

# Level measurement comparison between 3D vision system based on Kinect and ultrasonic industrial sensor

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**Abstract - This paper presents the development of a level control system using 3D machine vision techniques based on a single level-detector sensor (Kinect), in an arrangement of three tanks. Using image processing algorithms, each one of the tanks is classified within the video sequence as well as each one of the depth measurements is averaged regarding the cross sectional area of the surface of the tanks. This information is used to control the filling level for each tank using a proportional fuzzy controller and a proportional-derivative fuzzy controller. The measured levels by the Kinect were compared to those obtained with an industrial level sensor, concluding that Kinect offers a good measurement precision besides some others inherent advantages like price and multi-sensing.**

**Index terms - Fuzzy control, level control, 3D vision, Kinect, image processing, ultrasonic sensor.**

## I. INTRODUCTION

Nowadays, three-dimensional vision systems are used in many engineering fields. Its development has led to the production of highly reliable and robust products such as Kinect, which using a series of cameras is capable of delivering the color information (RGB) of the scene and the spatial position of the different objects into it (*Deepness*) [1].

Some of the Kinect applications use the depth information in order to recognize the hand movements of a person [2], which can be applied to control robot hands for tracking or imitation.

Similarly, there are some objects identification applications [3] where the characteristics of shape and size can be associated using a combined analysis of the characteristics of color and depth. In [4] a Kinect sensor is used to track a robotic arm on a stage and the possible obstacles in its workspace. In [5] the same sensor is used to control the height of a quadrotor, which shows the functionality of the depth information recollected for the development of all kind of control applications.

Among Other research in 3d vision systems [6] presents a 3D vision system for displacement of a mobile robot in an environment, using a laser and an augmented reality system based on a stereo microscope. In [7] a research is developed to improve the estimation of stereoscopic depth, contemplating the motion information based on edges. [8] and [9] presented studies regarding the use and improvements to stereoscopic vision systems.

On the other hand, fuzzy control systems are very appreciated nowadays because of their capacity to replace the "modeling stage" of the design by a series of rules that describe the behavior and the logic of the controller. In this way, fuzzy controllers are much easier to develop and their performance is usually better. In [10] a classical PID controller was improved connecting a fuzzy system in parallel in order to control the filling and draining cycles of a tank. In this schema, the PID controller operates over the small error variations, while the fuzzy system is in charge of the big ones. In [11] a fuzzy controller was developed, which is able to control the level of a tank in a boiler system in order to avoid the

interference effects of warming and maintain the system stability.

In the present work a simple sensor is used (*Kinect*) as a level measurement system in order to control the liquid level inside a group of three tanks coupled together. Using the deep information measured and applying image processing techniques over the RGB data recollected it was possible to obtain a multi-tank level sensor, which together with a group of fuzzy controllers was able to establish the desired levels of liquid in each tank.

In Engineering applications *Kinect* is a very recent device, for this reason, no previous studies about its use for tank level measurements could be found. The implementation of fuzzy controllers is already known [12] [13], but there is not a tacit comparison between them. Therefore, two fuzzy level controllers were developed (proportional and proportional- derivative) in order to compare their performance.

Next section describes the technique of measuring the liquid level in the three tanks using *Kinect*. At Section III the design of the fuzzy controllers is presented while the results are showed at Section IV. Finally at Section V the conclusions achieved are presented.

## II. LEVEL DETECTION:

The level detection system is based on a *Kinect* sensor which has an infrared projector that generates a pattern captured by a camera of the same type in order to build a depth map of the objects inside the captured image. Besides, it has a RGB camera with a resolution of 1280x1024 pixels, which allows the system to obtain the color information of the image.

Detecting the liquid level in each tank is performed by using image processing techniques in which by analysis of the video sequences captured by *Kinect*, it is possible to obtain both, the color information (RGB) for the segmentation of each tank, and the depth measures to establish the current level. The sensor is located so that covers the area occupied by all three tanks from the top, as shown in Figure 1.

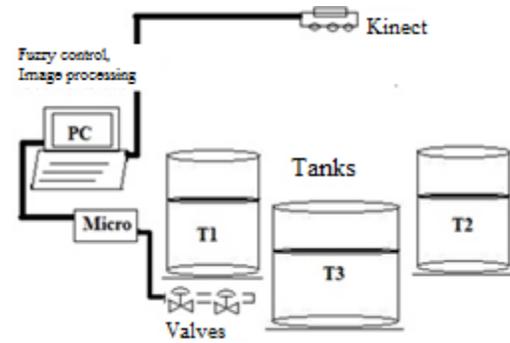


Figure 1. Application diagram

Considering the depth measurement technique used by *Kinect* (infrared), it is not possible to perform the measurement directly on the liquid; therefore, it becomes necessary to use a float in each tank which allows the measure of depth. Each float is highlighted in the scene using a color on its upper surface, which contrast with the surroundings and makes them easier to identify.

A gray scale transformation is performed over the captured RGB image in order to apply a thresholding process and thereby obtain a binary image in which it is possible to detect the edges of the present objects using the Canny algorithm [14]. Figure 2 shows the result of the procedure over one of the captured image using *Kinect*.

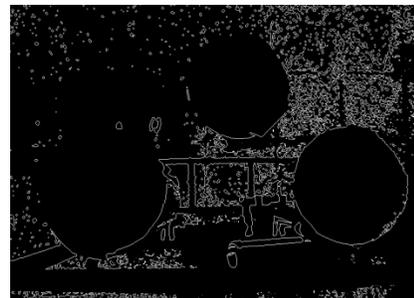


Figure 1. Edges of the captured image

The location of the tanks is determined by the circular structure of them; therefore, a Hough transformation [15] for detecting circles is applied over the obtained image. Figure 3 illustrates the found circles in the image after applying the transformation.



Figure 2. Detection of the tanks

It is noted that the lateral tanks (tank 1 on the left and 2 on the right) coincide with the centers of the found circles, however, this does not happen with the central tank (tank 3), where the radius calculated was higher, creating a mismatch between the centers. Nevertheless, tank 3 is still contained within the radial area.

The first step is identifying those pixels located between the center of any of the circle found and the perimeter of the respective tank. If a non-black pixel is found within the established perimeter, the center is recalculated as its current position minus half of the distance between the pixel found and the perimeter. This process sometimes generates a smaller circle than the tank, but ensures obtaining only pixels inside it.

The found pixels, which represent the surface of the float, are linked with their own depth measure, averaging each of the values and obtaining the current level of the tank ( $Nt$ ) using Equation 1.

$$Nt = \frac{1}{n} \sum_{i=0}^n [d_k - P_i(x, y)] \quad (1)$$

Where  $P_i(x, y)$  represents the depth values measured by the Kinect for each pixel of the cross sectional area of the tank, and  $dk$  is a constant as shown in Equation 2.

$$dk = Dt - (A + B + C) \quad (2)$$

$Dt$  represents the distance between Kinect and the ground;  $A$  represents the thickness of the float;  $B$  is the thickness of the tank base and  $C$  represents the distance from the tank to the ground, as illustrated in Figure 4.

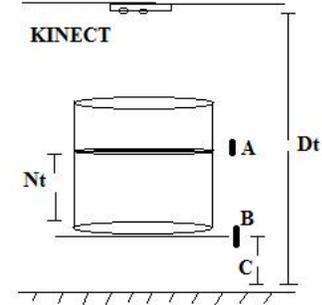


Figure 3. Representative measures

The number of pixels will depend on the radial segmented area for each tank and will be associated with the depth image generated by the Kinect, as shown in Figure 5.

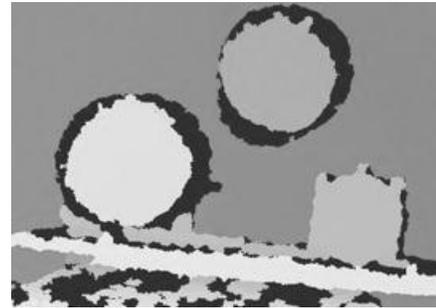


Figure 5. Depth image generated by Kinect

### III. FUZZY CONTROL SYSTEM:

The fuzzy control system developed makes it possible to have an analytical model of the group of tanks. Tanks 1 and 2 in Figure 1 will be controlled by a fuzzy system; A Proportional fuzzy controller will be developed for tank 1 and a Proportional Derivative fuzzy controller for tank 2. On the other hand, tank 3 will be controlled by an ON / OFF controller.

The system operates over a group of proportional control electro valves, which allow the filling and draining flows. The reference of these electro valves

is FESTO MPYE and their operation range is 0-5 volts (maximum and minimum flow respectively) as shown in Figure 6.

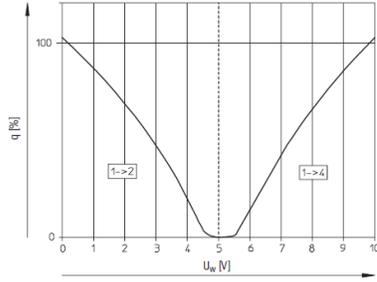


Figure 6. V-Q diagram of the proportional valves

According to the valves response, the membership functions and the universe of discourse of the outputs were deduced. This universe sets the duty cycle percentage (0-100) of a PWM signal, which will be responsible of controlling the actions of the valves.

#### A. Proportional fuzzy controller:

The controller input is determined by the difference between the current level of the tank ' $N_t$ ' and the reference level ' $N_r$ ', which corresponds to the control error ( $e$ ), as shown in Equation 3 and as evidenced by the control loop illustrated in Figure 7a with an ultrasonic sensor and Figure 7b with Kinect. The universe of discourse of the input is determined as a percentage between 0 and 100 when ' $N_r$ ' is greater than ' $N_t$ ' (positive error), and between 0 and -100 when ' $N_r$ ' is smaller than ' $N_t$ ' (negative error). The percentage relation is given by the filling level of each tank ( $N\%$ ) compared to its maximum capacity (100%), establishing the universe of discourse shown in Figure 6a for each one of the tanks under control.

$$e = N_r - N_t \quad (3)$$

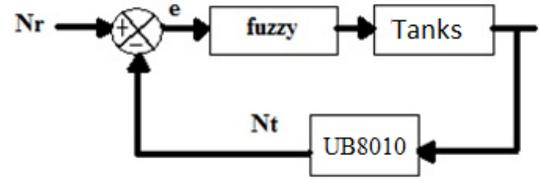


Figure 7a. Block diagram of the control loop with an industrial level sensor

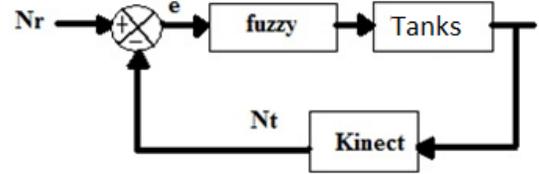


Figure 7b. Block diagram of the control loop with Kinect as level sensor

The outputs are the control signals of the filling and the drain valves ( $V_{in}$  and  $V_{out}$  respectively) and they activate the opening and closing of the valves. Therefore, they are set in a range between 0 and 100, generating the universe of discourse for  $V_{out}$  and  $V_{in}$  (Figure 8b).

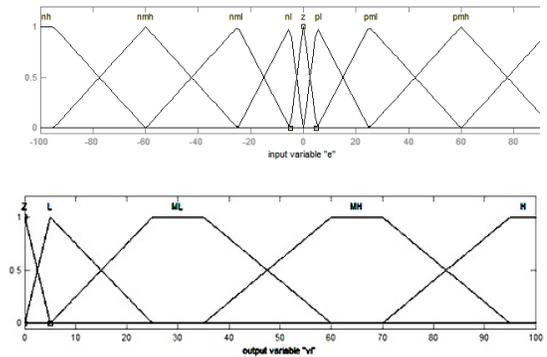


Figure 8. (a) Top: input fuzzy set. (b)Bottom: output fuzzy set.

#### B. Proportional-derivative fuzzy controller:

The implementation of the fuzzy proportional-derivative controller involves a new fuzzy set based on the rate of change of the error ( $de$ ) between a sample  $K$  and the immediately preceding ( $K-1$ ). Because of the 'slow' nature of the system, the error variation is small, therefore the fuzzy set is

established in the range of  $\{-10, \dots, 0, \dots, 10\}$ . Figure 9 illustrates the membership functions for the error variation.

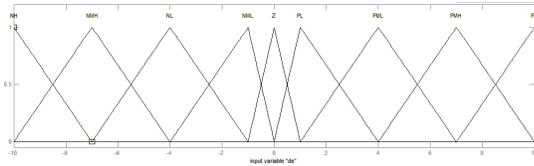


Figure 9. Fuzzy set of the derived output

Given the controller structure, the fuzzy rules for each valve are determined by each control operation to be performed according to both inputs: the error and its derivative.

The defuzzification method uses the center of gravity procedure, described by equation 4, which is the one that best matches the fuzzy set type [16].

$$y' = \frac{\sum_{j=1}^F \mu_{B'}(y_j) * y_j}{\sum_{j=1}^F \mu_{B'}(y_j)} \quad (4)$$

#### IV. RESULTS:

The experimental framework of the fuzzy system is done using the Open Source Fuzzy Logic Library (DotFuzzy); the image processing algorithms are developed using OpenCV 2.1, and all the development is implemented under Microsoft Visual C# 2010 Express Edition. The test computer has a 2.4GHz processor and 2GB of RAM.

Using a resolution of 640x480 pixels for both the depth and the RGB camera, the application operates at a speed of 50 frames per second; worth noting that this value its theoretical and correspond to the speed at which the application process a single frame; concerning the actual processing speed of the algorithm, it lays within the maximum refreshing rate of the Kinect, which is limited to 30 fps.

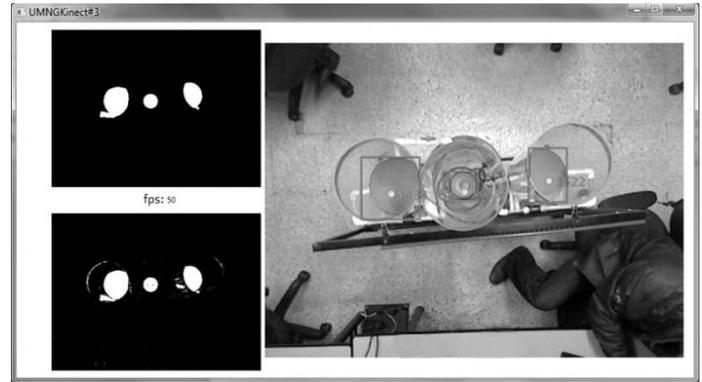


Figure 10. Snapshot of the developed application

The control signal of each tank is supplied by a PIC18F2550 microcontroller, which generates a PWM output for tanks 1 and 2 and an ON/OFF signal for tank 3. At the same time, this microcontroller receives a main control signal from C#.

The validation of the measurement system, was made through the comparison of the measurements taken with an industrial ultrasonic sensor (PEPPERL-FUCHS UB8010-18GM40-I-V1) which takes measures between 50mm and 800mm with a resolution of 0.4mm. At Table 1 the results of these measures are shown.

Table 1  
Level measurements with several sensors

<b>Kinect</b>	80	70	59.9	49.9	39.8	29.8	19.7	9.6	0.5
<b>UB8010</b>	80	70	60	50	40	30	20	10	5

The measures provided by the UB800 were taken as the real measures. From the table V an estimation of the measurement error between the Kinect and the UB800 is performed. This estimation showed an average error of 0.35% which denotes an appropriate performance against an industrial sensor.

At Figure 10, several error percentages are shown, it is observed that this error is directly proportional to the distance, yet the relation is exponential instead of linear; this is noted in [17].

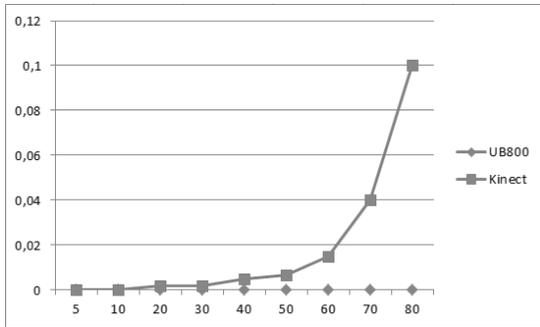


Figure 11. Measurement error between ultrasonic sensor and Kinect.

Figure 11 shows the control action results for a filling cycle when the same reference level is set for tanks 1 and 2; Figure 12 shows the same results but for a draining cycle.

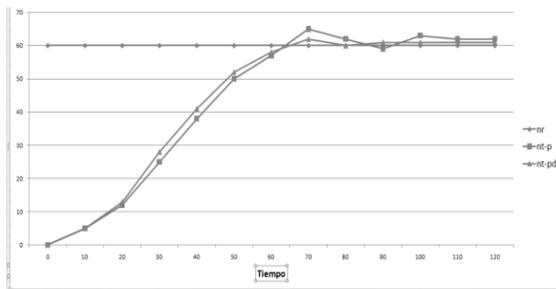


Figure 12. Controller's response to the filling process.

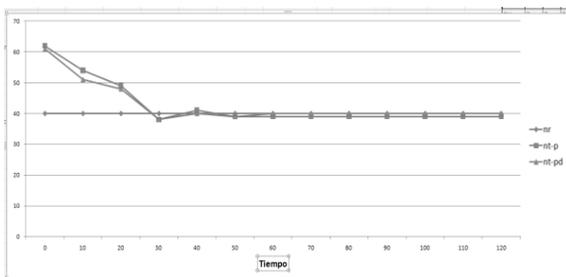


Figure 13. Controller's response to the draining process.

It can be seen that the 'PD' controller generates a filling and draining responses slightly higher, which makes the system responds in a shorter time. For different reference levels the response time is about 1.5 seconds.

The system presents a maximum overshoot of 6.6% and steady-state error of 4.2%. Borderline cases of these two parameters are given for small control actions in which the valves do not close completely due to the defuzzification method (center of gravity), which keeps a remaining voltage on the valves input. For this reason, a correction signal is generated by software on the microcontroller in order to bring this value to zero.

## V. CONCLUSIONS:

It was possible to develop a level measurement system using Kinect, which using image processing techniques is capable of sensing efficiently the filling rate from three different tanks in order to apply this information in an automation system. Thus decreasing the number of sensors, the respective communication lines and the acquisition times associated with each sensor.

Regarding the performance of the fuzzy proportional and the fuzzy proportional-derivative control systems implemented in the level control loop, very similar results were obtained. Given the slow nature of the process, the improvement in time of the 'PD' controller (1.5 seconds avg.), does not represent a significant advantage over the 'P' controller, but since the structure of the first one (PD) is more complex and does not generate an additional computational load to the system; this controller is ultimately more efficient.

It was established that for this punctual application the error level of the Kinect versus an industrial ultrasonic sensor was about of 0.35% in a range of less than 160cm, this is a good level of accuracy besides the cost of the implementation using ultrasonic sensors for the three tanks would cost up to 8 eight times the cost using a single Kinect.

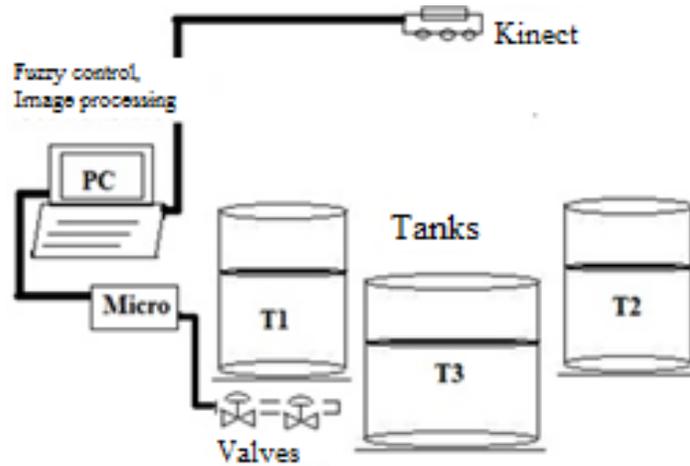
Using multiple Kinects with non-interference techniques may be used to further improve the efficiency and resilience of the system, moreover increasing the capabilities of the system to perform multi-sensing tasks over multi-processes systems. Besides, the use of more non-dedicated and low cost devices is proposed as future work.

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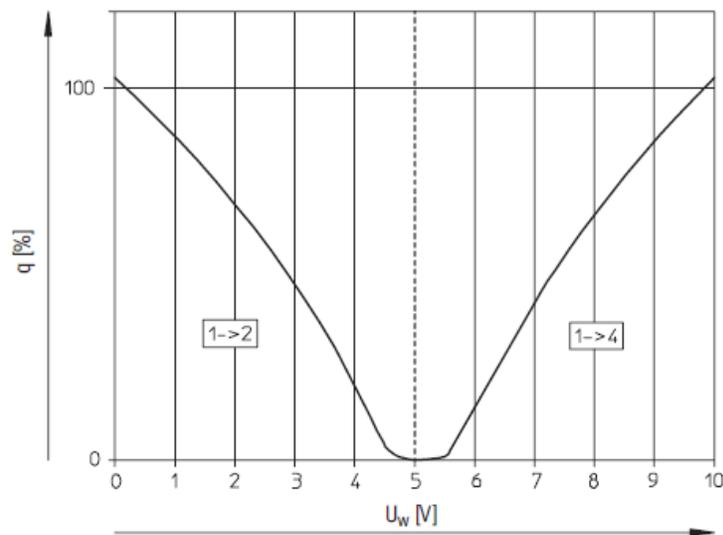
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**FIGURES WITH TEXT**



**Figure 1.** Application diagram



**Figure 6.** V-Q diagram of the proportional valves

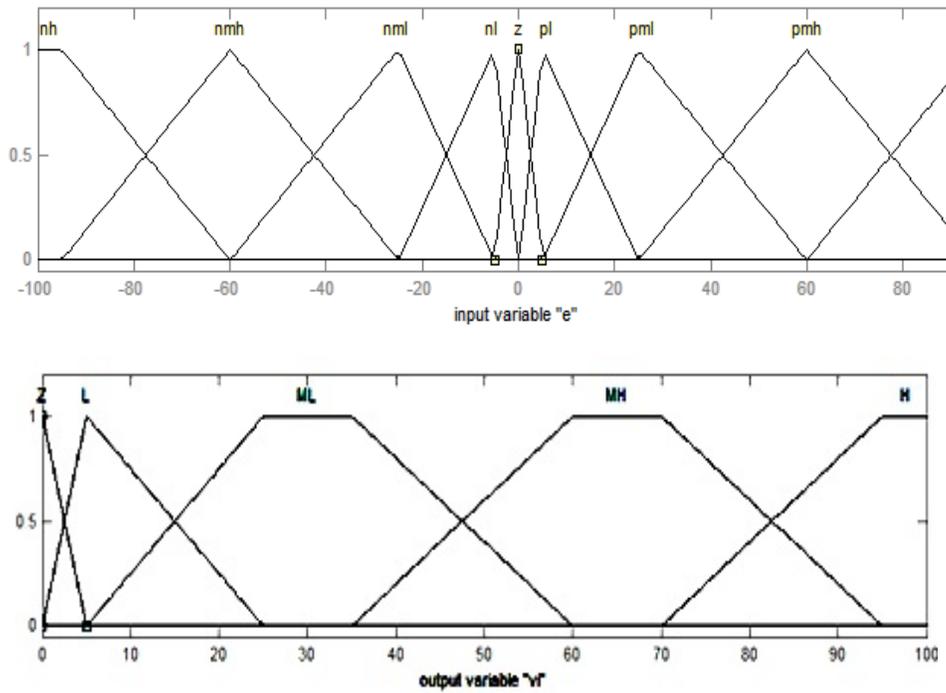


Figure 8. (a) Top: input fuzzy set. (b)Bottom: output fuzzy set.

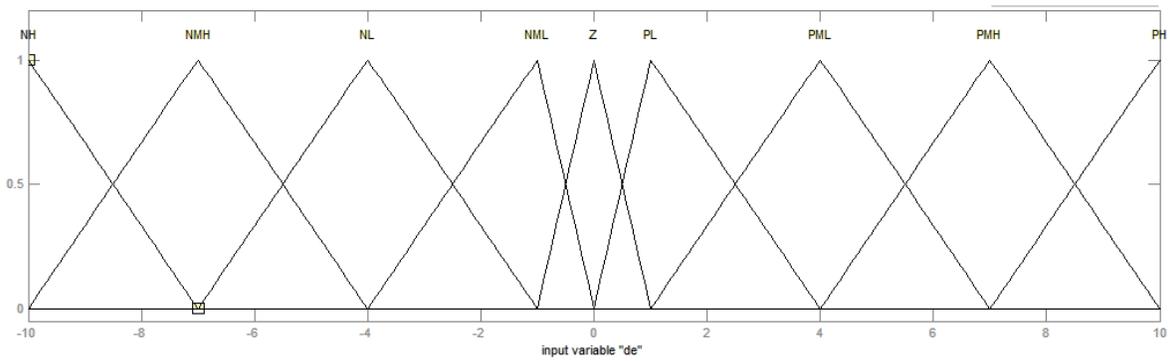


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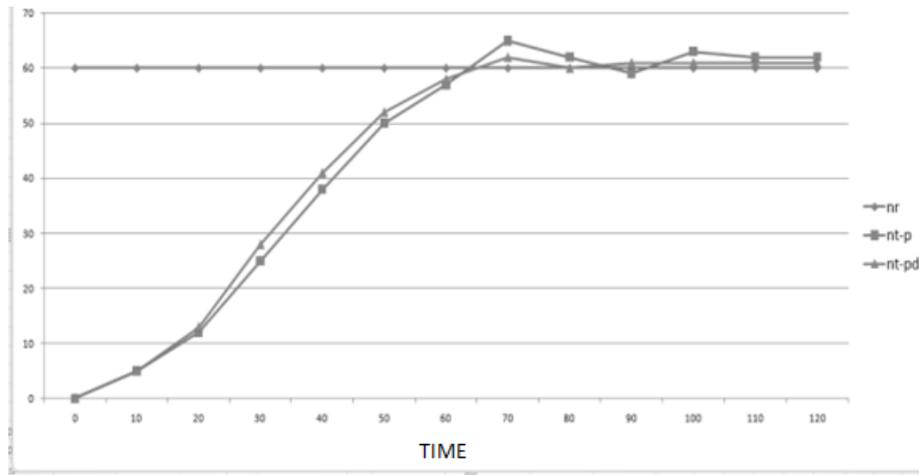


Figure 11. Controller's response to the filling process.

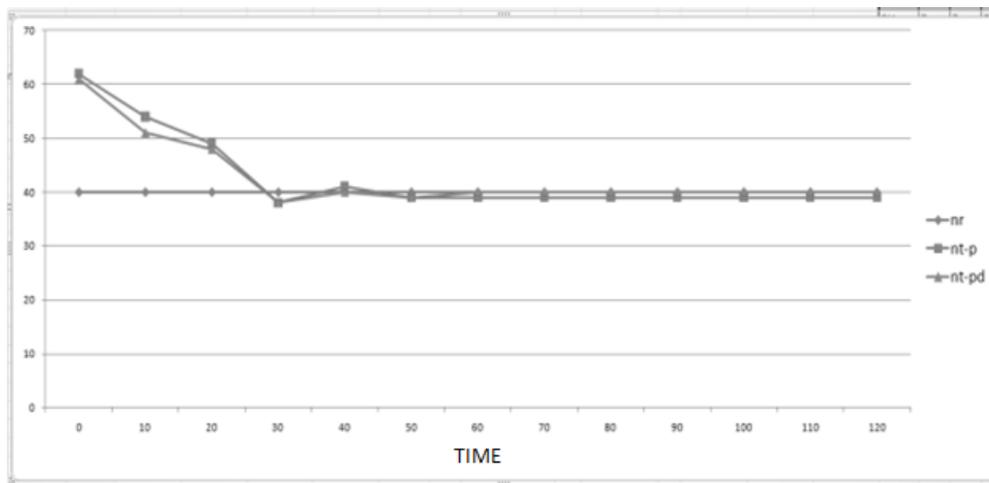


Figure 12. Controller's response to the draining process.