

Intelligent Water Consumption Control Using IoT and AI

Problem Statement

Water scarcity is becoming a critical global issue due to population growth, climate change, and inefficient consumption. Conventional water meters are passive systems that only measure usage for billing purposes and fail to provide real-time monitoring, anomaly detection, or active control against leaks and excessive consumption.

Objectives

- Monitor water consumption in real time using IoT sensors
- Detect abnormal water usage and potential leaks using AI
- Actively control water flow through automated intervention
- Provide users with real-time visualization and historical analytics
- Reduce water waste and promote sustainable consumption

Hardware Design

Main Components:

- ESP32 (sensing node with Wi-Fi)
 - YF-S201 Water Flow Sensor
 - Raspberry Pi 4 (Edge Gateway & Edge Computing Unit & Localized Server)
 - 12V Solenoid Valve
 - Relay Module with isolated power supply
- Design Rationale:
- ESP32 ensures accurate pulse counting and wireless data transmission
 - Raspberry Pi enables edge computing, local database, and TinyML inference
 - Relay isolation protects low-voltage logic from high-power actuation

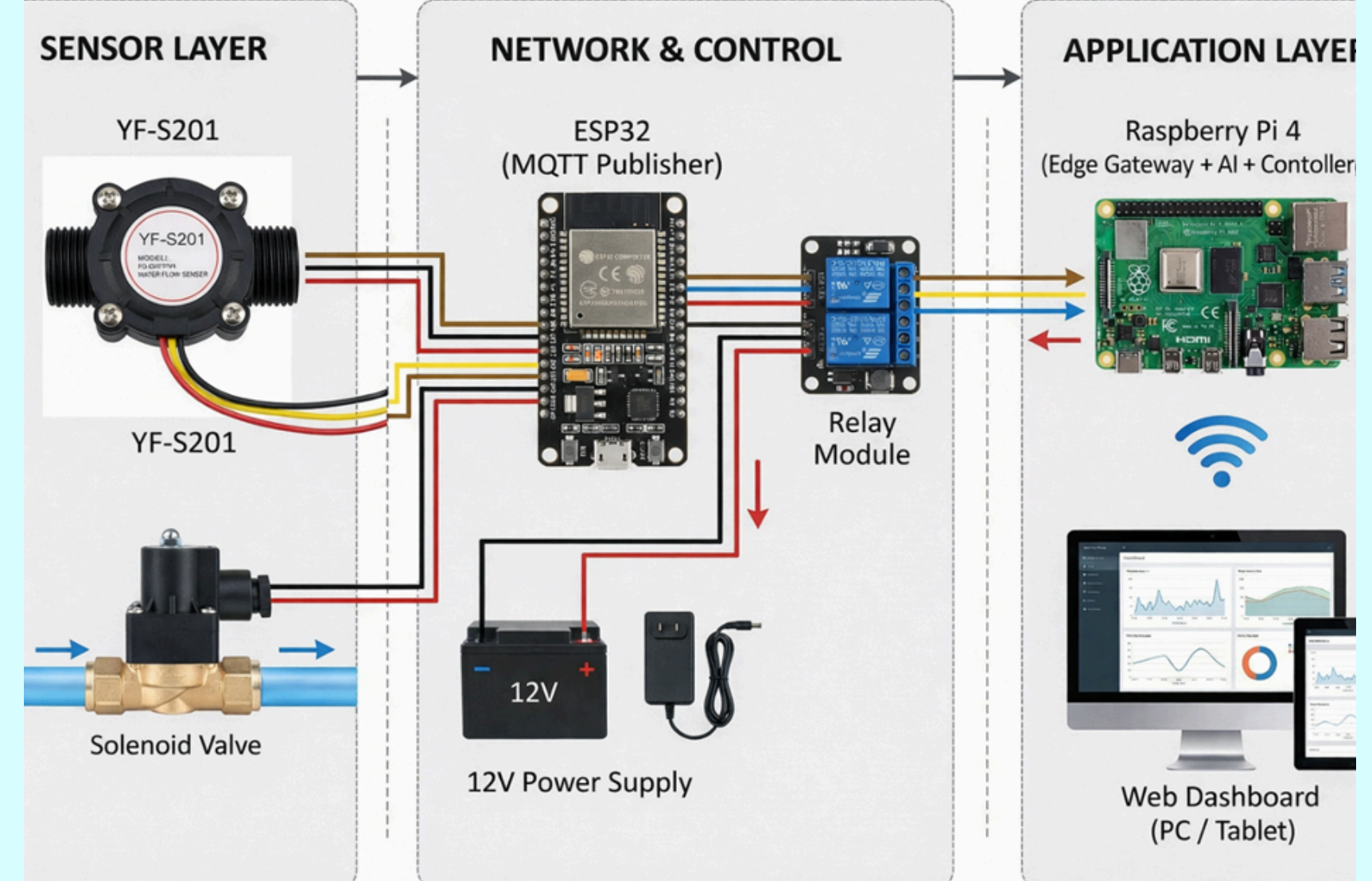
Proposed System

The system follows a three-layer IoT architecture

Perception Layer: ESP32 + YF-S201 water flow sensor collect real-time flow data

Network & Control Layer: Raspberry Pi 4 acts as an edge gateway, MQTT broker, and AI inference unit

Application Layer: Web dashboard for monitoring, analytics, and manual control



Software Design

Hybrid Communication Architecture

Real-Time Layer (MQTT): Usage: ESP32 → Edge Gateway

Application Layer (HTTP/REST): Usage: Edge Gateway → Backend & Frontend

Backend: Python-based MQTT subscriber and control logic, Java (Spring Boot framework)
Maven build tool
MySQL database

Frontend: Responsive Web Dashboard, Nginx web server (HTML/JavaScript), TypeScript

Dashboard Features:

Real-time water flow gauge (L/min)

Historical consumption charts, Manual emergency shut-off control

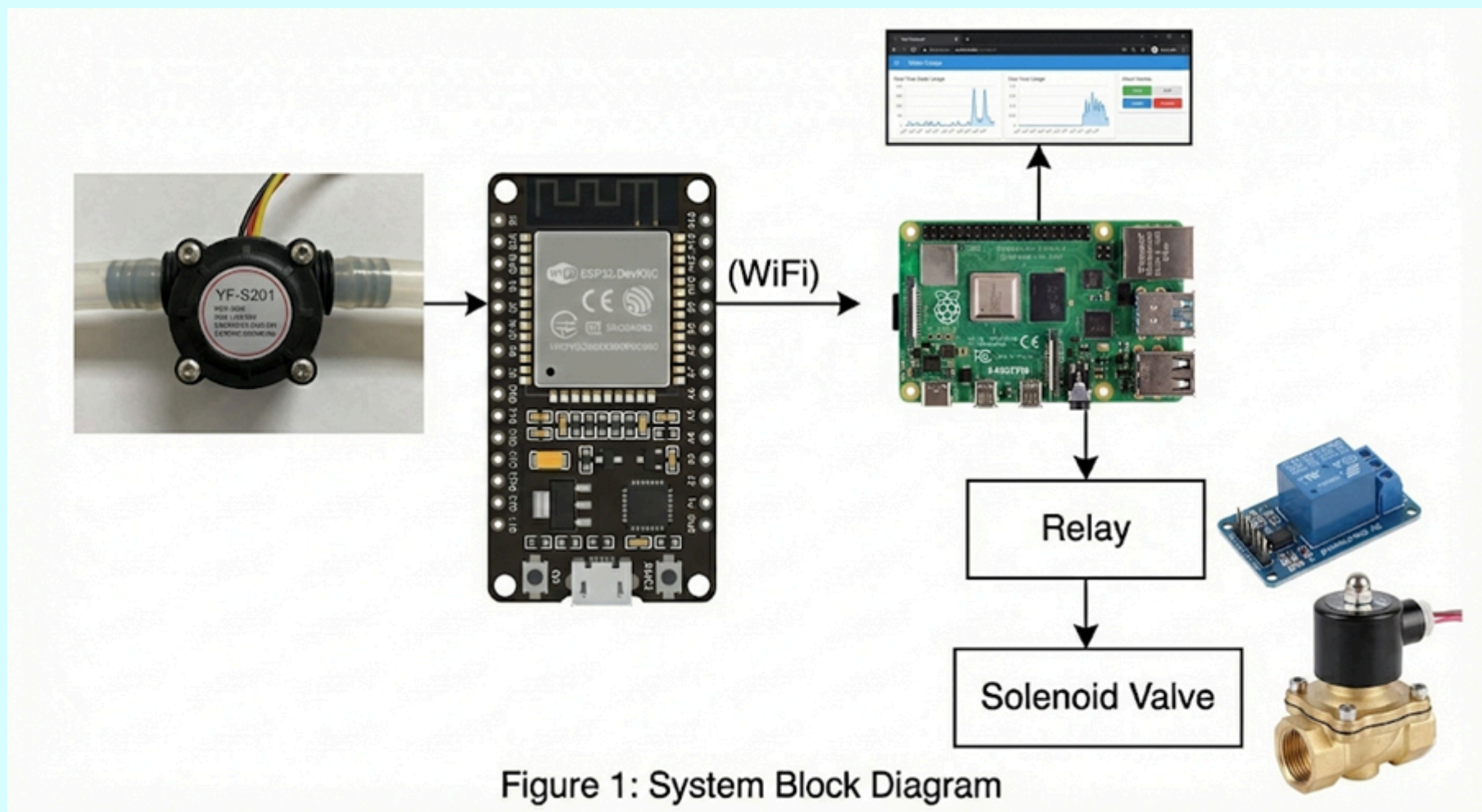
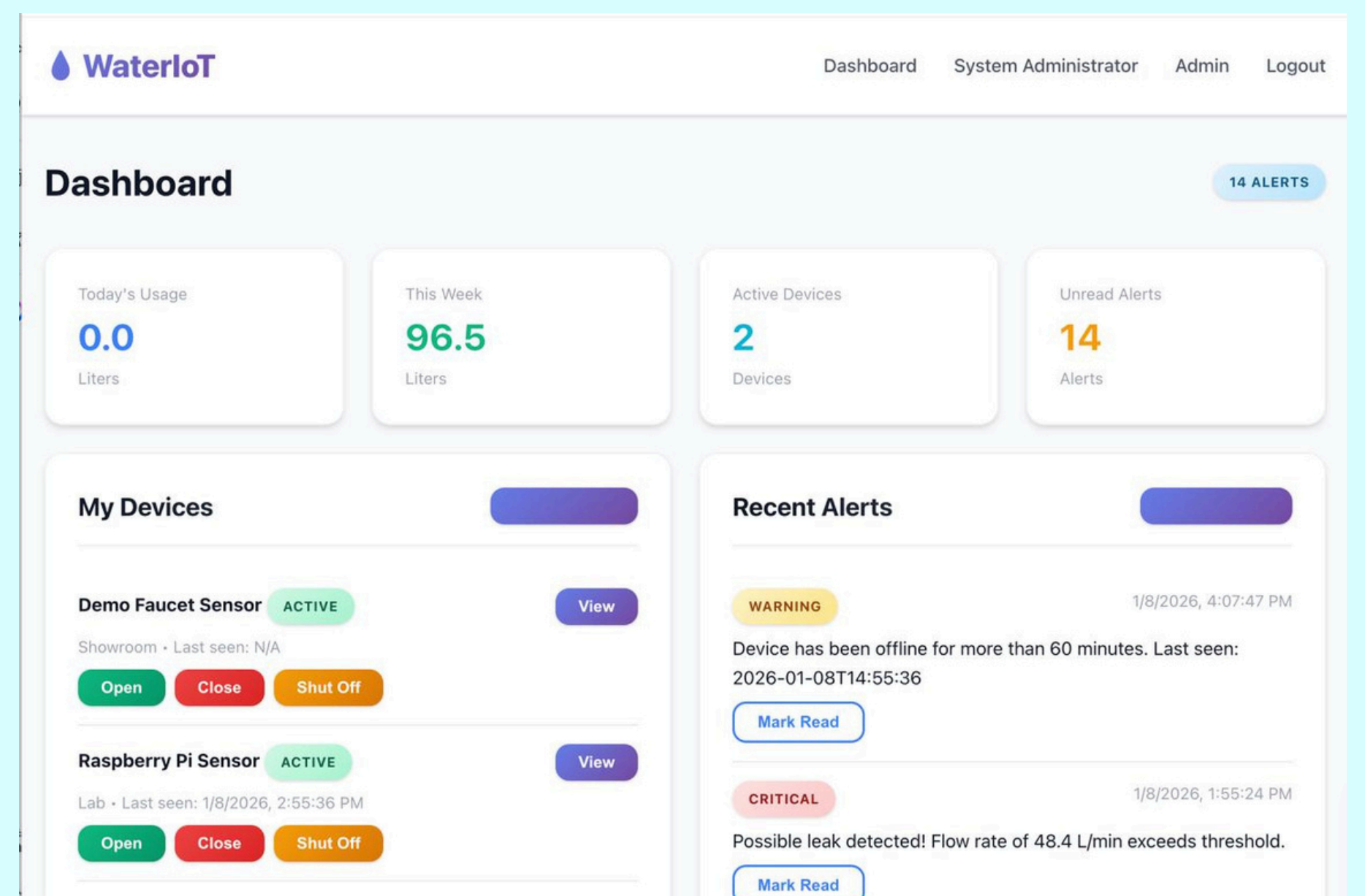


Figure 1: System Block Diagram

TinyML Design & Implementation

- Model trained on normal household consumption patterns
- Converted to RandomForestClassifier (Scikit-learn) for edge deployment

Anomaly Detection: If real-time flow deviates significantly from predicted values (e.g., high flow at night), the system automatically closes the solenoid valve and raises an alert.



EVENT TIME	FIXTURE / ROOM	DURATION	WATER USED	AVG FLOW	CONFIDENCE	CONTROL
8 Oct 2026 17:26:57	Kitchenfaucet	15.00 s	20.79 L	74.33 L/min	58.3%	FULL_OPEN
8 Oct 2026 17:23:17	Kitchenfaucet	8.99 s	12.98 L	78.40 L/min	69.8%	FULL_OPEN
8 Oct 2026 17:21:14	Kitchenfaucet	22.00 s	24.32 L	63.72 L/min	56.0%	FULL_OPEN
8 Oct 2026 17:20:10	Kitchenfaucet	19.00 s	19.49 L	58.54 L/min	60.8%	FULL_OPEN
8 Oct 2026 17:14:47	Kitchenfaucet	8.02 s	6.43 L	42.90 L/min	63.1%	FULL_OPEN
8 Oct 2026 17:12:42	Kitchenfaucet	17.00 s	14.59 L	48.62 L/min	61.0%	FULL_OPEN
8 Oct 2026 17:10:26	Kitchenfaucet	5.01 s	3.31 L	33.29 L/min	55.8%	FULL_OPEN
8 Oct 2026 16:57:19	Kitchenfaucet	5.99 s	5.07 L	43.91 L/min	57.1%	FULL_OPEN

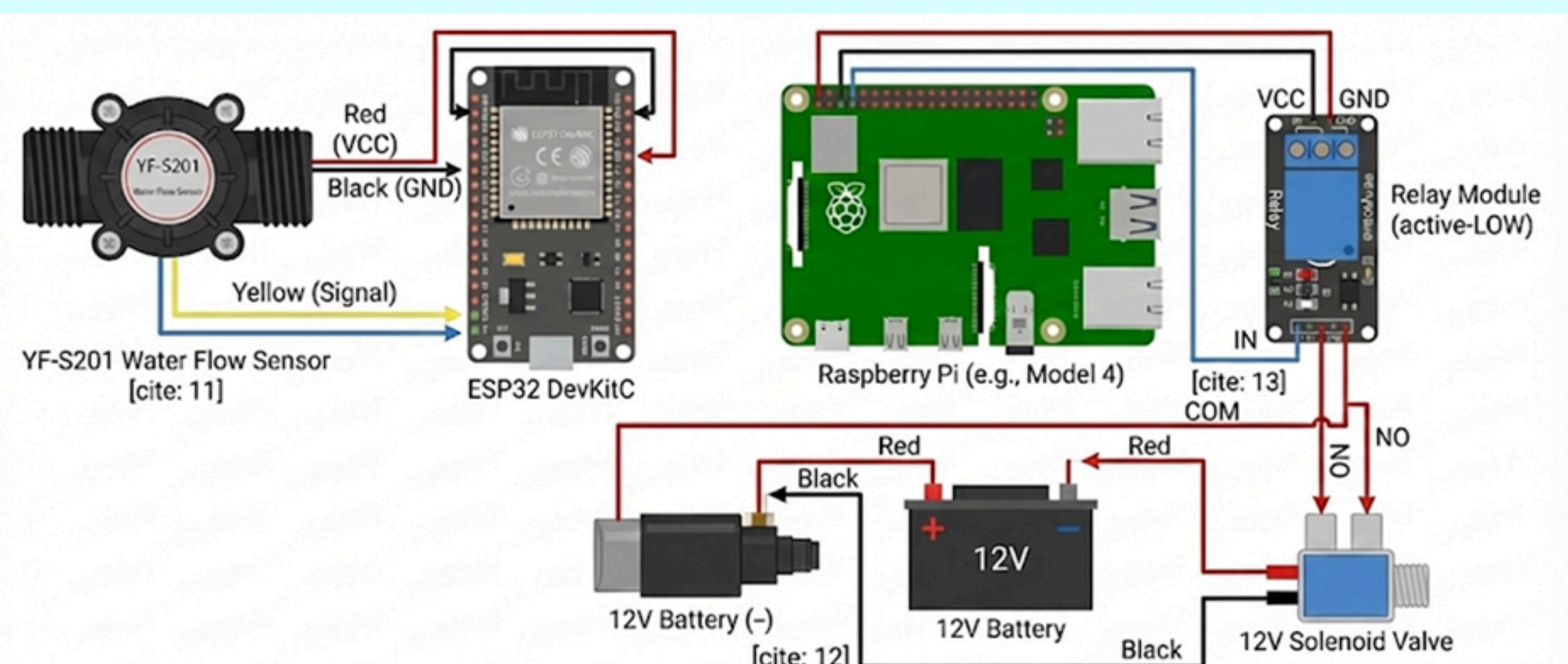


Figure 3: Integrated System Wiring Diagram: ESP32-Sensor, Pi-Relay Control, and 12V Power Circuit

Course: COMP 413 – Internet of Things

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References

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- [2] W. Yu et al., "A Survey on the Edge Computing for the Internet of Things," IEEE Access, vol. 6, pp. 6900-6919, 2018.
- [3] J. Walls et al., "Spring Boot in Action," Manning Publications, 2016.