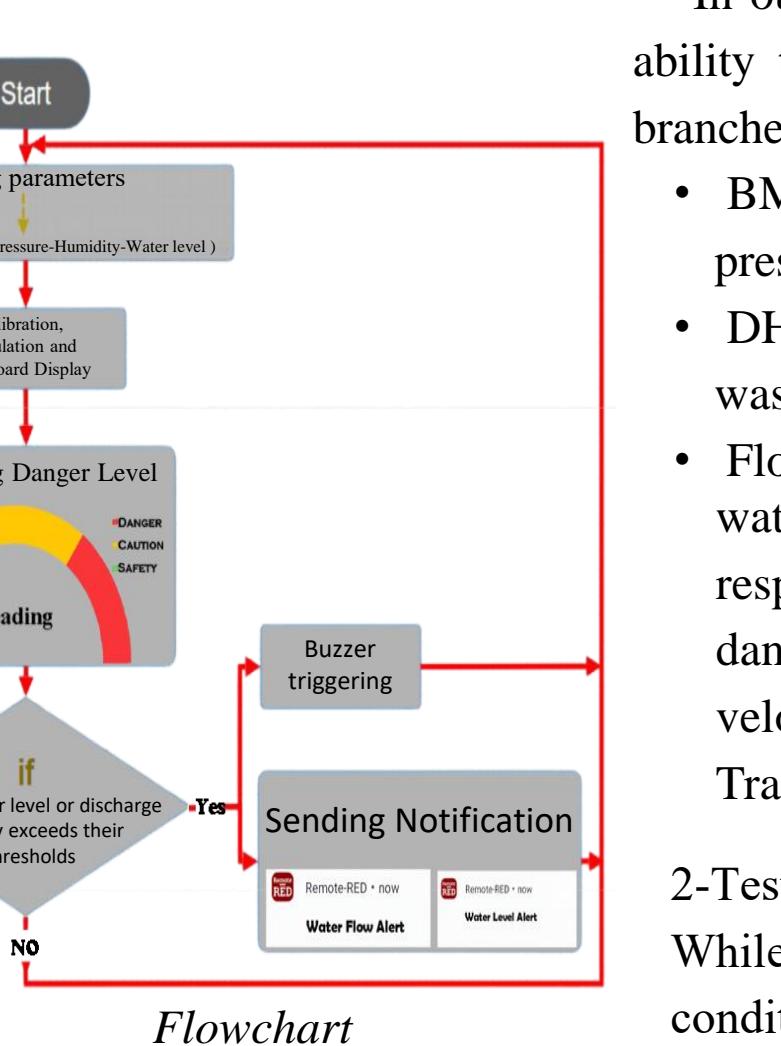
Key Words: Torricelli's Theorem - Multi-Sensor System - Regression line - Node-RED

Methods & Test Plan

Designing

After completing the connection stage, the next step is creating the final product with two primary objectives. The first is ensuring for an appropriate placement of the sensors, either on the designated box or other available components. Secondly, Adjust the codes of every sensor to work properly with the other sensors, and give correct readings in their new fixed position. Additionally, the ESP32 and the mobile phone were connected to the same Wi-Fi network for data transmission to the app's dashboard. This setup also enabled alert notifications and buzzer activation when needed. We used the Node-RED as a visual programming tool ideal for IoT and real-time data applications. by integrated process involved:

- Establishing MQTT communication to transmit real-time data.
- Creating flows to interpret the data and identify the flood risk based on the chosen parameters that we measure and give reading according to defined functions.
- Configuring notifications to alert the users of impending flood risks if the sensor readings reach dangerous conditions



Test Plan

In our project, the test plan mainly focuses on testing the functionality of each sensor and its ability to detect different factors. As shown in Figure 3&4. The test plan is divided into two branches: 1-Testing the ability of sensors to detect surrounding factors

- BMP280 Barometric Pressure: Tested by blowing on the sensor to create higher external pressure, then checking for increased pressure readings.
- DHT11: Exposed to hot water vapour to measure both temperature and humidity. The sensor was left in the vapour until it recorded high readings for both parameters.
- Flow Rate and Water Level Sensors: Tested by pouring water to reach the maximum height, observing the response of the water level sensors, and calculating the danger level percentage. The changes in flow rate and velocity were monitored using the Water Flow sensor Transparent YF-S201C.

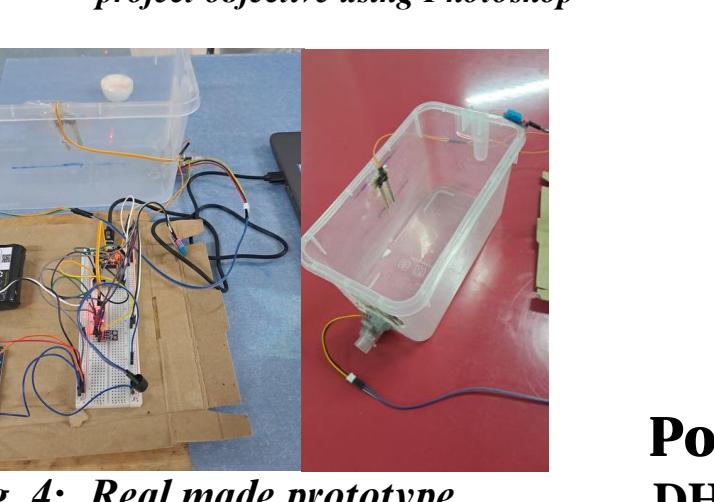
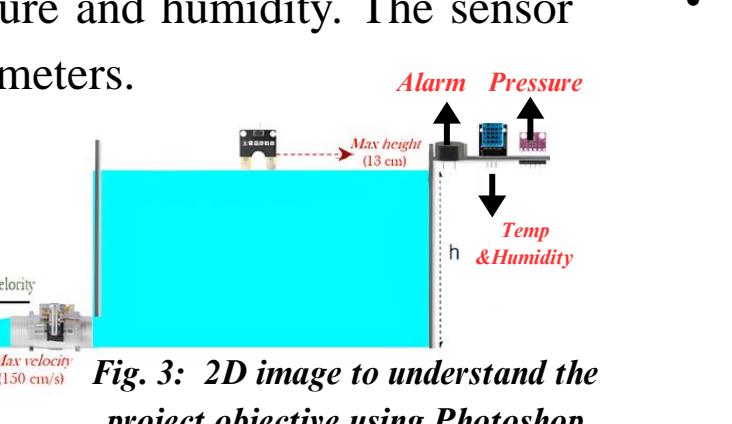


Fig. 4: Real made prototype

Results

Status

- | Reading | 1 st | 2 nd | 3 rd | 4 th | 5 th | 6 th | 7 th | 8 th | 9 th |
|-----------------|-----------------|-----------------|-----------------|-----------------|-----------------|-----------------|-----------------|-----------------|-----------------|
| Temperature (C) | 19.4 | 19.4 | 20.6 | 22.6 | 25.1 | 28.4 | 32.8 | 31.2 | 31 |
| Humidity(%) | 63 | 63 | 74 | 80 | 98 | 98 | 86 | 77 | 68 |
| Pressure (hPa) | 1015.71 | 1015.71 | 1016.54 | 1016.39 | 1016.26 | 1015.16 | 1016.09 | 1016.02 | 1015.7 |
| Velocity cm/s | 151 | 144 | 131 | 118 | 105 | 91 | --- | --- | --- |
- Table(3): Positive & Negative Results

Reading	1 st	2 nd	3 rd	4 th	5 th	6 th	7 th	8 th	9 th
Temperature (C)	19.4	19.4	20.6	22.6	25.1	28.4	32.8	31.2	31
Humidity(%)	63	63	74	80	98	98	86	77	68
Pressure (hPa)	1015.71	1015.71	1016.54	1016.39	1016.26	1015.16	1016.09	1016.02	1015.7
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Pressure (hPa)	1015.71	1015.71	1016.54	1016.39	1016.26	1015.16	1016.09	1016.02	1015.7
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Analysis

Where:

- X represents height and Y the velocity v^2
- X_i and Y_i are the individual data points.
- \bar{X} and \bar{Y} are the means of the X and Y variables, respectively.
- r is the Pearson correlation coefficient.

By substituting in the law the correlation ($r = 0.9961$)

In statistical concepts, this correlation is classified as very strong. The strength of the correlation, as illustrated in the in Figure(6), serves as a positive indicator of the accuracy of our calculations and the reliability of our data collection.

It is critically important to create a regression line to effectively visualize the trend or relationship between velocity squared and height by a single line. The equation of the regression line is calculated using the following formula: $v^2 = mH + c$

Where :

$$m = \frac{\sum(H_i - \bar{H})(V_i^2 - \bar{V}^2)}{\sum(H_i - \bar{H})^2}$$

$$m = 18.8976 \quad \& \quad c = -0.108$$

According to the original equation $v^2 = 2gh$,

the slope of the line $\frac{dv^2}{dh}$ should equal $(2g)$ which is constant .

By comparing the slope of our regression line to $(2g)$,

we can validate the accuracy and efficiency of our

calculations. In our case the slope represented by (m)then we

$$\text{can represent the accuracy by } \frac{m}{2g} = \frac{18.8976}{2 \times 9.8} = 96.35\%$$

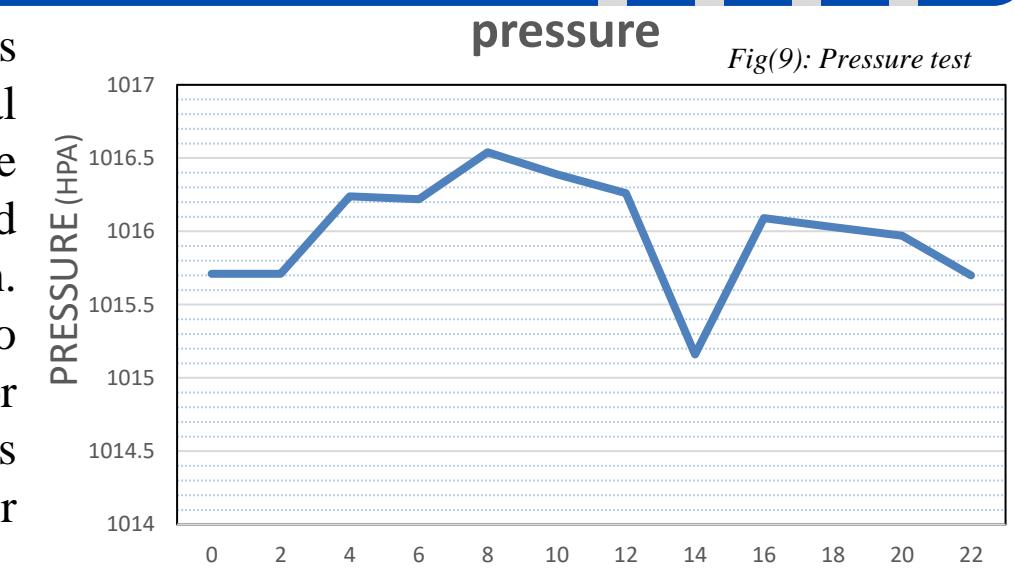
The high accuracy of our reading show that our data and calculation could be reliable and trusted

- The Water Flow Sensor contains an internal fan that rotates as water flows through it. This rotation generates pulses, which are translated into the water flow rate, typically measured in L/min, then we converted it to m^3/s . Then if we need to get the velocity of the water, we use the continuity equation ($Q = \text{Area}/\text{velocity}$). For testing, we filled the container with water and recorded the velocity over time. The system's threshold is set at a water velocity of 150 cm/s , which indicates a danger level.
- A soil moisture sensor can function as a water level detector by measuring changes in conductivity or resistance between two probes when they come into contact with water. The sensor is placed vertically at heights between 6 cm and 12 cm in the container. As the water rises, it alters the conductivity, which is expressed as a percentage. A range of 1% to 99% indicates a potential danger, and when it reaches 100% , the water level exceeds 12 cm . At this point, a notification is sent, and the buzzer is triggered.

Analysis

- The BMP280 Barometric Pressure Sensor is used to monitor external pressure. the normal pressure should stay above 1010 hPa because anything lower can encourage cloud formation and increase humidity, which often leads to heavy rain. To test the sensor, we tried blowing on it to simulate a sudden increase in pressure. The sensor responded perfectly, as shown in the graph. This test proves that it's accurate and reliable for detecting changes in pressure in real life.

Fig(9): Pressure test



- The DHT 11 is used to measure the temperature and humidity. The temperature should stay between 20°C and 30°C . If it drops below 20°C , condensation increases and this leads to heavy rainfall and potential flash floods, especially in poorly drained areas. On the other hand, when temperatures rise above 30°C , the atmosphere becomes unstable, which can also trigger heavy rain in terms of humidity, it should remain below 80% . When it exceeds this threshold, the air holds more moisture which increases the possibility of making heavy rain quickly

We boiled water in a glass and placed it beside the system. The sensor demonstrated its ability to adapt to environmental changes and the readings clearly indicated whether the conditions were normal or not, proving the sensors are reliable and effective.

Power and Memory Calculations

DHT11 : 2 values (temp and humidity) \times 2 bytes each = 4 bytes, Soil Moisture: 1 value \times 2 bytes = 2 bytes, Flow Rate: 1 value \times 4 bytes (float) = 4 bytes, BMP280 : 2 values (pressure and temperature) \times 4 bytes each = 8 bytes, Timestamp: 1 value \times 4 bytes = 4 bytes. Pre- record size = 22 bytes/record

Total records in one year: 2 records/minute \times 525,600 minutes/year = 1,051,200 records

Total storage = 22 bytes/record \times 1,051,200 records/year = 23,126,400 bytes/year \approx 22.05 MB

First, total sensors power = $\sum (I \times V) = 1.6846 \text{ Watt}$. Second, E battery = V battery \times C battery, to calculate it Runtime(hours) = E battery / P total, E battery = $4380 \text{ hours} \times 1.6846 \text{ Watt} = 7378.548 \text{ Wh}$, the battery desired must be $12 \text{V}, 614.879 \text{ ampere-hours}$.

Code	Learning Transfer
ST.3.02	We used to calculate the correlation between velocity squared, draw the regression line, and determine its equation.
PH.1.09	We studied the equation of continuity to calculate the water flow rate and applied Torricelli's Law to understand the relationship between the height of water and the velocity in the dam of discharged water
CH.3.01	applying scientific methodology by identifying flood risks, designing and testing a solution, collecting data, and analyzing results to confirm the system works.
PH.3.04 & PH.3.05	The communication methods learnt in these learning outcomes helped us in controlling communication between the components of the system.
ES.2.05	understanding river systems and drainage basins by showing how water interacts with landforms and the hydrologic cycle.

Conclusion & Recommendations

The project has showed its success as the sensors and modules work together with efficiency and the data processing by the algorithms of looping, if statements, and calculation function is done correctly to give a real-time output for the users. The microcontroller ESP32 makes communication with mobile via Wi-Fi to be able to get data everywhere. The DHT11 sensor measures the humidity and the BMP280 measures the atmospheric pressure, which are two different weather parameters. The soil moisture sensor has proved its ability to measure the water level in the tank, that classified as an important parameter we measure. The flow rate sensor measures the velocity of water in the water stream. Both flow rate sensor and water level sensor if they reach the dangerous conditions notify the user by the mobile application.