

Assistive guidance system for the visually impaired

Rohit Takhar¹, Tushar Sharma¹, Udit Arora¹, Sohil Verma¹

¹ Division of Computer Engineering, Netaji Subhas Institute of Technology, New Delhi 110078, India

Corresponding Author:

Udit Arora¹

Email address: udita1.co@nsit.net.in

ABSTRACT

In recent years, with the improvement in imaging technology, the quality of small cameras has improved significantly. Coupled with the introduction of credit-card sized single-board computers such as Raspberry Pi, it is now possible to integrate a small camera with a wearable computer. This paper aims to develop a low-cost product, using a webcam and Raspberry Pi, for visually-impaired people, which can assist them in detecting and recognizing pedestrian crosswalks and staircases. There are two steps involved in detection and recognition of the obstacles, i.e., pedestrian crosswalks and staircases. In detection algorithm, we extract Haar features from the video frames and push these features to our Haar classifier. In classification algorithm, we first convert the RGB image to HSV and apply histogram equalization to make the pixel intensity uniform, followed by image segmentation and contour detection. These detected contours are passed through a pre-processor which extracts the regions of interest (ROI). We applied different statistical methods on these ROI to differentiate between staircases and pedestrian crosswalks. The detection and recognition results on our datasets demonstrate the effectiveness of our system.

1. INTRODUCTION

Our sense of vision has developed over millions of years of evolution, and coupled with the visual cortex has been one of the key ingredients for the ascension of humans as the dominant species on our planet. Without proper vision, humans remain inert towards a wide range of stimuli perpetuating from the outside environment. Hence, there is an urgent demand for low-cost solutions which can assist people with impaired vision.

In the past few decades, many researchers and engineers have tackled this problem and developed prototypes for obstacle detection, object detection and classification using computer vision and sonar assisted obstacle detection. These solutions involve the manufacturing of customized hardware along with expensive cameras.

In the context of assistive technology, mobility takes the meaning of "moving safely, gracefully, and comfortably". It relies in large part on perceiving the properties of the immediate surroundings, and it entails avoiding obstacles, negotiating steps, drop-offs, and apertures such as doors, and maintaining a possibly rectilinear trajectory while walking. Although blind people are more in need of mobility aids, low-vision individuals may also occasionally trip onto unseen small obstacles or steps, especially in poor lighting conditions.[1]

2. RELATED WORK

This section reviews existing navigational assistance for the visually impaired. First, the mobility of vision-disabled people is presented in Section 2.1. Then, conventional mobility tools are

46 described in Section 2.2. Next, assistive technologies based on computer-vision are reviewed in
47 Section 2.3.

48

49 **2.1. Mobility of the visually impaired**

50 Mobility is the ability to travel safely and independently from one place to other places and is an
51 essential aspect of daily life. Vision plays a very significant role in mobility as it allows sighted
52 people to collect most of the information required for perceiving the surrounding environments.
53 People with vision loss must rely on other senses (e.g., hearing and touch) to gather information
54 about the surrounding objects, and therefore face great difficulties in traveling[2].

55 The mobility of blind people includes two aspects: perception and orientation. Perception
56 refers to a blind traveler obtaining information about the environment via the non-vision senses
57 in order to detect obstacles and identify landmarks. Orientation refers to the knowledge to
58 recognize the position in relation to surrounding objects and the location in the routes of the
59 entire journey. To understand the surrounding environments, vision disabled people employ
60 touching and hearing as primary modalities [8]. In near space, both touching and hearing are
61 used to perceive objects. In far space, the people rely mainly on hearing to find objects.
62 Orientation is performed based on the identification of landmarks. Blind people often employ
63 many different types of landmarks to determine their position in the environments. Example
64 landmarks include rises and falls in the walking path, changes in the texture of the walking path,
65 the presence of walls and hedges, traffic sounds and temperature changes.

66

67 People with vision impairments have three major barriers while traveling [3,4] as follows:

- 68 1) Relying mainly on touching and hearing, visually impaired people are limited in detecting
69 and avoiding obstacles, finding the travel path, and identifying hazards ahead.
- 70 2) They have great difficulties in determining routes for a journey, understanding the scene
71 layout, and identifying their position in the scene.
- 72 3) They cannot obtain visual or textual information such as road signs and bus numbers.

73 These barriers make visually impaired people unable to travel safely and independently in
74 unknown environments.

75

76 **2.2. Traditional aids**

77 There are two popular traditional aids for navigation of the visually impaired: the white cane and
78 the guide dog.

79 **The white cane** is a hand-held tool and is considered as an extension of the user's arms. This
80 tool is a lightweight and cylindrical cane, and is often made of aluminium. It is adjustable to the
81 height of the user and includes a purposely designed handle-grip and tip. To detect objects on the
82 walking path, the user swings the white cane from side to side while the tip of the cane directly
83 contacts with the ground. Based on vibrations and force feedback from the cane, the user can
84 recognize objects ahead and characteristics of the ground. To effectively use a white cane, blind
85 people often require about 100 hours of training [3], because an incorrect use of the cane can
86 lead to dangers for both the user and others.

87 The white cane is a cheap, reliable and robust tool to detect obstacles and drop-offs at
88 ground levels, identify the characteristics (e.g., texture and hardness) and conditions of the
89 ground, and find the walking path at close range [4]. However, this tool is unable to detect
90 obstacles at torso and face levels, which are dangerous for blind travelers [3, 4]. Furthermore,
91 using the white cane over a long distance causes arm fatigue for travelers [4].

92

93 **The guide dog** is a specially trained dog to assist the blind in mobility. A blind user can travel
94 safely with a trained dog. The dog is trained to have the following major skills [5]:

- 95 • Walking in a straight line and on the left-hand side slightly ahead of the user.
- 96 • Stopping at all curbs, the top, and bottom of stairs.
- 97 • Waiting for user's commands before crossing roads.
- 98 • Avoiding obstacles at head level.
- 99 • Avoiding narrow spaces where a person and a dog cannot walk through side by side.
- 100 • Boarding and traveling on all different types of public transportation.
- 101 • Taking the user to a lift.
- 102 • Refusing the user's commands that may cause dangers for the user.
- 103 • Recognizing the familiar routes.

104

105 The user controls the guide dog and receives information about its activities via a handle
106 encircled the dog's chest. With the above trained skills, the guide dog provides assistance for
107 safe travel to vision-disabled people in familiar environments. However, a guide dog typically
108 requires about two to three years and 40,000 USD to train and has a working life of only eight to
109 ten years [3]. Due to the high cost of training and maintenance, just three percent of the visually
110 impaired population is reported to use a guide dog. Furthermore, the guide dog cannot detect
111 obstacles at head levels.

112 Both the white cane and guide dog are useful aids for blind travelers in detecting and
113 avoiding obstacles at ground levels in the near space. However, these tools cannot assist blind
114 travelers in determining spatial and geographical orientations, recognizing locations and finding
115 routes for their journey in unfamiliar environments.

116

117 **2.3. Technology aids**

118 Many assistive technology systems have been developed to improve the mobility of vision
119 disabled people. Some computer-vision based assistive systems are explained below.

120

121 **Computer-vision based systems**

122 Computer-vision based systems use images captured from cameras to obtain information about
123 the environment. Many studies have been conducted on assistive navigation of blind people.

124 The Virtual White Cane system[14] detects objects using a smartphone and a laser pointer as
125 shown in Fig. 2.1. The object's distance from the user is estimated from the image of the laser's
126 reflection off the planar surface, using active triangulation. The image is captured by the camera
127 of the smartphone. The detected objects are conveyed to the user through vibration of the
128 smartphone.

129 Several vision-based systems focus on detecting crossings of zebra patterns at traffic
130 junctions for the visually impaired [6, 7, 8]. These systems apply the image processing and
131 pattern recognition techniques to detect lane markers in the image captured by a camera, and
132 determine then the walking region for users. For example, Se proposes an assistive system to
133 find the zebra-crossing lane in the image [7]. The walking lane is determined by finding lane
134 markers using the Hough transform.

135 In [8], Ivanchenko *et al.* design the Crosswatch system that employs a mobile phone to
136 find and locate zebra-crossings as shown in Fig 2.2. The authors proposed using a graphical

137 model with geometric and color cues to find the borders of lane markers in the image captured
138 from the phone camera.

139 In [7], Uddin and Shioyama employ the bipolar feature of zebra patterns to segment the
140 crossing region and verify the detected crossing using the constant width periodic feature of
141 white bands in a zebra-crossing. The system by Radvanyi *et al.* segments first the crossing region
142 using color features, and then verifies the detected region by finding the lane markers with an
143 adaptive thresholding technique [8].

144 In general, most existing assistive systems based on computer-vision focus on detecting
145 obstacles.

146

147 3. APPROACH

148 This paper proposes a robust approach for the identification of staircases as well as pedestrian
149 crosswalks adaptive towards varying lighting conditions.

150

- | |
|---|
| <ul style="list-style-type: none"> ● Capture frame through webcam. ● Apply classifier. <ul style="list-style-type: none"> ○ Object Detected <ul style="list-style-type: none"> ■ Convert original image to HSV color scale. Split image into H, S and V. Select V matrix and assign to <i>image</i> matrix for further calculations. ■ Apply CLAHE to the <i>image</i>. ■ Apply image segmentation and threshold to differentiate dark and light spots. ■ Find contours and traverse through all contours to remove the outliers. Apply following conditions: <ul style="list-style-type: none"> <input type="checkbox"/> Ratio of minor axis and major axis of ellipse. <input type="checkbox"/> Area of ellipse. <input type="checkbox"/> Orientation of ellipse. ■ Apply LIS(Longest Increasing Subsequence) algorithm to find the number of ellipses which can be part of the zebra-crossing. ■ <i>if len(sequence) >= 4:</i>
 “Pedestrian Crosswalk”
 <i>else:</i>
 “Stairs” ○ No object Detected |
|---|

151

TABLE 1: PROPOSED ALGORITHM

152

153 3.1. Identification of Staircase and Pedestrian Crosswalk

154 The first step involves image acquisition through suitable hardware, which in this case is a USB
155 webcam. The next step involves the identification of any staircase like object in the acquired
156 image and draws a boundary around it. The identification is done through Haar’s feature-based
157 classifier.

158

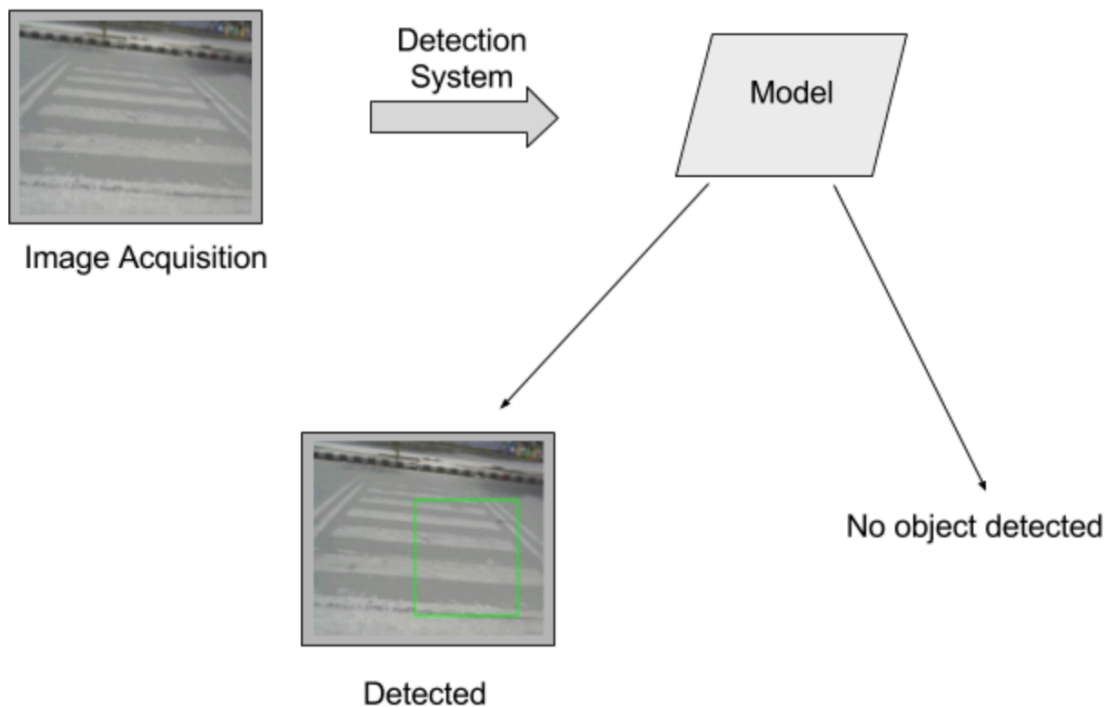


FIGURE 1: FLOW CHART OF DETECTION SYSTEM

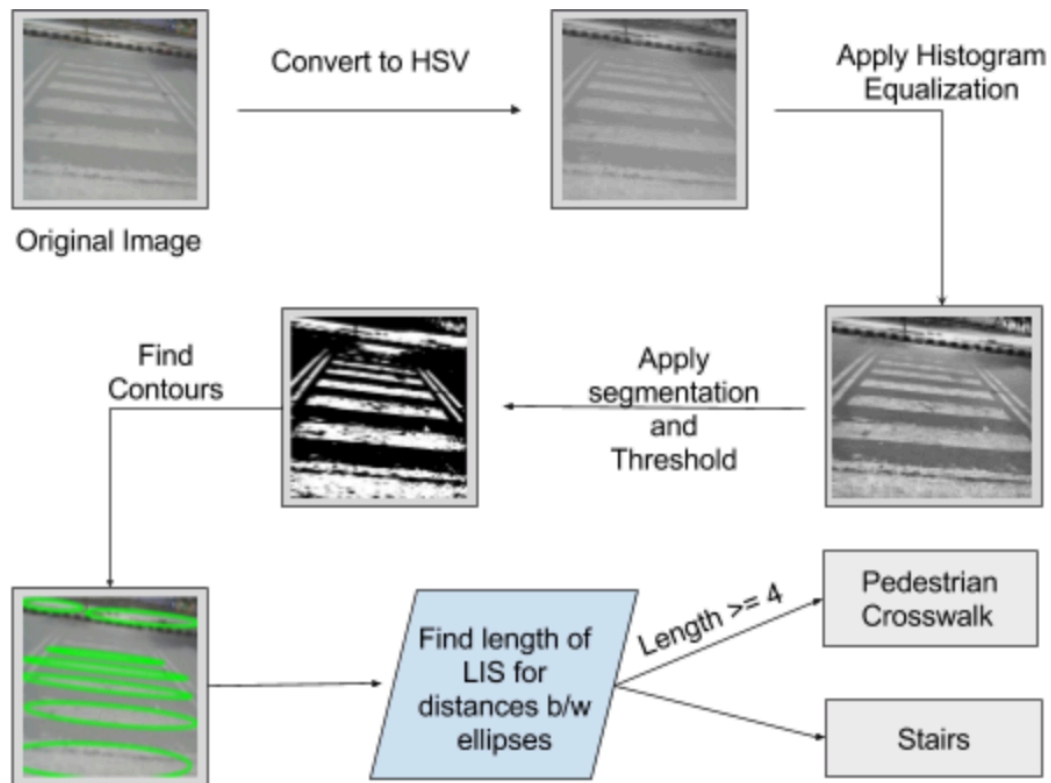
159
160
161
162
163
164
165
166
167
168
169
170
171

3.2. Classification between Stairs and Crosswalks

Upon detection, histogram equalization is applied to the acquired frame which increases the contrast of the image. Segmentation and thresholding are subsequently done to highlight the white and black patterns in the image which helps in differentiating zebra-crossings from staircases.

In the next step, contours are found in the observed frame and ellipses are drawn over them. The distances between individual ellipses are found and are arranged in increasing order. If the length of the longest increasing subsequence is found to be higher than 4 (this number is obtained through repeated field observations), then a prediction of zebra-crossing is made otherwise a staircase is predicted.

172



173
174
175
176
177
178
179
180
181
182
183
184
FIGURE 2: FLOW CHART DEPICTING IMAGE PROCESSING TECHNIQUES

3.3. Companion Android application

An Android application was built for activating the detection system through touch gestures. The application also provided audio feedback for identified obstacles.

3.3.1. Activation

In computing, multi-touch is technology that enables a surface (a trackpad or touchscreen) to recognize the presence of more than one or more than two points of contact with the surface. This plural-point awareness was used to implement additional functionality, such as pinch to start the data gathering, and the reverse gesture to deactivate the system.



185
186
FIGURE 3: GESTURE

187 3.3.2. Communication between phone and Raspberry-Pi

188 To establish communication with the Android smartphone the Raspberry Pi is configured as a
 189 Wi-Fi hotspot with WPA2 secure encryption. The phone is connected to this network with this
 190 one-time process. The Pi then starts broadcasting signals upon activation from the user's end. If a
 191 staircase or a crosswalk is identified, a UDP packet is generated and sent to the Android
 192 smartphone using Wi-Fi as the carrier.

193 3.3.3. Voice-over for audio feedback

194 Google Text-to-Speech is a screen reader application developed by Google Inc for its Android
 195 operating system. It powers applications to read aloud (speak) the text on the screen. In this
 196 implementation, it is used provide audio cues to the blind user so that he/she can avoid or
 197 maneuver the identified obstacle.

198

199 4. EXPERIMENTS AND RESULTS

200 In this paper, the proposed system was tested in different environments with different lighting
 201 conditions. We captured ~1200 frames in different environments and calculated the true
 202 positives and false positives. We observed that in low visibility areas, the performance of the
 203 system decreases, but it is still satisfactory. In the areas with sufficient light, we observed that the
 204 system accurately identifies staircases and zebra crossings most of the time.

Visibility	True Positives (%)	False Positives(%)
Low (Indoor)	52.33	14.88
Sufficient(Indoor)	72.12	4.13
High(Outdoor)	74.47	4.91

205 TABLE 2: ACCURACY OF THE SYSTEM

206 Here is a video showing the system in action – <https://youtu.be/5gA2s16UuW4>

207

208 5. CONCLUSION

209 The primary goal of this paper was to develop a low-cost wearable prototype for visually
 210 impaired people. The prototype can be used by them to detect and maneuver obstacles such as
 211 staircases and zebra-crossings. A thorough overview of the fundamentals of computer vision was
 212 presented, including the conclusions of historical research and newly proposed techniques. The
 213 shortcomings of various image processing techniques were discussed, and benefits of using one
 214 algorithm over the other were elaborated.

215 REFERENCES

- 216 [1] Blasch B, Wiener W, Welsh R. "Foundations of Orientation and Mobility". 2 AFB Press;
217 1997.
- 218 [2] Boser, B. E.; Guyon, I. M.; Vapnik, V. N. (1992). "A training algorithm for optimal margin
219 classifiers". *Proceedings of the fifth annual workshop on Computational learning theory – COLT*
- 220 [3] P. Strumillo, "Electronic interfaces aiding the visually impaired in environmental access,
221 mobility and navigation," in *Conference on Human System Interactions*, 2010, pp. 17–24.
- 222 [4] M. A. Hersh and M. A. Johnson, *Assistive technology for visually impaired and blind people*.
223 Springer, 2008.
- 224 [5] Guide Dogs NSW ACT, "Guide dog training," Tech. Rep., 2014. [Online]. Available:
225 <http://www.guidedogs.com.au/guide-dogs/guide-dog-training>
- 226 [6] M. S. Uddin and T. Shioyama, "Bipolarity and projective invariant-based zebra-crossing
227 detection for the visually impaired," in *IEEE Computer Society Conference on Computer Vision*
228 *and Pattern Recognition Workshops*, 2005, pp. 22–30
- 229 [7] S. Se, "Zebra-crossing detection for the partially sighted," in *IEEE Conference on Computer*
230 *Vision and Pattern Recognition*, 2000, pp. 211–217.
- 231 [8] V. Ivanchenko, J. Coughlan, and S. Huiying, "Detecting and locating crosswalks using a
232 camera phone," in *IEEE Computer Society Conference on Computer Vision and Pattern*
233 *Recognition Workshops*, 2008, pp. 1–8.
- 234 [9] D. Yuan and R. Manduchi, "Dynamic environment exploration using a virtual white cane", in
235 *IEEE Computer Society Conference on Computer Vision and Pattern Recognition*, 2005.