

# **EVA: An Intelligent Voice-Controlled Robotic Assistant Integrating ESP32-CAM-Based Vision and Spatial Memory**

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**Abstract**-This paper presents EVA, an intelligent voice-controlled robotic assistant integrating speech recognition, ESP32-CAM vision, spatial memory and a multi-degree-of-freedom robotic arm. The system combines perception, decision making and actuation through a modular architecture. Continuous listening, interrupt-driven control and asynchronous software enable responsive human-robot interaction. The robotic arm executes basic manipulation tasks while the vision module detects objects and supports contextual awareness. The platform demonstrates the feasibility of integrating embedded AI, computer vision and robotics into a unified assistant suitable for laboratories, smart homes and assistive environments.

**Keywords**-Voice assistant, ESP32-CAM, robotic arm, embedded AI, computer vision, spatial memory, Arduino, human-robot interaction.

## **I. INTRODUCTION**

Artificial intelligence and embedded robotics have enabled the development of interactive robotic assistants capable of performing both digital and physical tasks. Conventional voice assistants are largely restricted to software operations, whereas robotic manipulators generally execute predefined motions without contextual awareness. EVA integrates these domains into a unified architecture by combining voice interaction, computer vision and robotic manipulation. The system continuously monitors user commands, processes environmental information and executes appropriate responses. The design emphasizes modularity, scalability and real-time operation. The proposed methodology emphasizes robustness, modular implementation, efficient communication, low-latency response, and practical deployment in assistive environments. Experimental observations indicate that

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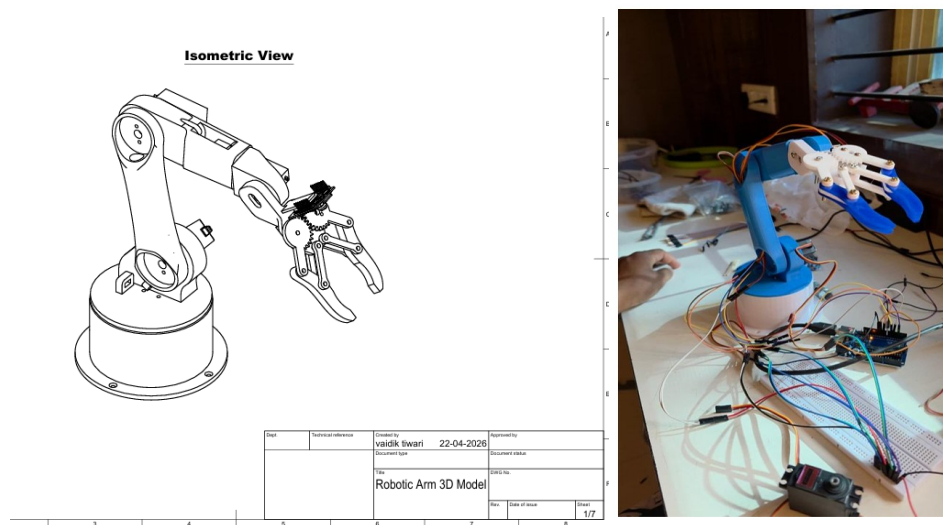
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**Figure 1.** Prototype of the developed robotic arm.

## II. SYSTEM ARCHITECTURE

The proposed platform consists of a robotic arm, Arduino controller, ESP32-CAM, microphone, speaker and power subsystem. The architecture follows perception–processing–action principles. Voice commands are processed by the software engine, visual information is obtained through the ESP32-CAM, and motion commands are transmitted to servo motors. Separate software modules handle speech recognition, synthesis, interrupt management and robotic control. The proposed methodology emphasizes robustness, modular implementation, efficient communication, low-latency response, and practical deployment in assistive environments. Experimental observations indicate that combining embedded vision, continuous speech interaction, and robotic manipulation improves usability compared with isolated subsystems.

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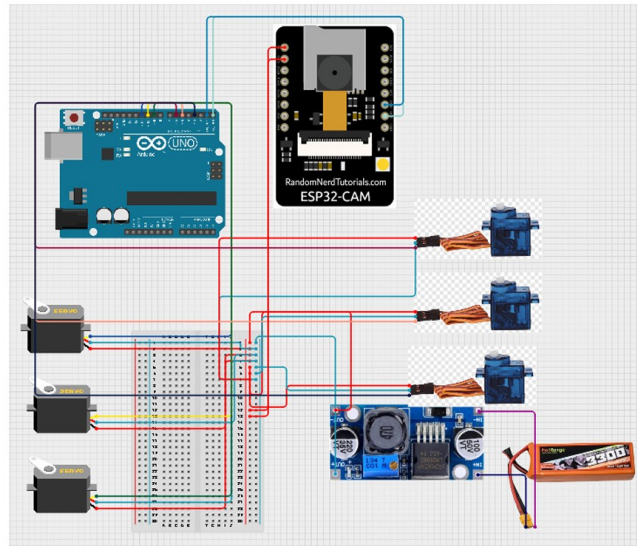


Figure 2. Electronic architecture of the proposed system.

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### **III. METHODOLOGY**

The robotic arm employs multiple servo motors providing several degrees of freedom for positioning and grasping. A lightweight end-effector carries the ESP32-CAM, enabling eye-in-hand perception. An external battery with a voltage regulator powers the actuators while maintaining stable controller operation. PWM signals control the servo positions, allowing repeatable movement. The proposed methodology emphasizes robustness, modular implementation, efficient communication, low-latency response, and practical deployment in assistive environments. Experimental observations indicate that combining embedded vision, continuous speech interaction, and robotic manipulation improves usability compared with isolated subsystems.

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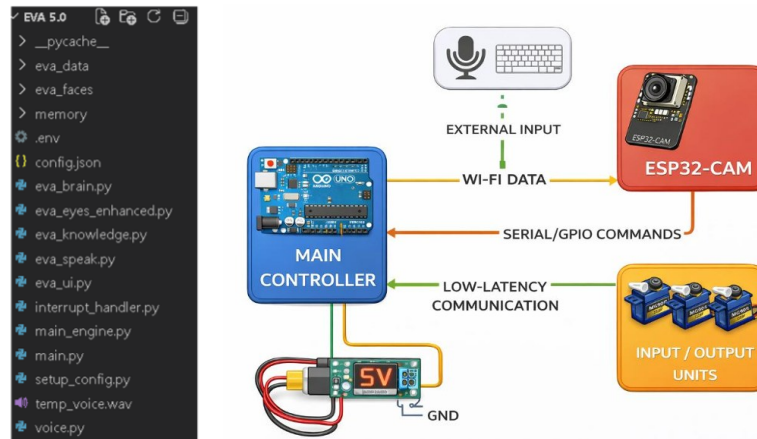


Figure 3. Software and Communication architecture with asynchronous modules.

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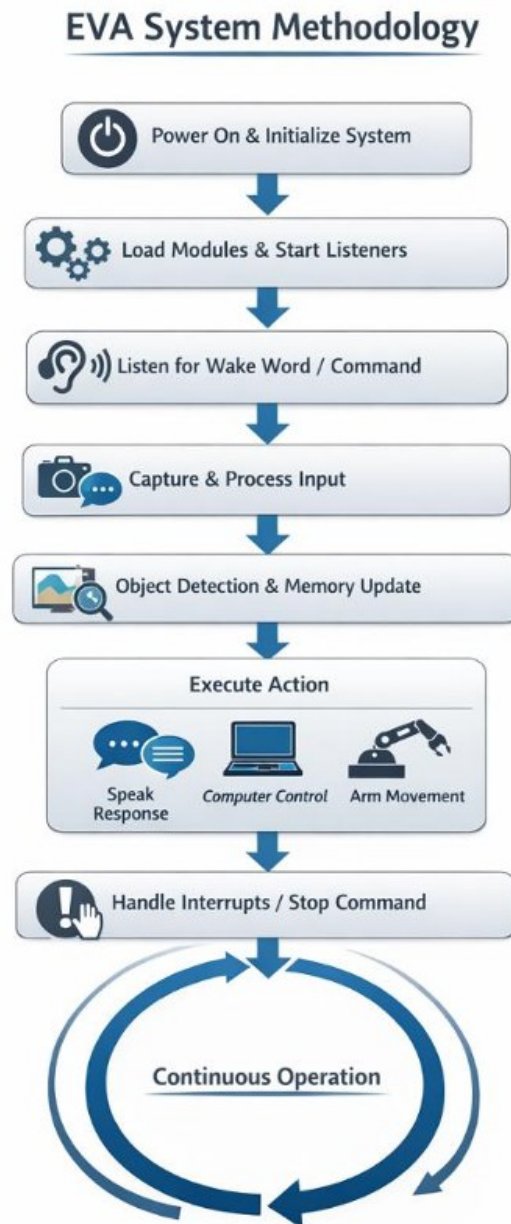


Figure 5. System workflow.

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#### IV. RESULTS AND DISCUSSION

Experimental evaluation demonstrated reliable speech-based computer control, stable robotic arm movement and successful object detection under indoor lighting. The integrated design reduced command latency and enabled coordinated digital and physical task execution. The ESP32-CAM effectively supported visual observation while spatial memory improved contextual interaction. The proposed methodology emphasizes robustness, modular implementation, efficient communication, low-latency response, and practical deployment in assistive environments. Experimental observations indicate that combining embedded vision, continuous speech interaction, and robotic manipulation improves usability compared with isolated subsystems.

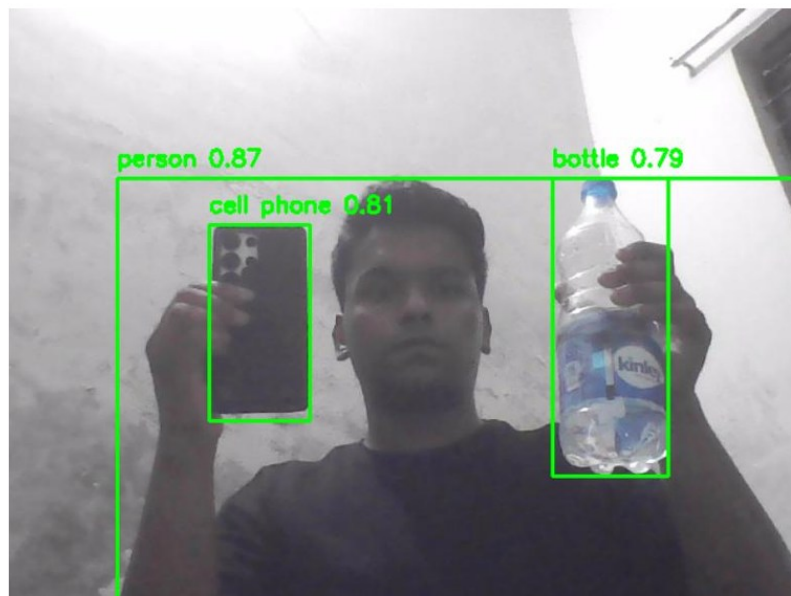


Figure 6: Object Detection using EVA Vision System (Indoor Test)

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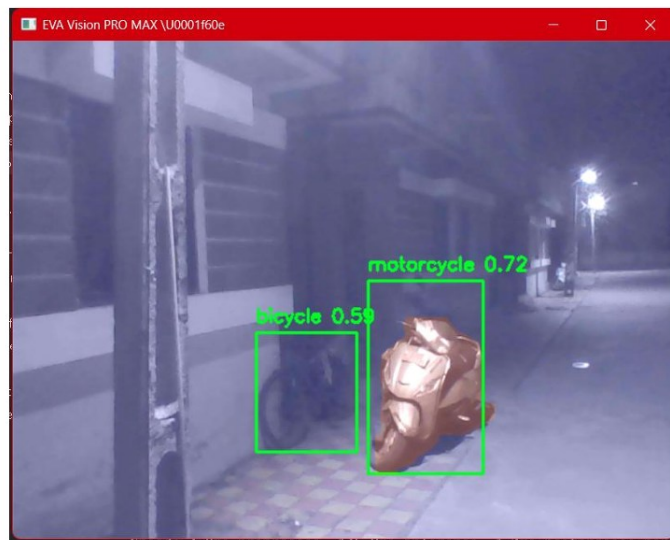


Figure 7: Outdoor Object Detection (Vehicle Identification)

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The proposed assistant bridges the gap between virtual assistants and robotic systems by integrating perception, reasoning and action. Although the prototype is intended for research and educational use, the architecture can be extended to industrial inspection, healthcare assistance and smart manufacturing. Future improvements include SLAM, deep learning-based object recognition and autonomous navigation. The proposed methodology emphasizes robustness, modular implementation, efficient communication, low-latency response, and practical deployment in assistive environments. Experimental observations indicate that combining embedded vision, continuous speech interaction, and robotic manipulation improves usability compared with isolated subsystems.

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## **V. CONCLUSION**

This study presented EVA, an intelligent voice-controlled robotic assistant that integrates speech recognition, computer vision, spatial memory, and robotic manipulation within a unified embedded platform. By combining continuous voice interaction with ESP32-CAM-based visual perception and a multi-degree-of-freedom robotic arm, the system enables seamless coordination between digital automation and physical task execution. The modular hardware and software architecture ensures scalability, low-latency communication, and reliable real-time performance, making the platform suitable for assistive robotics and smart automation applications. Experimental evaluation demonstrated that integrating voice control, object detection, and robotic actuation

significantly enhances system responsiveness and operational efficiency compared with conventional standalone assistants. Furthermore, the incorporation of spatial memory improves contextual awareness by enabling the system to recognize and recall object locations during interaction. Overall, EVA provides a practical framework for next-generation intelligent service robots and establishes a strong foundation for future research involving autonomous navigation, advanced machine learning, adaptive decision-making, and human-centered robotic assistance..

### **Author Contribution**

**Vaidik Tiwari:** Conceptualization, methodology, software development, hardware integration, robotic arm design, experimental investigation, data analysis, visualization, and writing original draft preparation.

**Jagdish Pampania:** Supervision, project administration, conceptual guidance, validation, technical review, manuscript editing, and final approval of the manuscript.

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### **Data Availability Statement**

No new data were generated during the study. All the data are contained within the manuscript.

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We acknowledge the use of Grammarly and Quill Bot, in correcting the English and Grammatical errors in the manuscript.

### **Conflicts of Interest**

The authors declare no conflict of interest.

### **Ethics Declaration**

This manuscript is a review article and does not involve any studies with human participants or animals performed by the authors. Therefore, ethical approval and informed consent were not required. The authors declare no conflict of interest, and all referenced works have been properly cited.

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