

# IMAGE BASED CAPTCHA SOLVER RESISTANCE SYSTEM

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**Abstract**— *With the growing sophistication of automated bot attacks, conventional CAPTCHA systems based on distorted text and image recognition have become less effective and often create accessibility challenges for users with disabilities. This research introduces a novel multi-modal CAPTCHA system that combines blink detection with One-Time Password (OTP) verification to improve both security and user experience. By applying computer vision techniques, the system tracks facial landmarks to identify natural blinking patterns, confirming human presence, while OTP authentication adds an extra layer of protection.*

*The system enhances resistance to automated bots by randomly switching between blink detection and OTP-based challenges, while also ensuring usability for individuals with visual, cognitive, or motor impairments. Experimental results show a blink detection accuracy of 95% and an OTP verification success rate of 98%, demonstrating the system's reliability and effectiveness. Future improvements may include adaptive security enhancements, expanded accessibility support, and rigorous testing against advanced adversarial attacks. This approach establishes a new benchmark for CAPTCHA systems by strengthening security, minimizing user inconvenience, and promoting an inclusive and seamless authentication process across diverse platforms.*

**Keywords**— *CAPTCHA, Blink Detection, OTP Verification, Computer Vision, Accessibility, Bot Detection, Cybersecurity, Multi-Modal Authentication.*

## 1. INTRODUCTION

With the rapid advancement of artificial

intelligence and the increasing prevalence of automated bot attacks, traditional CAPTCHA systems are facing significant challenges in effectively distinguishing between human users and bots. Conventional CAPTCHA methods, such as distorted text recognition and image-based puzzles, have become vulnerable to sophisticated machine learning algorithms capable of bypassing them with high accuracy. As a result, online security systems require more robust authentication mechanisms to prevent unauthorized access while maintaining a seamless user experience.

One of the primary concerns with traditional CAPTCHA systems is their declining effectiveness due to advances in deep learning and artificial intelligence. Bots can now analyze distorted text, recognize objects in images, and even solve complex puzzles with human-like accuracy. This has rendered many existing CAPTCHA systems obsolete, creating a security gap that attackers can exploit. Additionally, some CAPTCHAs rely on challenges that are repetitive and predictable, making it easier for bots to adapt and bypass them over time.

Beyond security concerns, traditional CAPTCHAs also present accessibility challenges for users with disabilities. Visually impaired users, for example, may struggle with text-based CAPTCHAs, while those with cognitive or

motor impairments may find it difficult to solve complex challenges that require precise interactions. Although audio CAPTCHAs exist as an alternative, they often suffer from poor clarity, background noise, and language limitations, making them frustrating for users. These accessibility issues highlight the need for a more inclusive and user-friendly CAPTCHA system that caters to a diverse range of users.

To address these challenges, this research proposes an innovative multi-modal CAPTCHA system that integrates blink detection and One-Time Password (OTP) verification. Blink detection leverages computer vision techniques to analyze facial landmarks and confirm the natural, involuntary blinking of a human user, ensuring a non-intrusive and user-friendly authentication process. On the other hand, OTP verification provides an additional layer of security by requiring users to enter a unique, time-sensitive code sent via email or SMS, ensuring that only genuine users can pass the authentication test.

By randomly alternating between blink detection and OTP verification, the proposed system introduces an element of unpredictability, making it more difficult for bots to bypass. This dynamic approach not only strengthens security but also enhances usability by reducing user friction. The combination of biometric-based authentication and a secondary verification mechanism ensures a more resilient, accessible, and efficient CAPTCHA system suitable for modern cybersecurity needs.

This paper presents a detailed exploration of the proposed multi-modal CAPTCHA system, including its design, implementation, security evaluation, and user experience testing. The remainder of this document is structured as follows: Section II provides a literature review on existing CAPTCHA systems and their limitations, Section III discusses the methodology and system architecture, Section IV presents preliminary results and analysis, and Section V concludes with discussions on future work and potential optimizations.

## II. RELATED WORK

Traditional CAPTCHA systems were designed to differentiate between human users and automated bots by utilizing text distortion, image-based challenges, and audio-based verification. However, as machine learning and artificial intelligence have advanced, these conventional CAPTCHA systems have become increasingly vulnerable to automated attacks. Ma and Deng [1] highlighted that modern optical character recognition (OCR) algorithms and deep learning techniques have significantly reduced the effectiveness of text-based CAPTCHAs. Their study emphasized the need for multi-modal CAPTCHA approaches that combine different verification techniques to enhance security and usability.

One of the emerging solutions in CAPTCHA development is biometric-based authentication, particularly blink detection. Zhang and Wang [2] explored the feasibility of using blink-based CAPTCHA by leveraging facial landmark detection to verify human presence. Their study demonstrated that blinking is a natural, involuntary action that bots struggle to replicate, making it a promising approach for CAPTCHA security. They further suggested that blink-based authentication can significantly improve user experience, as it requires minimal user effort compared to traditional CAPTCHAs.

Another enhancement in CAPTCHA security is One-Time Password (OTP) verification, which adds an extra layer of authentication. Chen and Lin [3] examined the integration of OTP authentication with CAPTCHA systems, arguing that time-sensitive verification codes provide an effective countermeasure against bots attempting to automate CAPTCHA-solving processes. Their research found that while OTP authentication improves security, it also introduces potential usability challenges, such as delays in OTP delivery or difficulty in entering the correct OTP within the time limit.

Comprehensive research on CAPTCHA mechanisms has been conducted to analyze their vulnerabilities and possible improvements. Johnson and Kuo [4] provided a detailed survey on various CAPTCHA methods, discussing how machine learning algorithms can bypass text-based and image-based CAPTCHAs. Their findings support the need for multi-modal CAPTCHAs that integrate biometric authentication, behavioral analysis, and additional verification layers to improve resilience against automated attacks.

A critical aspect of CAPTCHA design is ensuring a balance between security and usability. Wang and Xu [5] conducted a study on user experience in multi-modal CAPTCHA systems, concluding that CAPTCHAs should be secure yet easy to solve for human users. Their findings showed that biometric-based CAPTCHAs, such as blink detection, offer a more user-friendly experience compared to complex text or image-based challenges. However, they noted that lighting conditions and camera quality can impact the accuracy of blink detection, necessitating further refinements.

The integration of biometric-based CAPTCHA solutions has also been explored in recent research. Lee and Park [6] proposed an advanced CAPTCHA system incorporating facial recognition and eye-tracking technology to improve security while ensuring accessibility for individuals with visual or motor impairments. Their study indicated that biometric-based CAPTCHA solutions are harder for bots to bypass but must be optimized to function effectively across different device environments and user conditions.

Looking toward the future of CAPTCHA systems, researchers have examined adaptive and AI-driven

verification techniques. Li and Liu [7] proposed an adaptive CAPTCHA model that dynamically adjusts the difficulty level based on the user's behavior and security risk profile. Their findings indicated that such adaptive models can improve security while minimizing user frustration by providing simpler challenges for legitimate users and more complex verification steps for suspicious activity.

The combination of CAPTCHA and OTP security measures has been studied to assess their effectiveness against sophisticated attacks. Wu and Zeng [8] evaluated the impact of integrating OTP and CAPTCHA authentication, concluding that this multi-factor approach enhances security but can introduce latency issues due to network-related OTP delivery delays. They suggested using alternative verification methods, such as mobile app-based authentication or biometric verification, to address these concerns.

Security testing has also played a vital role in CAPTCHA research. Kumar et al. [9] explored machine learning-based bot detection techniques, demonstrating that bots are becoming increasingly capable of imitating human-like interactions to bypass CAPTCHA systems. Their study suggested implementing AI-driven CAPTCHA solutions that can analyze user interaction patterns in real-time to detect fraudulent attempts.

Finally, advancements in deep learning and computer vision have contributed to CAPTCHA evolution. Kumar and V.K.S. [10] studied computer vision-based approaches for security applications, emphasizing that deep learning techniques can both enhance and threaten CAPTCHA security. They suggested that future CAPTCHA systems must integrate real-time AI detection mechanisms to counteract adversarial AI attacks and GAN-generated CAPTCHA-solving models.

In summary, existing literature highlights the need for multi-modal CAPTCHA systems that incorporate biometric authentication, adaptive verification techniques, and AI-driven security measures to counteract evolving threats. While blink detection and OTP verification offer promising advancements, challenges related to usability, accessibility, and security testing must be addressed to ensure widespread adoption. Future research should focus on refining AI-based CAPTCHA defenses, optimizing biometric authentication accuracy, and enhancing real-time user adaptability to create a robust, user-friendly CAPTCHA system.

### III. MATERIALS AND METHODS

The proposed Multi-Modal CAPTCHA System integrates blink detection and One-Time Password (OTP) verification for enhanced security. The system is developed using computer vision techniques for facial landmark detection and secure authentication APIs for OTP generation and

verification. Blink detection is implemented using OpenCV and Dlib, which allow real-time analysis of user blinks to confirm human presence. The OTP authentication mechanism is integrated using services like Twilio or Authy, which generate and send time-sensitive verification codes to users via SMS or email. The system dynamically alternates between blink detection and OTP challenges to ensure enhanced security and accessibility.

Hardware requirements for the proposed solution include a working Windows or Linux operating system with a minimum RAM of 4 GB and a secondary storage capacity of 256 GB. The system requires a webcam for blink detection and a stable internet connection for OTP verification. The processor should be Intel Core i3 or above or an equivalent AMD Ryzen model for smooth execution of real-time facial recognition tasks.

Software requirements include a modern Internet browser (Google Chrome, Mozilla Firefox, or Microsoft Edge), a Python development environment (Anaconda or PyCharm), and libraries such as OpenCV, Dlib, NumPy, Pandas, Twilio API, and Flask for backend development. The system also requires email service integration for OTP delivery and a secure database (MySQL or Firebase) for temporary OTP storage and validation.

Tools and technologies used in the development include OpenCV for image processing, Dlib for facial landmark detection, Twilio API for OTP services, Flask for backend integration, and HTML, CSS, and JavaScript for the user interface. Additionally, adaptive challenge selection algorithms are implemented to randomly alternate between blink detection and OTP verification based on user behavior and risk factors.

### IV. EXISTING SYSTEM

Traditional CAPTCHA systems have been widely used to differentiate between human users and automated bots. These systems primarily rely on distorted text, image recognition, and audio challenges, requiring users to solve puzzles or identify specific objects within an image. The text-based CAPTCHA, one of the earliest implementations, presents warped or obscured letters and numbers that users must decipher. However, advancements in Optical Character Recognition (OCR) technology have made it easier for automated scripts and bots to bypass such challenges with high accuracy. Similarly, image-based CAPTCHAs, which require users to select specific objects from a grid, are now vulnerable to machine learning models trained on large datasets for object recognition.

A major drawback of the existing CAPTCHA systems is their poor accessibility and usability. Users with visual impairments struggle with text-based CAPTCHAs, while individuals with hearing disabilities may find audio CAPTCHAs difficult due to poor sound quality and background noise. Additionally, solving complex image or

text-based CAPTCHAs can be frustrating, leading to high user drop-off rates and negatively impacting the user experience. Studies have shown that these traditional CAPTCHAs increase cognitive load, making it difficult for users to interact smoothly with web applications.

Security vulnerabilities in existing CAPTCHA systems have also become a growing concern. Advanced artificial intelligence (AI) models and deep learning algorithms can now solve CAPTCHAs with high precision, rendering traditional methods ineffective. Attackers use automated CAPTCHA solvers, botnets, and adversarial AI techniques to bypass security measures, compromising website integrity and user data protection. Furthermore, captcha farms, where humans are paid to manually solve CAPTCHA challenges for bots, have further reduced the effectiveness of existing CAPTCHA solutions.

Despite efforts to improve security by introducing reCAPTCHA (Google's AI-based CAPTCHA system), issues still persist. Modern invisible reCAPTCHA attempts to verify users based on behavioral analysis and mouse movement tracking, but it raises privacy concerns as it collects user data for verification. Additionally, AI-generated adversarial attacks can still manipulate behavioral-based CAPTCHAs. Given these challenges, the need for a more secure, accessible, and user-friendly CAPTCHA system has become essential, leading to the exploration of biometric-based authentication and multi-modal CAPTCHA systems such as the blink detection and OTP verification approach.

## V. PROPOSED SYSTEM

Figure 1. Architecture of the proposed workflow

The proposed Multi-Modal CAPTCHA System integrates blink detection and One-Time Password (OTP) verification to enhance security, improve accessibility, and offer a seamless user experience. Unlike traditional text or image-based CAPTCHAs, which are increasingly vulnerable to machine learning-based attacks and automated solvers, this system leverages computer vision and secure authentication techniques to verify human presence. Blink detection is implemented using OpenCV and Dlib, which utilize facial landmark detection to track the user's eye movement and confirm natural blinks, an involuntary action that bots struggle to replicate. Simultaneously, OTP authentication adds an additional layer of security by generating a time-sensitive one-time password via Twilio or an equivalent secure API, which is sent to the user's registered email or phone number. To increase unpredictability and prevent bot bypass attempts, the system randomly alternates between these two verification methods based on a security risk assessment model. This approach ensures that bots cannot anticipate the CAPTCHA challenge, making it significantly harder to exploit. Furthermore, the proposed system is designed with accessibility in mind, offering audio-based verification options for visually impaired users and ensuring a smooth

experience for individuals with motor impairments. Unlike conventional CAPTCHAs that require manual interaction and cognitive effort, blink detection allows for a hands-free and effortless verification process, reducing frustration and improving usability. Additionally, behavioral analysis and AI-driven adaptive security mechanisms can be incorporated to further enhance bot detection capabilities and detect suspicious activity in real-time. The proposed Multi-Modal CAPTCHA System effectively addresses the security vulnerabilities, accessibility challenges, and usability issues associated with traditional CAPTCHAs, making it a more robust, user-friendly, and future-proof authentication solution for online platforms.

## VI. METHODOLOGY

The Multi-Modal CAPTCHA System is designed to improve security and accessibility by integrating blink detection and OTP verification as a dual authentication mechanism. This system addresses the vulnerabilities of traditional CAPTCHA methods by using computer vision-based human verification and secure one-time authentication codes to ensure that only legitimate users gain access. The proposed methodology follows a structured approach, incorporating facial recognition, random challenge selection, secure authentication, and accessibility features to create a robust, user-friendly CAPTCHA system. The following subsections outline the key components of the methodology:

### A. System Design

The system is designed as a multi-modal CAPTCHA solution, where blink detection and OTP verification work in an alternating manner to introduce unpredictability and enhance security. The backend system determines which verification method to use based on predefined security conditions or selects a method randomly to prevent bots from adapting. A web interface is developed to interact with users, while the backend logic handles verification, data storage, and security monitoring.

### B. Blink Detection Mechanism

Blink detection is implemented using computer vision techniques, primarily OpenCV and Dlib, to track facial landmarks and recognize natural blinks. When the system prompts for blink detection, the user must blink within a specified time frame, ensuring real-time human verification. The system analyzes eye movement patterns to confirm a natural blink, distinguishing between human users and automated scripts or fake images. If the blink is detected correctly, access is granted.

### C. OTP Generation and Authentication

If the system selects OTP verification, a secure one-time password (OTP) is

generated using an OTP service provider such as Twilio or Firebase. The OTP is sent to the user's registered email or phone number, ensuring that only the intended recipient can complete the authentication process. The user is required to enter the OTP within a limited validity period, after which the system verifies the code. If the entered OTP matches the system-generated one, the CAPTCHA is successfully solved.

#### D. Randomized Challenge Selection

To prevent automated attacks and improve security, the system dynamically alternates between blink detection and OTP authentication. This selection can be made randomly or based on security conditions such as IP address analysis, login frequency, or suspicious user behavior. By making the CAPTCHA unpredictable, bots cannot pre-program solutions, thereby increasing resilience against automated CAPTCHA solvers.

#### E. Accessibility and Usability Considerations

To make the system inclusive, accessibility features are incorporated. Users with visual impairments can opt for audio-based CAPTCHA, where an OTP is read aloud. Additionally, customizable verification alternatives, such as voice commands or gesture-based inputs, can be provided for users with motor impairments. The system ensures that all users, regardless of disabilities, can interact effectively with the CAPTCHA.

#### F. Security and Anti-Spoofing Measures

Advanced security features are integrated to prevent spoofing and automated attacks. For blink detection, liveness detection algorithms ensure that bots, deepfake videos, or static images cannot replicate natural human eye movements. For OTP verification, rate limiting and IP tracking mechanisms are used to prevent brute force attacks or unauthorized OTP requests. Additionally, machine learning-based anomaly detection can flag suspicious user behavior in real-time.

The performance of the system is assessed through usability testing, security analysis, and system optimization. Metrics such as blink detection accuracy, OTP delivery time, user success rates, and security breach attempts are analyzed to refine the system. The goal is to minimize false positives and negatives, optimize response times, and ensure a seamless experience while maintaining high-security standards. Continuous testing and feedback collection will help improve system efficiency and adapt to emerging threats in CAPTCHA security.

This methodology ensures that the Multi-Modal CAPTCHA System is secure, user-friendly, and resistant to modern automated attacks, providing a more effective alternative to traditional CAPTCHA systems.

## VII. RESULTS

The implementation of the Multi-Modal CAPTCHA System was evaluated based on security effectiveness, user experience, and accessibility performance. The blink detection module, developed using OpenCV and Dlib, achieved a 95% accuracy rate in correctly identifying natural blinks under standard lighting conditions, with minor performance degradation in low-light environments. The OTP verification module, integrated using Twilio API, demonstrated 98% successful OTP delivery within an average of 3–5 seconds, ensuring timely authentication. Randomized challenge selection effectively alternated between the two verification methods, preventing predictable attack patterns by bots. Usability tests conducted with a diverse set of users, including individuals with disabilities, indicated a high level of user acceptance, with 80% of participants preferring blink detection over traditional text-based CAPTCHAs due to its hands-free and effortless verification process.

## VIII. DISCUSSION

The results indicate that the Multi-Modal CAPTCHA System provides a more secure and user-friendly alternative to conventional CAPTCHAs, successfully addressing bot attacks, accessibility issues, and user experience concerns. Unlike traditional text-based and image-based CAPTCHAs, which are vulnerable to machine learning-based solvers, this system leverages biometric verification and real-time authentication, making it significantly harder for automated scripts to bypass. However, certain challenges remain, including slight accuracy drops in poor lighting conditions and occasional OTP delivery delays due to network dependencies. Future improvements will focus on enhancing the adaptability of the blink detection model using AI-driven liveness detection, optimizing OTP response times, and introducing alternative accessibility features for users with motor impairments. Overall, the proposed system demonstrates strong potential for widespread adoption, providing an efficient, secure, and inclusive CAPTCHA solution for modern cybersecurity needs.

## IX. CONCLUSION

The proposed Multi-Modal CAPTCHA System, integrating blink detection and OTP verification, addresses the security vulnerabilities and accessibility challenges of traditional CAPTCHA methods. By leveraging computer vision techniques for real-time facial landmark detection and secure authentication APIs for time-sensitive OTP verification, the system ensures robust protection against automated attacks while maintaining a user-friendly experience. The randomized challenge selection enhances unpredictability, making it difficult for bots to bypass the verification process. Additionally, accessibility features such as audio-based CAPTCHA and alternative input

methods ensure inclusivity for users with disabilities. Preliminary evaluations indicate high accuracy in blink detection and OTP success rates, demonstrating the system's efficacy in distinguishing humans from bots. Future enhancements will focus on AI-driven security optimizations, improved anti-spoofing techniques, and scalability improvements to adapt to emerging cybersecurity threats. The Multi-Modal CAPTCHA System sets a new standard for CAPTCHA security by providing a secure, adaptive, and accessible verification mechanism that enhances both user experience and system protection.

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