

Detection of Rail Switch Passages Through Image Processing on Railway Line and Use of Condition-Monitoring Approach

Mehmet Karakose *, Orhan Yaman*, Erhan Akin*

*Computer Engineering Department

Firat University Elazig, Turkey

{mkarakose, orhanyaman, eakin }@firat.edu.tr

Abstract— Today, railway transportation is one of the transport modes commonly used. Compared to other transport modes, railway traffic is highly critical. Multiple railway vehicles run constantly on one or two lines. Rail switch passages are used to prevent locomotives from colliding with one another and avoid traffic disruptions. Through switch passages, locomotives pass from one line to another. Friction between rail and wheels on switch passages is considerably high. This friction leads to failures on switch passages. Unless these failures are diagnosed early and remedied, significant accidents emerge.

In this study, a new approach based on image processing has been presented for detection of rail switch passages on railway lines. A test vehicle has been created in order to test the proposed approach and apply it on a real-time system. Railway line is monitored by digital cameras fixed on this test vehicle. Image-processing approach is developed on the real-time images captured from the railway line and the switch passages on the line are detected. The image-processing approach consists of three main parts including pre-processing, feature extraction and processing of the features obtained. At the pre-processing stage, the basic image processing methods are used. At the feature extraction stage, Canny edge extraction algorithm is used and hence the edges in the image are detected. Hough transform method is used at the stage of processing of the extracted features. Following Hough transform stage, straight lines and angles of these lines are obtained on the image. Taking into account the angle of each straight line, the junction points of the lines are calculated. Thus, rail switch passage and switch types are detected. The proposed image-processing approach is highly fast and real time-based. Compared to the existing studies in the literature, it is seen that the proposed method gives fast and successful results. This study intends to diagnose the failures on switch passages early and prevent potential accidents.

Keywords— Railway, Condition Monitoring, Fault Detection, Image Processing, Railroad Switches.

Railway transportation is commonly used in the freight and passenger transport. Railway vehicles provide transportation using a particular rail line [1,2]. These vehicles are preferred by people because they can carry quite heavy loads and large number of passengers. It is very important for rails to be robust because railway vehicles provide transportation on the rails [3]. The railway lines are the most critical components

for this transportation. The switch has been developed for more than one train can use the same railway line. Trains can change their direction, give way to another train or proceed to the standby points at the station due to the switch system. The development of the switch system has made the railway line quite utilizable. Therefore, the switch system has a great importance for the railway transportation. Excessive friction and wear occur during switching [4]. Such undesirable situations lead to major wear on the rail in time. Major accidents may occur in the event of failure in the early diagnosis of these wears. The accidents that may occur can cause loss of life and injury for passengers. However, the railway line hinders the traffic for a while during the accident, and major financial losses are formed. Today, the railway line is inspected at regular intervals, and the faults which may arise are diagnosed early. There are many literature studies related to switch system. Johansson et al. [5] conducted a study for the detection of the corrosion faults caused by friction and wear during switching. In the study, Switches & Crossings components are monitored. They modeled the contact of the wheel with the rail by simulation during switching. As a result of modeling, observation was made according to the loads that may occur on the rails. Bocciolone et al.[6] proposed a signal processing-based method for the detection of the faults that occur on the rail surface. In the proposed method, the vibration signals made by the train while going on the railway line were used. Fault detection is performed by improving the signal processing methods through vibration signals. Palsson [7] completed a thesis study on optimization in the Switches and Crossing sections on the railways. In this thesis study, Switches and Crossing components were monitored and the faults that occurred in these components were examined. The changes in the Switches and Crossing areas were observed during switch passages. The faults that occur on the rail surface from one rail to another rail surface. In the literature, many studies have been conducted for the detection of the faults occurring on the rail surface and in its components. Quinyong et al. [8] an image processing based contactless method is proposed and detected faults on rails in real-time. Rail surface is extracted by using railway images. Surface faults are determined by contrast stretching of the obtained surface. This method, is used only railway surface to detect faults. Limin et al. [9],

proposed a method that detects faults on the surface by using machine vision techniques. The roughness and cracks are determined with image processing algorithms. Edge detection and feature extraction methods are used.

Ying et al. [10] proposed an automatic monitoring approach to detect and evaluate the railway components. The proposed method is working on a train with 16km/h speed. Ray component detection is carried out by applying Sobel edge detection and Hough transform methods. Status and positions of the components are observed and analysis of them is carried out. The general architecture of the proposed railway inspection system is presented in Figure 1.

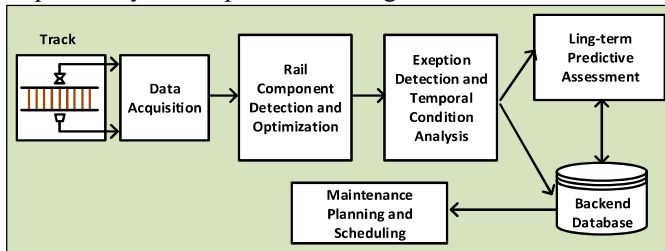


Fig. 1 General architecture of the railway inspection system proposed in the literature [10]

In this study, an image processing-based method was proposed for the detection of the crossing points on the railways. Images were taken for different types of switches using a camera fixed on the train. The switch crossings were determined using image processing algorithms on the images taken.

II. TYPES OF RAILWAY SWITCHES

The apparatus for changing the way that allow railway vehicles moving on the rail to pass from one rail to another rail are called "switch". In switch systems, the change of way of railway vehicles is carried out by ensuring the movement of the Switch part of the junctions with human or electric power [11]. Each switch has a right track and a diverted track. The sleepers of the switches on conventional lines are wooden. The radius of the curve located on the diverted track that provides transition to secondary roads is $R=300$ m, the maximum length standard of single turnout is 34.20 m. The sleepers of the switches on high-speed trains are concrete, the radius of the curve located on the diverted track that provides transition to secondary roads is $R=1500$ m, the maximum length standard of single turnout is 72 m.

Railways have switch crossings for the control of the traffic and a line to be used by more than one trains. A train moving on a line can easily switch to the other line due to switch crossings [12]. An example of a switch crossing system and components is given in Figure 2.

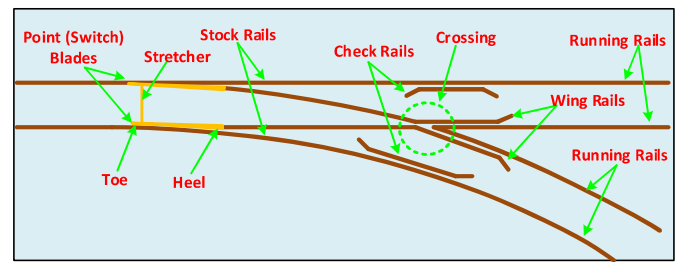


Fig. 2 Example of a switch crossing system and components

As it is seen in Figure 2, the switch crossing system has many components. The components of Switch Blades, Toe, Stretcher and Heel serve as a switching. The components of Crossing, Check Rails and Wing Rails allow train to switch to another rail in a secure way. The components of Check Rails and Wing Rails are the most important components that prevent the derailment of the train. The switch crossings are very important on the railway line. Major faults may occur as a result of the failures of the components on the switch crossings. Therefore, the railway line especially the switch crossings should consistently be inspected. It is required to make the necessary maintenance even considering the small faults. On switch crossings, to which direction the train will go is determined using the components of Switch Blades, Toe, Stretcher and Heel. The changes that occur on the switch crossings when a train goes straight or to the right direction on an example railway line are seen in Figure 3.

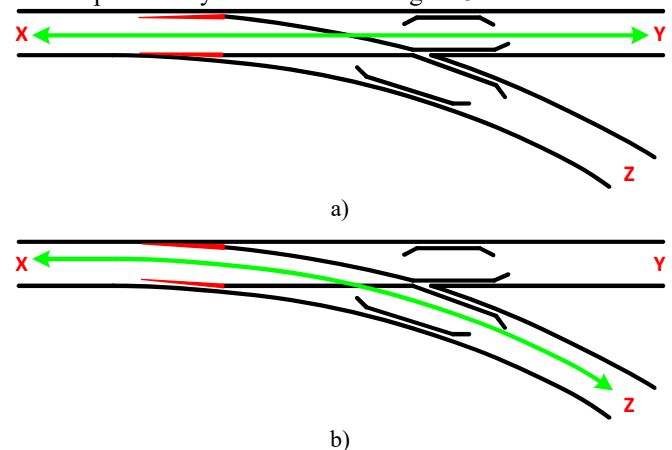


Fig. 3 The changes that occur on the switch crossings when a train goes straight or to the right direction on the railway line a) The state of the train going straight b) The state of the train turning right

In the sample images given in Figure 3, a train coming from X direction in Figure 3.a goes straight ahead, namely to the Y direction. In Figure 3.b, a train coming from X direction turns right, namely to the Z direction. There are many types of switch crossings which are commonly used on the railways. These types of switches vary depending on the number and position of the railway line. The types of switch crossings and their sample images are given in Figure 4.

The sample images of the switch crossing types given in Figure 4 are presented in Figure 5.

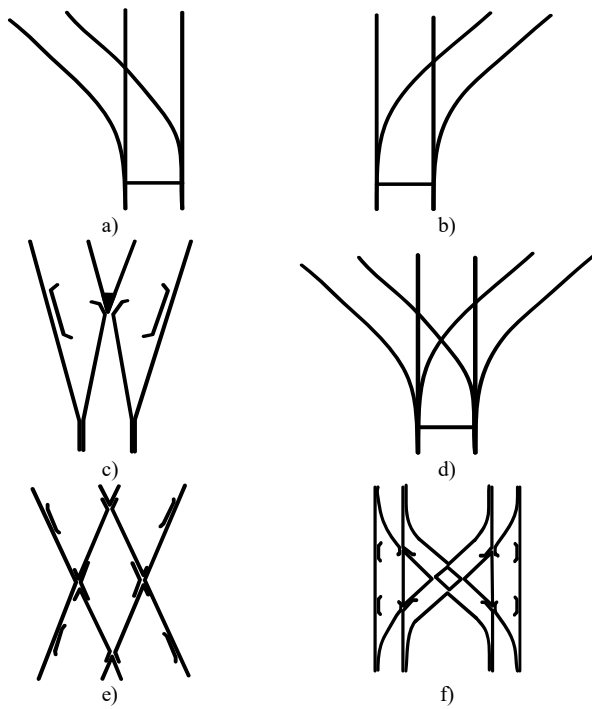


Fig. 4 Railway switch types a) Single left switch b) Single right switch c) Symmetric switch d) Compound switch e) Cross switch (full cross) f) crossover [13-15]

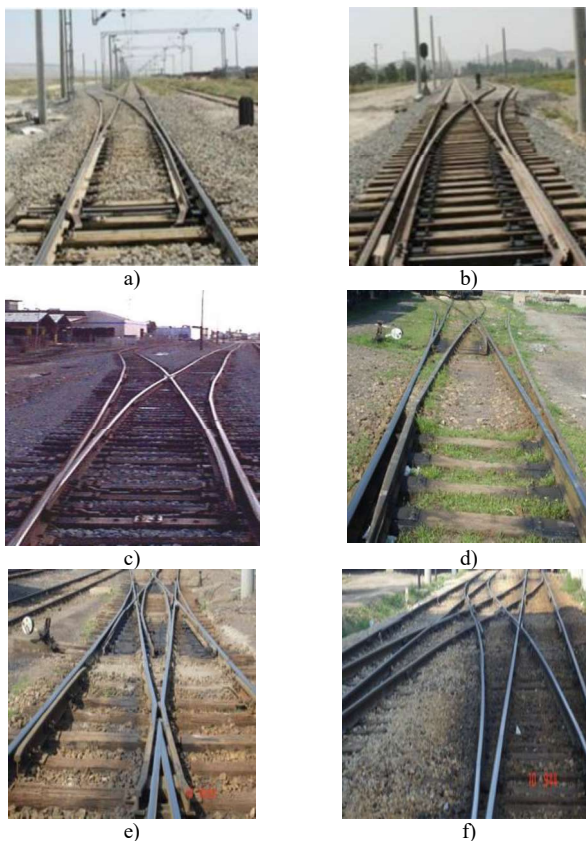


Fig. 5 Sample images of the railway switch types a) Single left switch b) Single right switch c) Symmetric switch d) Compound switch e) Cross switch (full cross) f) crossover [13-15]

III. THE METHOD PROPOSED FOR SWITCH DETECTION

In this study, an image processing-based method was developed for detecting the switch crossings on the railway line. In the proposed method, feature extraction was performed by performing pre-processing on the image. The lines on the image are obtained as a result of feature extraction. The intersection points are calculated by considering the start and end positions and slopes of the lines obtained. The calculated intersection points are called switch crossing zone. The flow chart showing the main stages of the proposed method is presented in Figure 6.

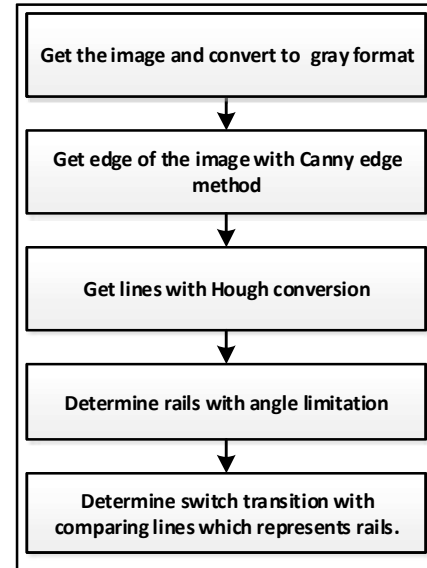


Fig. 6 Flow chart of the proposed method

In the proposed method, images are taken from a camera placed on the locomotive. The colorful images taken from the camera are primarily converted into gray format.

Canny edge extraction is performed using the gray image obtained. The edges on an image are defined as the sudden change of the pixel values on the edge zones. The edge extraction methods are very important for feature extraction in image processing applications [16,17]. Canny edge extraction algorithm is an edge extraction algorithm which is realized as multi-staged. The general steps of Canny edge extraction algorithm are as follows;

- Performing smoothing by applying Gauss filter to the image
- Performing edge extraction in the X and Y directions
- Calculation of gradient magnitude and angle
- Elimination of undesirable details and obtaining image

In Equation 1, $f(i, j)$ represents the gray level image matrix, $R(i, j)$ represents the red image matrix, $G(i, j)$ represents the green image matrix and $B(i, j)$ represents blue image matrix. The image smoothing process is performed by applying gauss filter to the image in gray format obtained. A gauss mask in 5x5 size is used in the Gauss filter.

After the gauss masking process, edge extraction is performed by using the canny masking matrix of the image from X and Y direction. The image is formed according to certain angle limits by calculating the gradient magnitude on the image matrices obtained.

The edge extraction process is achieved more successfully because Canny edge extraction algorithm uses the Gauss filter before performing edge extraction on the image. Canny edge extraction algorithm gives better results compared to other edge extraction algorithms.

Hough transform is used to obtain the lines on the image after the edge extraction process [18]. Hough transform is generally used in detecting geometric models on the image in combination with the edge extraction methods. Hough transform is applied onto the images, the edge information of which are obtained. The method transforms the problem of finding a shape into a problem of finding a density by transferring the information in the image space to the parameter space [19-21]. Hough transform does not require that the points on the image are side-by-side or connected to detect an image. The pseudo code of Hough transform used in the proposed method is given in Figure 7.

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Image ← Edge detection image in binary format
Column ← Width of Image
Row ← Height of Image
for i=1:Column
    for j=1:Row
        if (Image(i,j) is an edge detected image)
            for θ=0:θmax
                r = Column*cos(θ) + Row*sin(θ)
                r = round(r)
                H(θ,r) = H(θ,r)+1
            end
        end
    end
end

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Fig. 7 Pseudo code for Hough transform [20]

A binary-based image obtained as a result of edge extraction is primarily used in the pseudo code of Hough transform given in Figure 7. The straight lines are expressed as in equation 1.

$$y = mx + n \quad (1)$$

This equation has two parameters to specify any straight right. These parameters are slope (m) and junction point (n). A point in the straight lines parameter space is shown as (m, n). The straight lines are transformed into coordinate system and shown in equation 2.

$$r = x \cos(\theta) + y \sin(\theta) \quad (2)$$

In this equation, the nearest vector from the origin is represented by a pair of (r, θ). All lines with angles ranging between 60 and 150 on the detected lines are taken into account. The junction points of the lines are calculated using

the angle values of the lines obtained and the start and end points of the line. The method of the slope of a line two points of which are known was used in obtaining these lines. The lines representing the rails are detected by considering the slope ranges of the lines on the railway image. The slope of a line two points of which are known is given in equation 3.

$$m = \frac{B_y - A_y}{B_x - A_x} \quad (3)$$

In equation 3, m represents the slope, (A_x, A_y) represents the location information of the A point and (B_x, B_y) represents the location information of the B point. The switch crossings on the railway line are detected according to the slope value obtained by applying this equation on the gray image. Two points of each line are known. These two lines intersect and give the switch crossing zone. For the calculation of the crossing point, the required calculations were made for the junction point of the line two points of which are known. First of all, the points of the line taken from the normal image are A, B, C and D . The junction point of these two lines is E . The junction point of two lines two points of which are given is shown in the analytical plane as in Figure 8.

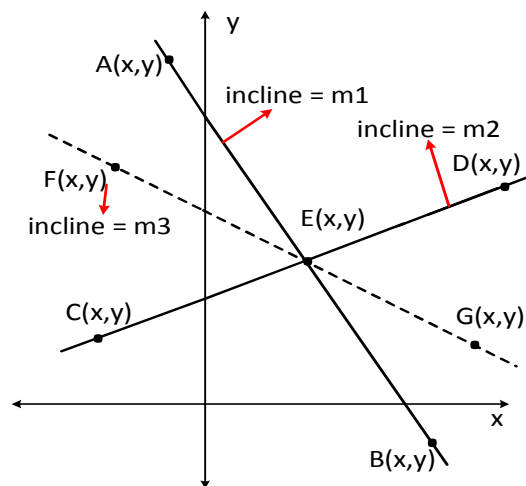


Fig. 8 Representation of the junction point of two lines two points of which are known in the analytical plane

In order to calculate the values of the E point given in Figure 8, another line passing from E, F and G point is drawn. The equation of the line the slope of which is m_3 is given in equation 4.

$$m_3 = \frac{G_y - F_y}{G_x - F_x} \quad (4)$$

By calculating the location information of F and G given in equation 4, the final state of the equation 4 is given in equation 5.

$$m_3 = \frac{((D_y - C_y) * (C_x - A_x)) - ((C_y - A_y) * (D_x - C_x))}{((D_y - C_y) * (B_x - A_x)) - ((B_y - A_y) * (D_x - C_x))} \quad (5)$$

By using m_3 value given in equation 5, E_x and E_y values are calculated in equation 6.

$$E_x = A_x + (B_x - A_x) * m_3$$

$$E_y = A_y + (B_y - A_y) * m_3 \quad (6)$$

The switch crossing zone is detected on the image using E_x and E_y values obtained in equation 6.

IV. EXPERIMENTAL RESULTS

In this study, an image taking environment was created to take sample images on the railway line. The images of the railway are taken by the camera installed on the train as in Figure 9.

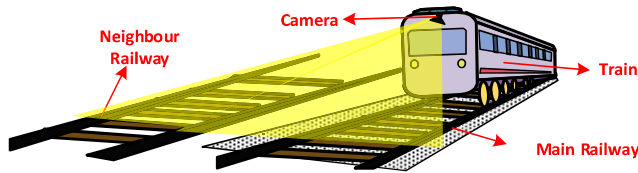


Fig. 9 The test vehicle used to take image

In the proposed method, the switch crossings were detected by taking images in different situations on the railway line. The dimensions of the images taken are 640 x 480 pixels. The color image was firstly transformed into gray image, and then edge extraction was performed. The lines are obtained by Hough transform on the image obtained. The images used during realization of the proposed method are given in Figure 10. The results were obtained by performing edge extraction on the image given in Figure 10. These results are presented in Figure 11.

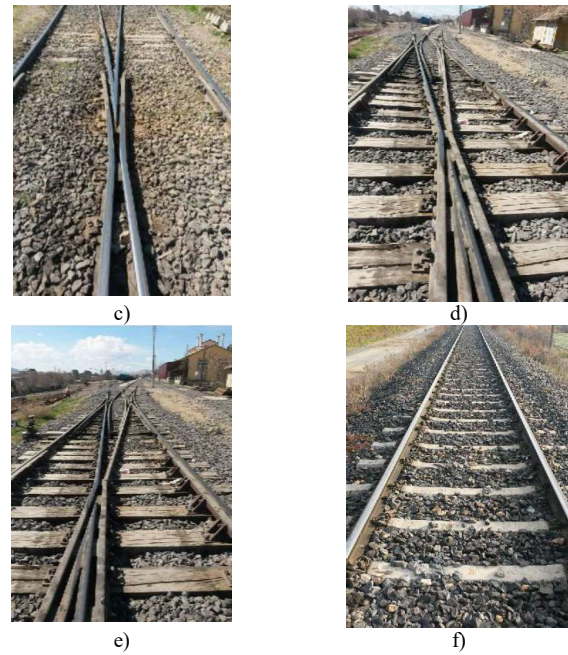
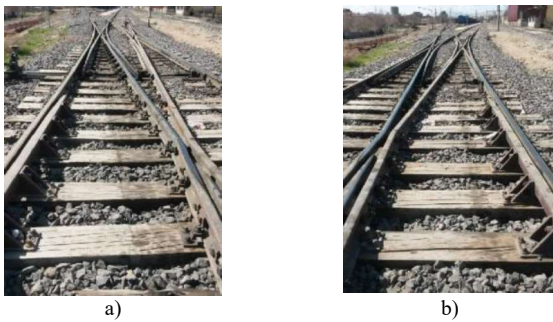


Fig. 10 The images belonging to different types of switch crossings taken from the railway

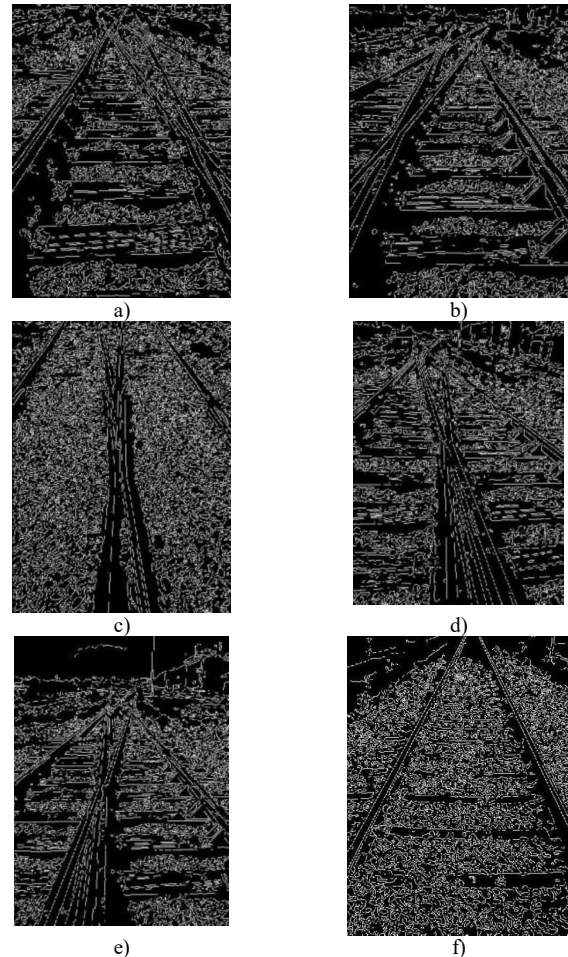


Fig. 11 Performing feature extraction by applying edge extraction on the image

The edge extraction results of different images are presented in Figure 11. After the edge extraction process, Hough transform was used in order to detect the lines on the image. The lines were detected by taking into account the angle value and the positions of the line during Hough transform. The switch crossings are detected by improving a method on the lines detected. The result of detecting the switch crossings on the sample images is presented in Figure 12.

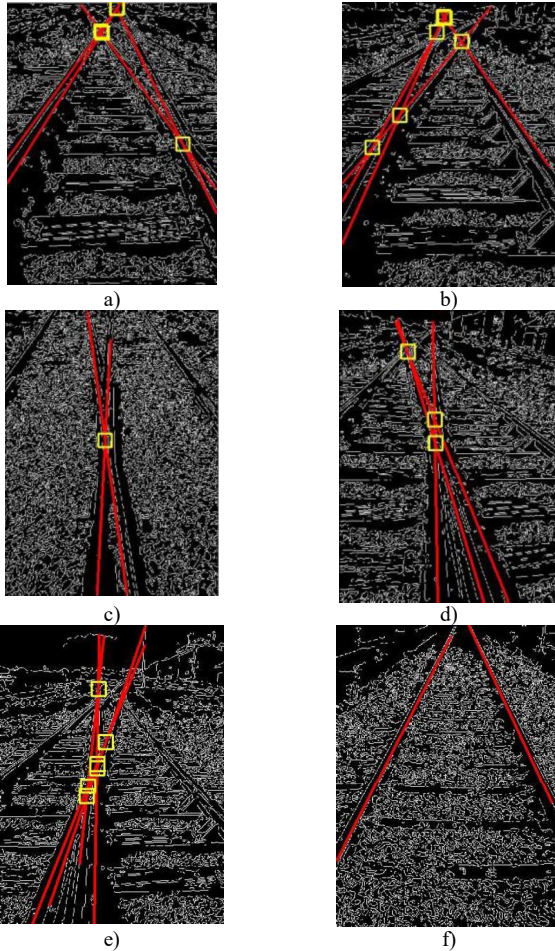


Fig. 12 Detection of the switch crossings on the railway line by Hough transform

The switch crossings in Figure 12 were detected. The image of a normal railway without switch crossing is presented in Figure 12.f. No switch crossing was detected there. The switch crossings were detected successfully in other images. Within the context of this study, many images were obtained for the test procedure. There are 100 switch images on these images. These switch images belong to different types. The numbers of successful and unsuccessful images when the proposed method was applied on these images are given in Table 1.

TABLE I
THE IMPLEMENTATION OF THE PROPOSED METHOD FOR THE IMAGES

| Railway switch types | Number of Test Image | Number of Successful Image | Number of Failed Image | Success rate (%) |
|----------------------|----------------------|----------------------------|------------------------|------------------|
| Single left switch | 29 | 24 | 5 | 82,7 |
| Single right switch | 26 | 23 | 3 | 88,4 |
| Symmetric switch | 17 | 11 | 6 | 64,7 |
| Compound switch | 13 | 9 | 4 | 69,2 |
| Cross switch | 9 | 6 | 3 | 66,6 |
| Crossover | 6 | 4 | 2 | 66,6 |

As it is seen in Table 1, the proposed method is highly successful in the images belonging to the types of single left and right switch. The success ratios of other types of switch are not very good because they are a little more complicated and due to the clarity of the image.

V. CONSLUSIONS

An image processing based method for the detection of railway switch crossings was proposed in this study. In the proposed method, the switch crossings were detected by image processing methods using railway images. Canny edge extraction and Hough transform were used as the image processing methods. When the studies in the literature are analyzed, it is seen that there is not a study on the detection of switch crossings and condition monitoring by using image processing based methods. It is seen that the proposed method is successful when its results are taken into account. However, different methods will be used in future studies to make study more qualified and successful. In addition, the detection of the faults occurring in the components on the switch crossings and the detection of the rail surfaces on the switch crossings are aimed in future studies.

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