A Robotic Assistance Machine Vision Technique for an Effective Inspection and Analysis

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Article Info ABSTRACT Article history: An Inspection is a study of methods and techniques that can be suitably employed in protections. In this paper, a new activity is proposed

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employed in practical applications. In this paper, a new activity is proposed and analysis framework to facilitate the inspection of an object using machine vision technique in which maximum efficiency can be achieved. By using LABVIEW software and vision builder software the quality of output images such as image compression, image restoration and multimedia streaming are achieved successfully. So the proposed design makes use of various image processing functions like special filters and classifiers to compute the optimum results. Using smart camera in the inspection system the static as well as the dynamic object is captured in fraction of seconds without any blurs; as a result the optimum image quality without any distortion is obtained for better analysis. The proposed system is very precise, accurate and flexible with reasonable development cost compared to other model. With the aid of an Industrial robotic system with simulation software the object is replaced immediately when the same is rejected by the machine vision model. Apart from this, the proposed model can be implemented for any type of Automation work

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1. INTRODUCTION

It is an obligatory step to evaluate the quality of output images in many image processing applications such as image acquisition, compression, restoration, transmission, etc. Since human beings are the ultimate observers of the processed images and thus the judges of image quality, it is highly desired to develop automatic approaches that can predict perceptual image quality consistently with human subjective evaluation.

Machine vision is a branch of engineering that uses computer vision in the context of manufacturing, where the images analysis is done to extract data for controlling a processor activity. Machine vision processes are targeted at recognizing the actual objects in an image and assigning properties to those objects-understanding what they mean. Engineers are adding Machine vision systems to a number of industrial applications to reduce costs, increase throughput, and improve customer satisfaction. All machine vision systems include a combination of hardware and software to acquire and process images, usually resulting in a response from a secondary system connected to the inspection system. There are two possible designs for this reference architecture. One includes a low-cost, rugged, embedded solution while the other makes use of the power of a computer to acquire and process images at a higher rate and resolution. Vision systems can be used to precisely measure number of variables such as length, angle, position, orientation, color and so on. The main advantage of a vision measuring system in these applications is its non-contact feature, which is important in cases when it is difficult to implement contact measurement.

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This paper proposed the devices and method to develop an effective inspection system. The system consists of Smart Camera 1722, a motion platform, Lab view Window and NI vision development module. In this proposed model the investigation of the circular linearity by using the NI Vision system which consist of the following Parts such as

[D] Image processor

[C] Smart Camera

 Illumination
 Smart camera

 [NI-1722]
 Image processor

 Object to be inspected
 Robot

 Controller

Figure 1. Schematic Block diagram of Object Inspection System

A. Illumination:

Lighting includes use of both artificial light sources such as lamps and natural illumination of interiors from daylight. Artificial lighting is most commonly provided today by electric lights. When the information from the samples is accumulated and analyzed, with respect to the specific sample and inspection requirements, we can achieve the primary goal of NI vision lighting analysis - to provide sample appropriate lighting that meets three acceptance criteria consistently:

- Maximize the contrast on those features of interest
- ✤ Minimize the contrast elsewhere
- Provide for a measure of robustness.
- B. Object:

The part which is going to be inspected in terms of circular linearity. In this particular paper the cylindrical bar is placed for investigating its circular linearity.

C. Smart Camera 1722

The selection of camera is heavily dependent on our application. If select an appropriate camera, lens and lightning setup, the efforts can then be focused on developing our solution, rather than wrestling with poor image data and saves processing time at execution. Here NI smart camera1722 is selected and having resolution of 1280x1024 with 7.4x7.4 µm pixel size.

D. Image processor

It is used for processing the acquired image from the smart camera for results with actual predefined parameters.

E. Controller (Robotic):

The Robotic controller gives the command signal to the Robot after successful analysis by means of circularity matching with predefined parameters; as a result the defective objects can be replaced form the test bench. The Block diagram of Object Inspection System presented in Figure 1.

2. ALGORITHM

The algorithm comprises the various analyzing and processing functions illustrated below.

[A] Illumination

[B] Object

[E] Controller (Robotic)

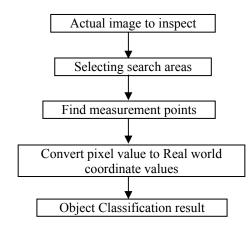


Figure 2. Algorithm chart for Inspection of an object

The image is captured by the NI SMART CAMERA (NI-1722). Vision Assistant offers three types of acquisitions: snap, grab, and sequence. A snap acquires and displays a single image. The grab acquires and displays images in a continuous mode at maximum rate, which is useful when you need to focus the camera. The sequence acquires images according to settings that specify in the Sequence tab of the Acquisition Interface window and sends the images to the Image Browser.



Figure 3. Actual image captured by Smart Camera

The origin image's size is 1360×1024 , it is so big that it will waste long time to process. So it is necessary to resize the origin image by scale of 0.2. But the result must multiply the same scale of 5 correspondingly to recover its real position in origin image.

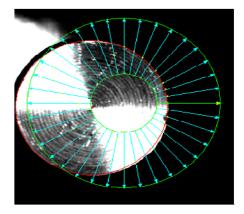


Figure 4. Selecting the desired mark points

Classification identifies an unknown sample by comparing a set of its significant features to a set of features that conceptually represent classes of known samples. The object classifier uses feature vectors to identify samples based on their shape. Color classifier uses color features to identify samples based on their color. Classification involves two phases: training and classifying. Training is a phase during which we teach the machine vision software the types of samples want to classify during the classifying phase. It can train any number of samples to create a set of classes, which, later compare to unknown samples during the classifying phase and stored in the classes in a classifier file. Training might be a one-time process, or it might be an incremental process repeat to add new samples to existing classes or to create several classes, thus broadening the scope of samples to classify. The classifying phase classifies a sample according to how similar the sample features are to the same features of the trained samples. The need to classify is common in many machine vision applications. Typical applications involving classification include the following:

- Sorting
- Inspection

Here the geometric matching algorithm is used to detect the mark of circle. Geometric matching is an important tool for machine vision applications; it must work reliably under various, sometimes harsh, conditions. In automated machine vision applications – especially those incorporated into manufacturing process – the visual appearance of materials or components under inspection can change because of factors such as varying part orientation, scale, and lighting. The geometric matching tool must maintain its ability to locate the template patterns despite these changes.

The geometric matching process consists of two stages: learning and matching. During the learning stage, the geometric features from the template image. The algorithm organizes and stores these features and the spatial relationships between these features in a manner that facilitates faster searching in the inspection image. During the matching stage, the geometric matching algorithm extracts geometric feature from the inspection image that correspond to the features in the template image. Then, the algorithm finds matches by locating regions in the inspection image where features align themselves in spatial patterns similar to the spatial patterns of the features in the templates.

In Lab view Windows/CVI, the function imaq Detect Circles based geometric matching is used to perform the task. In the function, the threshold which specifies the minimum contrast a seed point must have in order to begin a curve must be set carefully. The inspection result is listed in Table I. The coordinate origin is in top left corner of the inspected image by default. There exits a little difference corresponding to different resize scale. The reason lies in the interpolation method when resize the origin image using Resample function in Lab View Windows. But it doesn't affect the inspection result if we convert the coordinate to the benchmark.

Scale	0.6	0.5	0.4	0.35	0.3	0.2
X (pixel)	883.33	880	882.5	887.14	833.33	815
Y (pixel)	685	688	652.5	652.86	656.67	660
Time (ms)	22.2	16.9	16.8	12.2	11.8	9.2

Table 1. Mark inspection result

2.1 Coordinate Conversion

Before inspection, the system must be able to relate regions of particular images to be inspected. However, in an inline inspection system, when an object moving into the inspection position, there are some minor positional or orientation errors because of the mechanical factor and the interpolation factor. To eliminate the deviation, it is necessary to align the object prior to inspection. This paper presents an algorithm to rectify the error from the above factor.

As shown in Figure 5, P and Q are marks in database with coordinate (Xp, Yp) and (Xq, Yq), P' and Q' are marks inspected from a sample object with coordinate (Xp', Yp') and (Xq', Yq'). The midpoint of P and Q is O (Xo, Yo), and the midpoint of P' and Q' is O' (Xo', Yo'). Their coordinate can be calculated as the following equations.

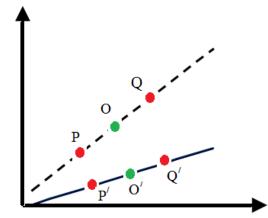


Figure.5 : Marks in an inspected object and reference data in database.

$$X_{0} = \frac{X_{p} + X_{q}}{2}$$
[1]

$$Y_{0} = \frac{Y_{p} + Y_{q}}{2}$$
[2]

and
$$X'_{0} = \frac{X'_{p} + X'_{q}}{2}$$
 [3]

$$Y'_{0} = \frac{Y'_{p} + Y'_{q}}{2}$$
[4]

$$P_{X} = \frac{X_{0}}{X_{0}^{\prime}} = \frac{X_{p} + X_{q}}{X_{p}^{\prime} + X_{q}^{\prime}}$$
[5]

$$P_{Y} = \frac{Y_{O}}{Y'_{O}} = \frac{Y_{p} + Y_{q}}{Y'_{p} + Y'_{q}}$$
[6]

According to above equation, we can convert the coordinate system of the real object in to the database coordinate system. A pixel point in an object with coordinate (P', Q') can convert to database coordinate point (P, Q) as following equation

$$P = P_X P', Q = P_Y Q'$$
^[7]

2.2 Pattern Matching and Inspect the Wrong Parts

Pattern matching quickly locates regions of a grayscale image that match a known reference pattern, also referred to as a model or template. When using pattern matching, first create a template that represents the objects for which you are searching. Pattern matching application then searches for instance of the template in each acquired image, calculating a score for each match. This score relates how closely the template resembles the located matches. Pattern matching finds template matches regardless of lighting variation, blur, noise, and geometric transformation such as shifting, rotation, or scaling of the template. NI pattern matching techniques include normalized cross-correlation, pyramidal matching, scale-and rotation-invariant matching, and image understanding. Normalized cross-correlation is the most common method for building a template in an image. The following is the basic concept of correlation:

R(p,q) of size $k \times l$ within an image f(p,q) of size $m \times m$, where $k \le m$ and $l \le n$. The correlation between R(p,q) and f(p,q) at point (i, j) is given by

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$$C(i,j) = \sum_{p=0}^{l-1} \sum_{q=0}^{k-1} R(p,q) f(p+i,q+j)$$
[8]

Where i = 0, 1...m-1, j = 0, 1, ..., n-1, and the summation is taken over the region in the image where R and f overlap. Basic correlation is very sensitive to amplitude changes in the image, such as intensity, and in the template.

For example, if the intensity of the image f is doubled. So are the values of C. We can overcome sensitivity by computing the normalized correlation coefficient, which is defined as

$$S(i,j) = \frac{\sigma(i,j)}{\sqrt{D(R)}\sqrt{D[f(i,j)]}}$$
[9]

$$\sigma(i,j) = \sum_{p=0}^{l-1} \sum_{q=0}^{k-1} (R(p,q) - \overline{R}) (f(p+i,q+j) - \overline{f}(i,j)$$
[10]

$$D(R) = \sum_{p=0}^{l-1} \sum_{q=0}^{k-1} (R(p,q) - \overline{R})$$
[11]

$$D[f(i,j)] = \sum_{p=0}^{l-1} \sum_{q=0}^{k-1} (f(p+i,q+j) - f(i,j))^2$$
[12]

where R (calculated only once) is the average intensity value of the pixels in the template R. The variable f is the average value of f in the region coincident with the current location of R. The value of S(i, j) lies in the range -1 to 1 and is independent of scale changes in the intensity values of f and R. Normalized cross-correlation is a good technique for finding patterns in an image when the patterns in the image are not scaled or rotated. Typically, cross-correlation can detect patterns of the same size up to a rotation of 5° to 10° . Scale invariant matching adds a significant amount of computation to matching process. To resolve the problem, NI improves the computation time of pattern matching by reducing the size of the image and the template by pyramidal matching. In pyramidal matching, both the image and the template are sampled to smaller spatial resolutions. Because the image is smaller, matching areas in the original image.

3. RESULTS AND ANALYSIS

We performed the proposed algorithm and method with Lab View Windows/CVI to detect the quality of an circular object. Figures 6-8 are the software user interface which are designed to detect the different edge position. All the aforesaid techniques are followed by using the NI vision in which the investigation of the circularity of an object as follows

👷 NI Vision Builder AI - Configuration - Untitled Inspection 1 * - (Remote Target: NI-1722-01774143 - 169.254.66.100)	X
File Edit View Operate Tools Target Help	
	2
	Start
State: Inspect	interest.
Sale Sale Sale Sale Sale Sale Sale Sale	Geometric Matching: Locates grayscale features based on edge information in the entire image or a region of interest.
PASS PASS <th< td=""><td>Set Coordinate System: Builds a coordinate system based on the location and orientation of a reference feature.</td></th<>	Set Coordinate System: Builds a coordinate system based on the location and orientation of a reference feature.
Acquere image (shart mier image 1 intestriou image 1 catorate image 1 create keglon or int mit Groude roge 1 betect objects 1 Frame Index = 23689 Threshold = 128 Diameter = 25.81 mm	Detect Objects: Locates objects of homogenous intensity (particle analysis),

Figure 6. Simulated object by the software

Find Circular Edge Setup						
Main	Settings	Advanced	Limits			
Edge Strength Profile Search Line Index 1 95						
Edge Po	ints Found Strength	STR (dB)	TNR (dB)	SNR (dB)		
1	43.33	6.72	3.31	10.02	Ê.	
2	33.33	4.44	8.33	12.77		
3	20.67	0.28	8.33	8.61		
4	24.00	1.58	-0.96	0.63		
5	32.00	4.08	11.48	15.56		
6	26.67	2.50	3.97	6.47		
7	28.67	3.13	6.94	10.06		
8	21.67	0.70	13.28	13.98	Ŧ	

Figure 7. Circular edge parameters setting points

Main	Settings	Advanced	Limits	,				
Pass Inspection if								
-	Minimum Diameter (pix)				26.00	٢		
Maximum Diameter (pix)					46.00			
	Maximum Deviation (Pixels)							
Step F	Results							
Name	e		Value		Unit		*	
Cent	er X Position	l i i i i i i i i i i i i i i i i i i i	115.82		pix			
Cent	Center Y Position				pix			
Radi	Radius				pix			
Diam	eter		227.10		pix			
Devia	ation		6.29		pix			
Strer	ngth		61.36					
Aver	age SNR		58.73		dB			
							Ŧ	
Step St	tatus PA	ASS	ОК) Ca	ance		

Figure 8. Circular edge parameters setting points

4. CONCLUSION

The computational speed of image processing has been improved using Machine Vision builder. A new system based on machine vision helps to calculate the results with precise and accurate values. Object Classification is an effective technique to classify or inspect the Circular linearity automatically and give results to match with the previously saved template. Filters help to improve the quality of an image. Image quality is basically depends upon the illumination conditions which produces noise in the image.

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