

## Automatic Reading System of Scale Values on Webcam-Based Measuring Vessel

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### ABSTRACT

**Keywords:** reading error, automatic reading, measuring vessel, image processing, webcam camera.

The fuel measurement pump is an installation used to measure the quantity of liquid fuel (BBM) to be sold or delivered to consumers. In the transaction of buying and selling fuel at gas stations, the fuel measurement pump plays a crucial role. Therefore, this measuring device needs to be calibrated or re-verified, and supervision is carried out using a standard measuring vessel with capacities of 10 L or 20 L. Errors in reading the measuring vessel can occur due to differences in individual visual acuity. To address this issue, a prototype for automatic reading of measuring vessels is created using a webcam camera. This prototype utilizes the Raspberry Pi 3 B+ as an image processor and the Logitech C270 webcam as an image capturer. The image processing process involves segmentation steps, transforming the image into a grayscale and then into a binary image. This research involves two different types of fluids in the volume measurement of the measuring vessel, namely distilled water (equates) and petalite. The measurement results with the prototype show an average accuracy level of around 99,23%, an average precision level of about 99,37%, and an average error rate of about 0,77%. For the equated fluid, the accuracy, precision, and error levels are 99,48%; 99,76%; and 0,52%, respectively. Meanwhile, for the petalite fluid, the accuracy, precision, and error levels are 98,98%; 98,80%; and 1,02%, respectively.



### Introduction

Fuel demand in Indonesia continues to increase in line with the growth of motor vehicles and the spread of petrol stations (Putri, Novita, & Maskar, 2022). In fuel transactions, measurement accuracy is an important factor. For this reason, fuel gauge pump measuring instruments must be marked and re-marked as well as periodic supervision (Sinaga, 2017). This activity is usually carried out by volumetric method using a standard measuring vessel (BUS) as the standard. A measuring vessel is a volume-measuring instrument that serves as a reference in testing other volume-measuring instruments (Buana, 2018).

However, differences in the ability to read the BUS scale by the individual eye can result in reading errors (Pertiwi & Setyawan, 2020). Therefore, an automatic reading prototype using the camera sensor is required. Previous research has tried using ultrasonic sensors and capacitive sensors, but it still has limitations (Adriansyah, GM, & Yuliza, 2014).

The study used camera sensors to capture BUS-scale images, which were then processed automatically (Irmayanti & Andini, 2019). This enables accurate automatic readings and generates evidence in the form of images for surveillance activities. The test results show that this prototype is effective in improving the accuracy of BUS measurements at petrol stations (Laoli, 2018). Thus, the use of camera sensor technology can be a solution to improve measurement accuracy in fuel transactions at petrol stations and facilitate supervision by metrology supervisors.

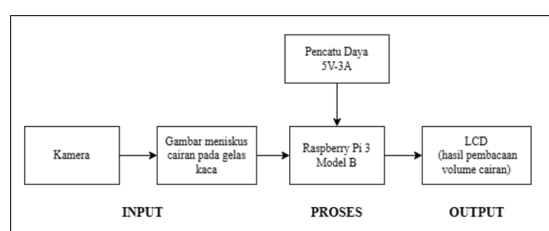


Figure 1. System block diagram

## Research Methods

In the prototype of the reading of the scale value on the measuring vessel, a method was carried out by simulating synthetic images. The synthetic imagery is made to resemble the image of a measuring vessel captured by a webcam. This image is made to resemble the colour of a petalite liquid, which is green.

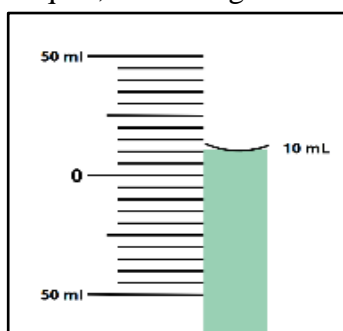


Figure 2. Synthetic imagery

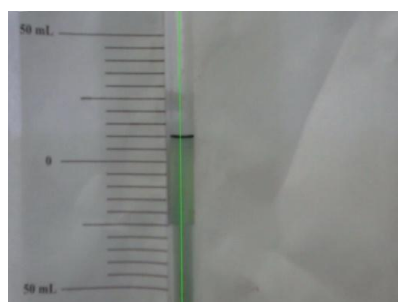


Figure 3. Image of the acquisition using a webcam camera

The image processing techniques used are as follows:

1. Pre-processing, preparing the correct camera position and determining the appropriate ROI (Region of Interest) to be detected by the camera.
2. The segmentation process, changing the RGB image to a grayscale image which is then changed again to a binary image (black and white).
3. Post-processing, image analysis by detecting black pixels using the principle of liner interpolation then calculating what volume the meniscus point is at with equation (1) as follows.

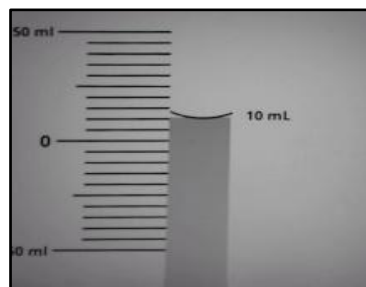
## Results and Discussion

### Simulation Results

In this study, a simulation was carried out using synthetic images as shown in Figure 2 which aims to determine the reliability of the program that has been made. This synthetic image is made to resemble the original image taken using a webcam camera. In the synthetic image that has been made, the ROI image contains at least a liquid meniscus object in the glass of the measuring vessel and scale lines on the glass wall, which is -50 mL to 50 mL (Zahara, 2019).

After that, the synthetic image that has been created is tested using the programming in the prototype to obtain the results of the volume reading on the measuring vessel. In synthetic images, the meniscus line is at a volume of 25 mL. Meanwhile, in prototype programming, it reads 25 mL (Alfonsius et al., 2023). This indicates that the prototype can run well because it has successfully read artificial images. The reading results have a difference of 0 mL.

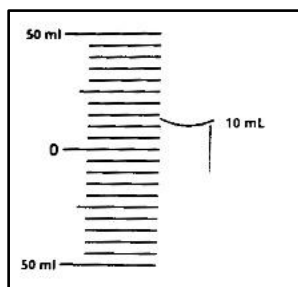
This image processing uses several stages of image processing. First, the segmentation stage by changing the RGB colour image to grayscale or grayscale.



**Figure 4 Segmentation of RGB images to grayscale images**

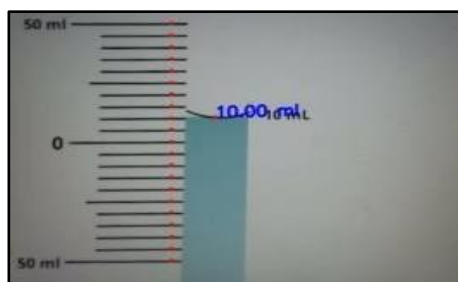
After that, for the image to be at a certain threshold, the adaptive threshold method is carried out. This method is performed to convert a grayscale image into a binary (black-and-white) image taking into account the pixel intensity in about each pixel. This helps in overcoming uneven lighting variations across the imagery (Abdi, 2020). In this process, the grayscale image will be changed to black or white only, where black will have a binary value of 0 while white will have a value of 1. This is shown in the meniscus producing

the colour black. In programming, the detected black colour is considered a black pixel. The detected black pixels are marked with a red dot that has x coordinates of 290.



**Figure 5. Results of segmentation of grayscale images to binary images**

Next, linear interpolation is carried out to determine the volume read on the glass of a standard measuring vessel. This interpolation is a method for estimating the value between 2 previously known points on a straight line of black pixels of scale. The reference used to determine the value of the detected meniscus black pixel is the black pixel on the artificial scale.



**Figure 6. Reading results**

It is known that the value of the px coordinate is 193 pixels. In addition, the start and end of the pixels are 193 and 211 pixels, and then the start and end of the mL values are 10 mL and 5 mL. Thus, the value of the meniscus point in the volume is obtained as follows.

$$Volume = starting\_ml\_value + (ending\_ml\_value - starting\_ml\_value) \times (px\_coordinate - starting\_pixel) / (ending\_pixel - starting\_pixel)$$

$$Volume = 10 \text{ mL} + (5 - 10) \text{ mL} \times (193 - 193) \text{ Px} / (211 - 193) \text{ Px}$$

$$Volume = 10 \text{ mL} - 5 \text{ mL} \times (0) / (18)$$

$$Volume = 10 \text{ mL} - 0 \text{ mL}$$

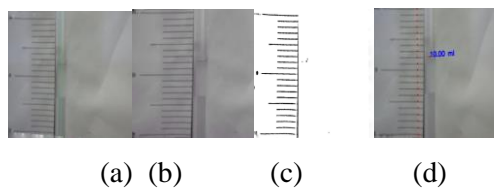
$$Volume = 10 \text{ mL}$$

The simulations that have been carried out using synthetic images, show that the image processing carried out has succeeded in reading the volume of liquid on the measuring vessel. The results of the prototype reading showed the same results as the

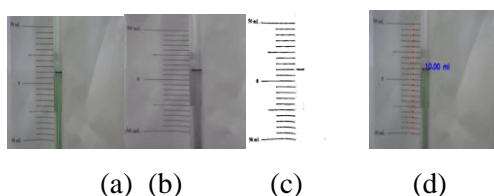
synthetic images that had been made. In addition, synthetic images have also been made at other meniscus points and produced an error range of 0 mL at 11 meniscus points, namely 25 mL to -25 mL with a difference of 5 mL at each point.

### Experiment Results

Synthetic image processing has been successfully carried out. Furthermore, the researcher conducted a direct test with real images using 2 different liquids, namely equates and petalite. Viewed in Figure 7, a real image with liquid equates reads 10 mL. That way, the results read by the prototype are not much different from the manual reading, which is 10 mL.



**Figure 7**  
Aquades (a) RGB imagery (b) Grayscale imagery (c) Binary imagery (d) Imagery



**Figure 8**  
Pertalite (a) RGB image (b) Grayscale image (c) Binary image (d) Image processing results

Viewed in Figure 8, a real image with petalite liquid reads 10 mL. That way, the results read by the prototype are not much different from the manual reading, which is 10 mL.

Conducting experiments using 2 different liquids, will prove the reliability of the prototype in taking volume readings on different liquid colours. In addition, petalite is a type of fuel so it has different properties and densities compared to equates. In the calibration of the measuring vessel, equate liquid is used. That way, in addition to being able to be used for surveillance and re-stamping purposes, this prototype can also be used for calibration purposes.

### Conclusion

In this study, a prototype of an automatic reading system for scale values on a measuring vessel using a webcam camera has been made, where this prototype can automatically read the volume of the measuring vessel on equates and petalite liquids through the integration of prototype-supporting hardware and software to be able to detect meniscus points through image processing with OpenCV Python.

Method validation for the prototype has been carried out by comparing the readings on the prototype with manual readings or references using a nonius scale of a measuring

vessel with an accuracy of 0.5 mL and an artificial line with an accuracy of 0.05 mL. Based on the results of the method validation, the prototype reading has approached the manual reading.

The measurement of the volume of the measuring vessel with the automatic reading prototype has an average accuracy rate of 99.23%; the average precision is 99.28%; and the average error is 0.77%. In equates liquid, the level of accuracy, precision, and error is 99.48%; 99,76%; and 0.52%. Meanwhile, in petalite liquid, the level of accuracy, precision, and error is 98.98%; 98,80%; and 1.02%. Not only used for surveillance activities but the prototype can also be used for calibration of measuring vessels because the prototype can take readings with adequate liquid.

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