

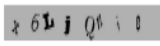

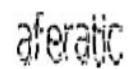

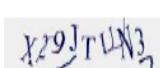


Some popular examples of each are shown in Table 1 and most existing CAPTCHAs are text-based. The user is presented with visually distorted text and asked to type it in correctly to prove he or she is a human and not a computer algorithm masquerading as a person. Many mobile devices lack a physical keyboard, which makes text-based input cumbersome and error-prone [10]. Further, most text based CAPTCHAs are (English) language-dependent and not suitable for multilingual worldwide usage. This paper mitigates the shortcomings of existing approaches and proposes a new CAPTCHA, termed as CAPTCHA, which leverages touch screen technology in mobile devices to make CAPTCHAs user-friendly and intuitive. CAPTCHA presents users with a composite image containing several visually distorted human faces along with other objects and non-real faces embedded in a complex background pattern. To prove that a user is human, users must solve the CAPTCHA by correctly selecting only the real human faces without choosing any other objects or non-real face images. If this is successfully done, the user is considered to be human and granted access to the secured resource. Fig. 1 shows an

example of how a CAPTCHA test can be correctly solved. In most cases, solving an instance only requires two or three taps from the user, making it extremely quick to complete and mobile device-friendly.

**Key Contributions of this Research Includes:**

- Design of an interactive non-keyboard-based (touch screen-compatible) image CAPTCHA to facilitate easy use on mobile devices.
- Generation of computationally-challenging face detection CAPTCHA tests to provide enhanced security.
- Utilization of genetic learning algorithms to optimize CAPTCHA parameters for better human performance and drastically lower the attack success rates of computer algorithms.
- Development of large-scale human and automated testing processes to evaluate performance of the proposed image-based face detection CAPTCHA.

CAPTCHA	Modality	Mode of Operation	Human Accuracy	Attack Accuracy	Sample
AltaVista [11], [12]	Text: 8 random characters	Each character is rendered in a different font, ransom-note style. Different rotations and distortions are applied to each letter.	-	Reduced page accesses by "over 95%"	
EZ-GIMPY [11], [13]	Text: 1 English word	Word is randomly distorted by adding white lines and deformations.	-	92%	
ScatterType [14], [15], [16]	Text: English word-like non-dictionary string of 6-8 characters	Characters are segmented into many pieces then systematically scattered.	Up to 95%	-	
BaffleText [17]	Text: Pronounceable English non-word of 5-8 characters	Text overlaid with random geometric shapes, difference masking applied.	89%	25%	
MSN [18], [19]	Text: 8 random characters	Characters are distorted and rotated, then arcs added to visually connect characters.	-	over 90%	

Handwritten [20]	Text: Full name of a city	Uses images of handwritten city names taken from U.S. mail.	100%	4-9%	
reCAPTCHA [21], [22], [23]	Text: 2 English words	Scanned words that failed OCR presented side-by-side. Some additional distortions may be applied.	-	30%	
ESP-PIX [24]	Images: 4 images	User selects category to describe images from predefined list.	-	High random guess rate	
Asirra [25], [26]	Images: 12 images of cats and dogs	User identifies cats or dogs.	High	82.7%	
Scene Tagging [27]	Images: Small number of images on background image	User answers questions relating to number of images, placement, or relationship to each other.	96.6%	2.6%	
MosaHIP [28]	Images: Collage of many images	User drags descriptor labels on top of images they represent.	98%	4.1%	
IMAGINATION [29], [30]	Images: Collage of images	User clicks on center of one image then categorizes that image.	70%	4.95%	
Digg Audio [31]	Audio: Audio recording of random letters and numbers	User enters information from audio.	-	71%	
Video [32], [33]	Video: Flash video	User types three words describing video.	90%	13%	

Table 1: Summary of selected existing CAPTCHAs

## 2. Literature Survey

### Image-based face detection CAPTCHA for improved Security

They demonstrate an implementation of novel image-based face detection CAPTCHA to add an additional layer of security in web-based services. Existing CAPTCHAs are vulnerable to computer attacks. Text-based CAPTCHAs are vulnerable to advanced OCR technologies. Image-based CAPTCHAs use a small subset of images and are susceptible to random guessing. When the images or videos are selected

from a large database, the users are presented with limited options making it susceptible to random guessing or machine learning techniques. Speech recognition software is used to exploit audio-based CAPTCHAs. Minimizing the vulnerabilities to prevent computers from solving the CAPTCHAs also makes it challenging for humans, often requiring multiple attempts to successfully solve the CAPTCHA.

They proposed an algorithm to generate an image-based CAPTCHA that uses the concept of face detection. The proposed algorithm embeds multiple human faces and non-

human faces in a background image to create image CAPTCHAs. The background image contains randomly generated overlapping blocks of different shapes and contrast levels. The faces were selected from the CMU face database and were subjected to known distortions. By varying different parameters, the intensity of distortion is controlled to produce low, medium, and high levels of distortion. All these processing make it very challenging for face detection algorithm to accurately select all human faces embedded in the CAPTCHA image, while humans generally are able to identify the embedded human faces with relative ease. The design objective Image-based face detection CAPTCHA for improved security [28] is to generate CAPTCHA images such that the computers attack rates are minimized while human accuracy to solve the same CAPTCHA is considerably increased. The use of image quality metrics to study the characteristics of images and design optimal images is briefly presented Here.

An extensive experimental study demonstrates these important features of the image-based face detection CAPTCHA. In addition, key factors that need to be considered in designing image-based face CAPTCHAs are described in detail. The proliferation of new generation mobile devices increasingly uses Internet-based applications and it is imperative they be made secure and resilient to attacks. These devices generally do not have a convenient keyboard and therefore the proposed image-based face detection CAPTCHA is ideally suited for clicking to solve the CAPTCHA rather than typing. Since there is no text involved, this CAPTCHA is language-independent and can be widely used by a large audience.

### CAPTCHA Based on Human Cognitive Factor

Here, illustrates a new design for CAPTCHA system based on human cognition. This model demonstrates the ability of human to find the answer that other bots and external programs fail to interpret and evaluate. The conducted survey explains the usability of this new form of CAPTCHA and provides valuable feedback to design the overall system and types of question pattern. This framework can easily be extended to specific website to include question of any particular area of interest.

### 3. Conclusion

The unique touch screen technology of mobile devices can be leveraged to create an additional layer of security that is both effective and user friendly. The proposed genetically optimized CAPTCHA works efficiently on both touch screens used by tablets and smart phones and on traditional computers, achieving a high 88% human accuracy rate during evaluation. It does so without compromising performance, offering an effective 0% automated attack rate. This combination of low attack rates, high human accuracy rates, and convenient mobile device usage provides major improvements over existing desktop centric security CAPTCHAs in widespread use today.

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