The goal of this project is to write a program in the C++ language that can recognize motions made by a subject in front of a camera. To do this, in the first place, a sequence of distance images has been obtained using a depth camera. Later, these images are processed through a series of blocks into which the program has been divided; each of them will yield a numerical or logical result, which will be used later by the following blocks. The blocks into which the program has been divided are three; the first detects the subject’s hands, the second detects if there has been movement (and therefore a gesture has been made), and the last detects the type of gesture that has been made accomplished. On the other hand, it intends to present to the reader three unique techniques for acquiring 3D images: stereovision, structured light, and flight time, in addition to exposing some of the most used techniques in image processing, such as morphology and segmentation.

1. Introduction

New technologies tend to develop interfaces with a high degree of usability. Human-computer interaction (HCI) is a research field that is in continuous evolution. Technology companies, particularly leading companies in video games, are investing an essential part of their resources in the development of new interfaces that seduce the user with new forms of communication with the machine, applicable to both video games and control of multimedia environments. In [1], a visual memory system (VMS) is used that stores different static patterns of the postures to be recognized and uses the Hausdorff distance for their separation. In [2], proportions and geometric characteristics of the hand are used for their description, and in [3], the silhouette is transformed into a function of the distance from the outline to the base, in addition to using your orientation angle. The projection of a 3D world to a 2D one entails the loss of depth information, which implies that, given a 2D image, the geometry of the observed 3D scene cannot be unambiguously reconstructed. Nature has satisfied the need for depth perception for humans and most animals with a binocular vision system, which, based on binocular disparity, extracts information from a pair of 2D projections. For the human being, it is an action that does not require any effort. However, the reconstruction of 3D geometries using different sensor systems has become a major technological challenge. Over the past decades, various technologies have been proposed for this purpose; the latest advances in electronic optics, sensor design, and computational power have achieved high resolutions (>300 kpx) and a temporal sampling as close as possible to real time (≥30 Hz) [4–8] in the acquisition of 3D images [9–17]. The project’s objective is to create a program that processes the images obtained using a depth camera, which uses the structured light technique to get the depth of the scene, in front of which a subject is located, to decide if it is or not making a gesture.

2. Methodology

2.1. Depth Cameras. The devices used for capturing 3D images are called depth cameras.
The leading technologies used by these devices are stereoscopic vision, structured light, and time of flight, explained in previous paragraphs.

Among the models that exist on the market, depending on the type of technology used, are the following models:

- **Stereoscopic vision:** Panasonic AG-3DA1
- **Structured light cameras:**
  - (i) Kinect
  - (ii) Asus Xtion PRO LIVE
  - (iii) Ensenso
  - (iv) SR4000 modulated light (Swiss Ranger)—Pulsed light Tiger Eye (Advanced Scientific Concepts, Inc.)

### 2.1.1. Main Elements

The main features of a depth camera are a low-resolution RGB camera, infrared camera, and infrared projector.

### 2.2. Operating Principle

To know the distance of an object in the scene to the camera, it is also necessary to have a projector, which in the case of the model is an infrared projector, so the camera that captures the scene must be as well. The operation is as follows.

1. The infrared light projector is turned on, projecting an irregular pattern of dots, with a wavelength between 700 nm and 1 mm, which belongs to the infrared light spectrum, which is not visible to the human eye. It is possible to visualize this pattern with a night vision camera
2. With the pattern projected on the scene, the infrared camera will be in charge of capturing the infrared light returned to the camera because, unlike RGB cameras, infrared cameras contain CMOS sensors that can detect infrared light that "bounces" off all objects found in the scene
3. At this point, the depth for the pixel in the scene is calculated; this is done by taking individual images, the one captured by the sensor, and the initially projected pattern of points, finding the corresponding points between both images, to later use triangulation and thus calculate the final distance to the object

### 2.3. Image of Distances or Depth Map

The appearance of a distance image presents as small values (dark pixels) in the areas where objects are closer to the camera and higher values for objects further away; the black areas of the image correspond to errors when the image was captured. In the picture, black regions correspond to the shelves and armrests of the chair, which, as seen in the left image, are dark.

### 2.4. Image Treatment Techniques

#### 2.4.1. Morphology

Morphology is signal processing based on maxima and minima. Erosion and dilation are the most widely used techniques for this type of processing.

- **Structural Element.** The structural part is in mathematical morphology that a convolution mask (or kernel) is in linear filters. It has a centre (or anchor point).
- **Erosion.** Erosion consists of the expansion of the dark areas of an image. For the process, the structuring element is moved through all the image pixels, placing the centre of the EE on each of these. In the resulting image, the value of each pixel will be the minimum of the pixels under the mask in the original image
- **Dilation.** It is a process similar to that of erosion, but what is produced is an extension of the transparent areas, in this case, taking the maximums under the mask. It is observed that when moving the cover through all the pixels of the image (left), the result (right) finally obtained is an enlargement of the light areas (or reduction of the dark spots)

#### 2.4.2. Segmentation

The different objects that can appear in an image are formed by a set of pixels. To identify them, it is necessary to group pixels with similar characteristics since each one of them presents features such as

- (i) colour
- (ii) movement
- (iii) texture

Segmentation by flat areas consists of grouping pixels with similar characteristics, such as those listed above, also made up of connected pixels. This is the one that is applied in the project.

### 2.5. Gesture Recognition

As previously introduced, the objective of this project is the recognition of gestures made by a subject placed in front of a depth camera. After having exposed some of the 3D image capture techniques, the main elements of a depth camera, and its operating principle, as well as some image processing techniques, precisely the most important ones used in this project, in the following sections, we will see that they will present the different stages that have been used to achieve this objective. Previously, it was necessary to know the type of gestures that will be detected, so the variation of their coordinates in pixels over time will be analysed.

**Nature of gestures.** The type of gestures to be detected contains movement in a single direction depending on the gesture. Specifically, these are gestures in which the hands only perform horizontal or vertical movements, as appropriate. The movements to be analysed consist of 3 stages. The stages are called initial, movement, and final; in the initial and final stages, the subject remains almost static since the movement’s beginning and end [18, 19]. The gestures that have been captured in the sequences, and the stages of each movement, will be shown below. Due to the similarity of the movements to each other, two horizontal movements will be analyzed.
In Figure 1, it can be observed how the movement is carried out by varying only the \( x \) coordinates of the left hand; the three stages of movement are also differentiated:

(i) At the moment of the start, the graph hardly varies; this is because the hands are not perfectly static; if that were the case, that area would have the shape of a horizontal straight line.

(ii) Movement is the time-space in which the hands begin to move; their displacement is to the left, so the \( x \) coordinate value increases.

(iii) End, a moment in which the gesture has come to an end, has the same characteristics as the start moment.

You must stay in the last state for a while before starting another gesture. Otherwise, neither of you would be detected.

In the gesture in Figures 2 and 3, both hands carry out the movement, which is also horizontally, so, as can be seen in the graphs of the \( x \) coordinates of both hands, the phases of the movement are as follows. Same for each hand, as was the case in the previous gesture (1.a), in this case, it should be noted that in the case of the right hand, the \( x \) coordinate decreases since the movement is to the right (the pixel coordinates are such, that the pixel in the upper left corner has the coordinates \((0, 0)\).

**Design stages.** For the elaboration of the work, three subobjectives have been proposed, each of which comprises a design stage with its corresponding blocks, and the function of these stages will be explained in the following paragraphs.

2.5.1. **Detection of Objects (Hands).** This is the first stage of development of the program, as it works directly with the distance image captured by the depth camera. It is possible to distinguish two blocks, that of "detection of minima" and "classifier of minima."

1. **Detection of Minima.** This block has been made by creating a function that uses the erosion method, a previously...
explained technique. The goal from here is to detect which objects in the scene could be hands.

Taking Figure 4 as an example, obtained from one of the test sequences used, it is observed that the subject is sitting in a chair facing the camera with his hands up and around are other objects such as shelves and chairs. To begin with the detection of the gesture, first, the detection of the hands must be carried out; for this, the minimum of the image must be found, among which two must be found, those corresponding to the hands.

**Distance image preprocessing:** as mentioned above, due to measurement failures, black areas appear in the image, the pixels of which have a value of zero. The area of interest, that is, the one where the subject must be located for the correct detection of the hands, is around 1.5 m; it is also observed that behind the subject there is a wall, which is located about 3 meters, the value was chosen as maximum in the image. For this reason, the distance image will be modified before starting the minimum detection process.

(i) Firstly, the pixel value of the image is cut to the maximum value chosen so that an image like the one in Figure 5(b) would result since the objects near the wall will not be of interest since the subject is not going to be in that area

(a) The maximum value then replaces the black pixels. The result is Figures 5(a)–5(c)

Once the preprocessing of the image is finished, you can start with the detection of minima.

**Minimum detection:**

(i) The first step is to erode the image; as explained above, this is done to expand the dark areas of an image. The structuring element is finally chosen, after carrying out tests on different frames, which has been a rectangular element with a vertical component twice as significant as the horizontal one. So, it has dimensions designed so that the minimum corresponding to one hand does not interfere with the other; that is, the two minimums can be detected

(ii) Once the erosion has been carried out, we compare the two images, the original and the eroded one. After performing the erosion, the minimum values of the image have been expanded to surrounding areas; in the resulting image, only the pixels that coincide in both images will be present, which will correspond to the minimums

(iii) As can be seen in the resulting Figure 5(f), there are several areas considered minimal. Still, it can also be observed that most are on the sides of the image that will not be regarded as areas where the subject could be located since it must be located preferably around the central area. Therefore, we proceed to “clean” the image to stay only with the central region; from the result, we will proceed with the next block

(2) **Minimum Classifier.** Based on the result obtained in the “minimum detection” block, this block will be in charge of deciding whether or not the current minimums in an image correspond to a pair of hands, according to a series of criteria that must be met.

A previous step before starting to classify these minimums is to segment the image, for which we will apply the method of segmentation by flat areas that have been exposed previously so that we will have the minimum of the image labelled. As can be seen in Figure 5(g), the minimums are points in the image that correspond to the hands. Still, to classify them, we need to obtain their natural shape, or in other words, we must extract from the original image the entire contour of the object. A function has been created that tracks pixels similar to a given one to carry out this work. Once the coordinates of the pixel of interest are known, which the minimum is, the coordinates of those pixels that are in the vicinity of the point are stored, considering that
they are pixels that belong to the hands. The decision as to whether or not a pixel belongs to the hand is made because the difference between the minimum corresponding to it and that of the evaluated pixel does not exceed a certain threshold. In Figure 6, you can see the result after applying the function to the minimum image.

The criteria according to which it will be decided if there is a pair of hands in front of the camera in the scene will be the dimensions, since, as is evident, there cannot be excessively large or tiny hands.

The separation between both hands changes depending on the subject under test, but it should not be too high a distance either. Once hands have been detected in the scene, it is assumed that in front of the camera, there is a subject who will make a gesture; therefore, once this block is finished, the movement detection block is passed. Otherwise, they continue to be analyzed distance images waiting for hand detection. The final result of this project is put to the test using sequences; each frame of the sequence must go through this block, finally obtaining its 3D coordinates for each detected hand.

2.5.2. Motion Detection. Once the detection of the hands has been carried out, using the previous block, to detect whether the subject in front of the camera is making a gesture or not, it is necessary to check if the hands are making any
movement, which can be recognized as a gesture that the program can detect [20]. For this, the following block has been implemented, which is made up of two blocks.

Frame by frame, the coordinates of points corresponding to both hands are obtained, and they are stored in different arrays, one for each hand.

(1) Movement Blocks. The gestures will take place during a series of consecutive frames, generating 3D points in each of them. For a considerable variation to be perceived from one frame to the next, the subject should perform swift gestures, in addition to not remaining perfectly static while not making any movement; the points oscillate around a position. For this reason, to analyse the gestures, a series of blocks will be used, which we will call movement blocks; each gesture will be made up of a series of these blocks.

Each of these blocks is created using a series of 3D points, specifically 7; in later sections, the reasons for this choice will be explained. Taking the first and last point of these, its Euclidean distance is calculated and compared with a certain threshold; if this distance exceeds it, it is considered a block of movement. Otherwise, it will be a block without movement.

These blocks must also be stored, to be analysed later, by the following processing block.

With the movement blocks obtained, the value is of a logical type since it is “1” if there has been movement and “0” if there has not been, obtaining a sequence of “1’s” and “0’s.” To perform the analysis of these sequences, the phases of the gesture explained in previous paragraphs will be used. The use is made of the block called “Pattern Analyser” to recognize these phases of the gesture.

(2) Pattern Analyser. This block is in charge of analysing the binary sequences to detect gestures; the way to do it will be to compare these sequences with some patterns, which will be dynamically generated based on the input to the block. Taking into account the aforementioned parts that make up a gesture of minimum duration, we obtain the pattern with which the sequence must be compared as follows, assigning the corresponding block value:

(i) Block 1: initial moment (semistatic-movement): 0

(ii) Block 2: instances corresponding to the effective movement of the hand (movement-movement): 1

(iii) Block 3: instances corresponding to the effective movement of the hand (movement-movement): 1

(iv) Block 4: final moment of the gesture (movement-semistatic): 0

In this way, the minimum length pattern has been obtained.

In this way, the minimum length pattern to which the input blocks to the analyser should be adjusted (pattern: 0 1 1 0) has been obtained. Since it could be a gesture of longer duration, or that the subject takes longer to perform it, the pattern must extend its length dynamically, that is, depending on the input sequence, and it can have the following lengths and shapes:

\[
\begin{align*}
-0 & 1 1 1 0 \\
-0 & 1 1 1 1 0
\end{align*}
\]  

As can be seen, these are patterns with the same structure as the minimum size but that contain in the central part of the pattern a more significant number of blocks of value 1. These will be the patterns used to detect gestures of longer duration, and therefore, the part corresponding to the actual movement of the hand is more extensive. Once we have the four blocks that are minimally needed at the input of the pattern analyser, we proceed to compare the input sequence block by block with the minimum length pattern that we have obtained previously; each and every one must match one of the blocks with the corresponding pattern blocks.

The comparison of the pattern with the sequences has been carried out using a state machine, which is made up of three states:

(i) Initial state (“0”): corresponds to the static part at the beginning of the gesture

(ii) State of movement (“1”): corresponds to the active part of the gesture

(iii) Final state (“0”): corresponds to the final part of the gesture and the last of the states through which the sequence must pass

Its operation is as follows:

(i) Initial state: it is assumed that the first block of the sequence must be equal to “0,” which corresponds to the first block of the pattern. If the comparison is accurate, it advances to the state of movement; if not, the state next should be the initial

(ii) Movement state: if this state is reached, it means that the first block of the sequence that could correspond with gesture has the value of “0.” Therefore, the next one must have a value equal to “1,” which corresponds to the moving part. We carry out the
2.5.3. Gesture Recognition. This is the last block of the program design, as it is responsible for identifying what type of gesture is being carried out once it has been decided that a gesture is being made. The previously stored 3D coordinates will now also be helpful to identify the type of gesture performed, as the pattern analyser has taken note of the start and endpoint of the gesture.

The operation is as follows:

(i) In the first place, it must be verified whether the gesture is carried out by moving a single hand (the other remains in the initial position of the hands) or with both; for this, the two logical values that the pattern analyser returns are checked. Therefore, from here on, a first selection has been made between two possible groups, gestures with two hands and gestures with one hand

(ii) Subsequently, the sequences of points corresponding to the gesture performed must be obtained; this is carried out by using the pairs of values returned by the pattern analyser, which indicate the beginning and end of the movement that is being carried out. Performing each hand, these values are the indices that correspond to the points within the arrays

(iii) Before proceeding to the next step, it is necessary to clarify once the information contained in the z coordinate has been used. As previously stated, the gestures that have been proposed to detect have only a vertical or horizontal component (x or y); these will be the coordinates that will be used in the next step

(iv) At this point, it only remains to determine if the movement carried out is horizontal or vertical, whether the movement has been carried out with one hand or with both, the procedure that has been followed to determine the direction is the same, the generation of some horizontal and vertical variables (deltas), that is, the amount of movement that has occurred in one direction or another, which have been obtained by calculating the difference between the x and y coordinates, respectively, between the beginning and the end of the sequence of the duration of the gesture of the corresponding point vectors

2.5.4. Obtaining Parameters. To create each of the movement blocks of the movement detection stage, 7 points have been used as mentioned above. Still, to determine whether or not there had been movement in a block, it was necessary to obtain a parameter. Obtaining this parameter has been carried out by analysing the sequences of points corresponding to the different gestures for each hand in particular.

The following graph shows the path followed by one of the hands during a horizontal movement. In the results (image not given), parts corresponding to a gesture described above can be distinguished: the initial part, the part corresponding to the movement itself, and the final part.

The oscillating points at the beginning correspond to the moment in which the hands are located in front of the camera, which ideally should remain static. Still, since the subject cannot be perfectly static, the points will have different locations (speaking in pixels), that is within of a prudent radius.

(i) The intermediate part corresponds to the instants concerning the movement; it can be seen how the points have a greater separation from each other, which means that from one frame to another, there has been a more significant amount of movement than there was in the initial part

(ii) In addition, these extend horizontally, as expected since the movement was of this type; it is also clear that this phase lasts longer than the others because, as said at the beginning, it corresponds to the effective moment in which it is performing the gesture

(iii) The end part is the smallest, but in it, it is perceived that the subject has finished the movement, since the points belong to this part of the sequence; even despite having some oscillation, this is relatively reduced. The points are not very distant from each other

Focusing on the central part of the graph corresponding to the movement, it can be observed that already in seven points, that is, a sequence of seven consecutive points, taking into account the initial and final parts of the said sequence
the Euclidean distance obtained with the 2D coordinates ("x" and "y") is already 41, which is much higher than if we took the initial and final parts of the initial or final instants of the graph since indeed there has been movement; for this reason, this number has been taken of frames to form the so-called “movement blocks” in previous sections, also as mentioned above due to the condition about the minimum duration of a gesture; a single movement block is not enough to consider that there has been a gesture.

On the other hand, to determine the threshold from which it is considered that there has been movement, different Euclidean distances of different sequences of gestures have been taken, initially concluding that on average, this value should be equal to 20; however, after testing it in different sequences, it has been varied to fit them better, so that it is valid for most of them, finally leaving a value of 17.

3. Results

During the performance of the work, drawbacks have been encountered, such as the nondetection of the hands, as in the example in Figure 6. It is observed that the hands are not detected when they are too close together because due to the resolution of the distance image, a single object is detected instead of 2.

Finally, after going through each of the blocks, the following results have been achieved until finally detecting a gesture made by the subject in front of the camera.

3.1. Hand Detection. An image of distances has been generated from the scene shown, which is used to detect the hands.

3.2. Generation of Movement Blocks. With the hands detected, from their 3D coordinates, the so-called movement blocks are generated, which will later make it possible to decide whether or not there has been a gesture.

3.3. Gesture Identification. Analysing the coordinates of the hands again, once it has been detected that there has been a gesture, the type of gesture performed must be identified; it is a gesture of the horizontal left hand and to the left.

In addition, it is worth mentioning the conditions in which the program works properly:

(i) In the first place, it must be detected that there has been no movement; then after a series of moving blocks, another block without movement must be detected again, at the one that the gesture was detected. Although this fact was already exposed in previous sections, the fact of making two consecutive gestures without there being a minimum pause...
between them would cause the nondetection of the said gesture

(ii) If the movement lasts more than 2 seconds, it will not be detected, as the analyser will not find the pattern, or rather the sequence will be overwritten with a more recent one, so it will never reach it

Finally, it is necessary to list the results obtained; that is, the gestures that have been detected are those shown in Table 1.

4. Conclusions

Throughout this project, two stages have been passed: minimum detection, in which the distance Figures 7 and 8 is worked directly, and the stage in which the information obtained from the distance image is worked, corresponding to the “motion detection” and “gesture recognition” blocks. Because the processed information is obtained exclusively from the distance image, certain aspects must be taken into account when both are capturing it and working with it:

(i) The subject should be around 1.5 meters from the camera

(ii) If the hands are too close together, they will appear as a single object. Therefore, there must be a separation of at least 10 cm between the centres of both

(iii) A pretreatment to the image must be carried out to eliminate the black areas in the image. These appear both by the shadows created by the subject itself, preventing the pattern from being projected on certain surrounding regions, in addition to the type of material and colour of the objects in the scene

The possible improvements that can be made to the project are duration and type of detectable gesture [21]. As previously mentioned, the duration of the gestures should not be greater than 2 seconds, as it will not be detected. If you want to see gestures with a more extended period, a simple improvement would be to increase this time, modifying the corresponding parameter. Regarding the types of detectable gestures, the “gesture recognition” block can be modified to use all the 3D points corresponding to the hand (or hands) that performs the gesture to identify the path that follows. You must create a record with the trajectory of the different gestures to compare the detected revolution with the document.

Data Availability

The data underlying the results presented in the study are available within the manuscript.

Conflicts of Interest

The authors declare no conflict of interest.

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