

Fast Obstacle Detection by Variable Step of 3D Laser Scanning for Robot Navigation on Unknown Planet

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Abstract

In the exploration of another planet by mobile robot the slow 3D scanning caused by small step can be a problem. It can be increased by combined scanning step for faster search of n obstacles in unknown surrounding. This is of keynote importance in automatic robot navigation, especially on the surface of another planet. To maintain a reasonable speed robot must to detect dangerous obstacles as soon as possible, and then calculate a safe trajectory in real time. So, the scanning with variable speed and precise digital mapping only for selected spatial sectors is under consideration here. Wide range of simulations in MATLAB of several scenes with variable n obstacles, scanning it with angle value since 0.6° up to 15° was provided aiming to detect such angular values still permitting to get the most information about obstacles without undesired time loss. Three of such angles were obtained in simulation and then rectified by Levenberg-Marquardt Algorithm application and were applied to micro electro transmission design for practical realization of variable combined step scanning on our previously known laser scanner.

Keywords: Planet exploration; Mobile robot; TVS; Scanning step; NN; Micro Electro-Mechanic Transmission

Introduction

The exploration on other planet usually is special task for robot navigation vision, optical scanning with laser is attractive to use over others approaches. When the navigation needs: direct information acquisition, low noise, acceptable accuracy with low uncertainty, ability to work in complete darkness, laser scanning is the one option. Optical sensors give reliability and data stability, can see a real trend in the non-contact sensors based in optical technologies [1]. Contactless acquire a large number of points. Furthermore, laser sensors currently on the market do not have the accuracy of the dynamic contact sensors. But, this accuracy depends on different parameters: surfaces color, texture, materials properties.

Recent research [2] tells diverse optical methodologies have been implemented in tracking systems, could for example make do into [2,3]. In MacFarlane et al [4] was demonstrated two-dimensional x-y laser scanning, using a computer controlled, and rotating reflective cube. Advance the simple/durable optic scanning method which is able to give still more open angle-of-view than any known techniques, except the omni-directional vision (fish-eye lens) [5]. Application depth map by an efficient epipolar plane analysis method [6,7] was analyzed on [8] for omni-directional stereo vision. In application on another planet additionally is growing the possibility to detect something absolutely unknown on Earth, so it is impossible navigation based on analysis of camera images [9]. The laser scanner developed in [10] does not provide 3D shape information of the detected obstacles because of the absence of enough computational power for 3D shape reconstruction. The avoidance strategy proposed in [10] is to avoid obstacles regardless of their shape. A good review of 3D model acquisition techniques can be found in [11,12]. The most crucial point in this case is a time of scanning. Obstacle must be detected in a shortest possible time, but the precise coordinate's measurements is not necessary for all obstacle shape, but only for its edge nearest to the robot's desired safe trajectory (see [9,13,14]). This paper, modeling sundry scenes with geometrical obstacles, for lay down simulation tool to find optimal angles which will be combined-step scanning, purpose of design based on stepper-motors

in laser positioning system, designing a micro electromechanical transmission for that job.

Problem statement

Our previous research [9,13-19], TVS we allow 3D Cartesian coordinates measurements for obstacles inside scene during navigation. TVS can give a digital map of the obstacles within the Field-of-View (FOV). The current TVS use constant scanning pitch which is not optimally efficient for robot navigation. The objective in unknown scenes is the scanning with combined scanning step, which will minimize scanning time.

We introduce basic principle of dynamic triangulation in [9,13-19], TVS works under dynamic triangulation theory in order to obtain 3D coordinates of S_{ij} points on objects, as shown in Figure 1 and Figure 2, laser beam is projected by Positioning Laser (PL) into obstacle surface, reflecting back is captured by revolving sensor called Scanning Aperture (SA) [20].

TVS has registered two dynamic angles C_i and B_j [14]. This system is able to track any point in front of the mobile robot calculating the instant triangle with two known angles and precisely known fixed distance between PL and SA.

In order to calculate any coordinate in Cartesian plane was used sine laws on dynamics triangulation by the following equations [17].

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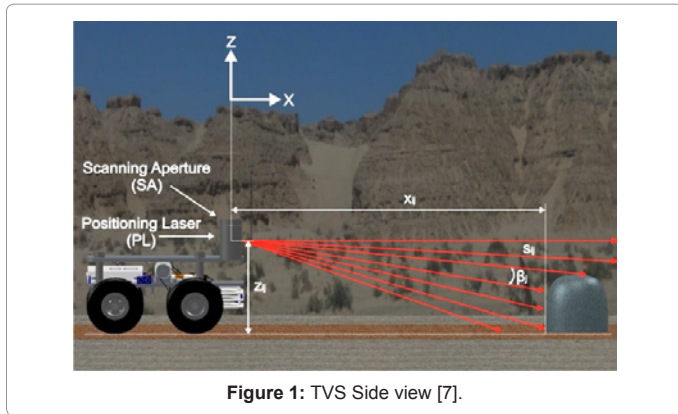


Figure 1: TVS Side view [7].

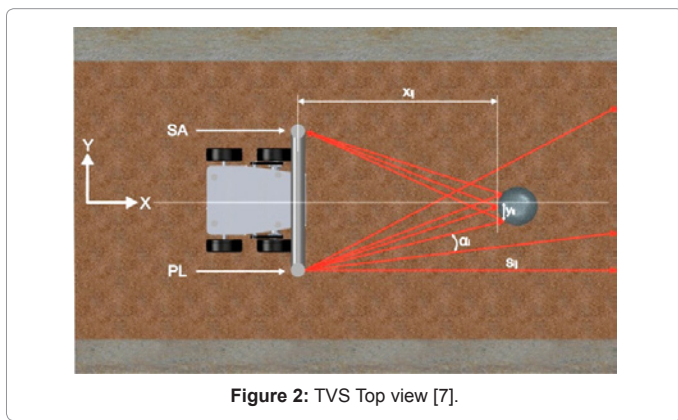


Figure 2: TVS Top view [7].

$$x_{ij} = a \frac{\sin B_{ij} \cdot \sin C_{ij} \cdot \cos \sum_{j=1}^j \beta_j}{\sin [180^\circ - (B_{ij} + C_{ij})]} \quad (1)$$

$$y_{ij} = a \left(\frac{1}{2} - \frac{\sin B_{ij} \cdot \cos C_{ij}}{\sin [180^\circ - (B_{ij} + C_{ij})]} \right) \text{ at } B_{ij} \leq 90^\circ, \quad (2)$$

$$y_{ij} = -a \left(\frac{1}{2} + \frac{\sin B_{ij} \cdot \cos C_{ij}}{\sin [180^\circ - (B_{ij} + C_{ij})]} \right) \text{ at } B_{ij} \geq 90^\circ, \quad (3)$$

$$z_{ij} = a \cdot \frac{\sin B_{ij} \cdot \sin C_{ij} \cdot \sin \sum_{j=1}^j \beta_j}{\sin [180^\circ - (B_{ij} + C_{ij})]} \quad (4)$$

Application in PL of TVS Micro-Electromechanical Transmission (MET) with combined scanning step. This is the most permissive of possible ways in resolution constraints improvement [11]. Some of them can be ignored under the simple criterion that robot cannot path there. Information about obstacles can be reflected by the number of detected 3D points on the obstacle. In Figure 3 is shown the proposal MET.

How simulation was?

The next frame shown a scene (see Figure 4) shown: step angle between two consecutive positions of laser ray, points which the system

detected, are the locations where laser ray is reflected to scanning aperture (Intersection Point Of Simulated Laser Ray With The Obstacle), and figures within the scene are simulations of obstacles. TVS FOV: beginning at 10 degrees and finish at 170 degrees. Within present scene (sizes: 30 m width and 15 m depth).

To simulate laser ray, which unfortunately due to MATLAB'S nature must have discretization for finite time of calculations, the following model based on Euler's identity was used [21]:

$$Ae^{i\theta} = A \cos \theta + iA \sin \theta \quad (5)$$

Where θ is current step angle, A is the length of ray

For simulated obstacles were used next lines equations Y_{r1} , Y_{r2} , Y_{r3} in order to draw a triangle:

$$Y_{r1} = \frac{(y_1 - y_2)(x - x_1) + y_1}{(x_1 - x_2)} \quad (6)$$

$$Y_{r2} = \frac{(y_2 - y_3)(x_2 - x_1) + y_2}{(x_2 - x_3)} \quad (7)$$

$$Y_{r3} = \frac{(y_1 - y_3)(x_3 - x_1) + y_3}{(x_1 - x_3)} \quad (8)$$

Where x_1 , x_2 , x_3 and y_1 , y_2 , y_3 are Cartesian coordinates of the triangle vertices.

For simulated circles were used arbitrary radius (r) and center (C), and discretization of the circle (DC) in grade [20]:

$$\Delta\alpha = \frac{(DC \times \pi)}{180^\circ} \quad (9)$$

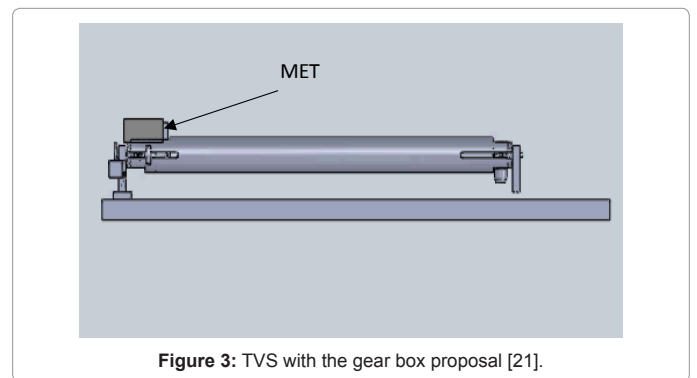


Figure 3: TVS with the gear box proposal [21].

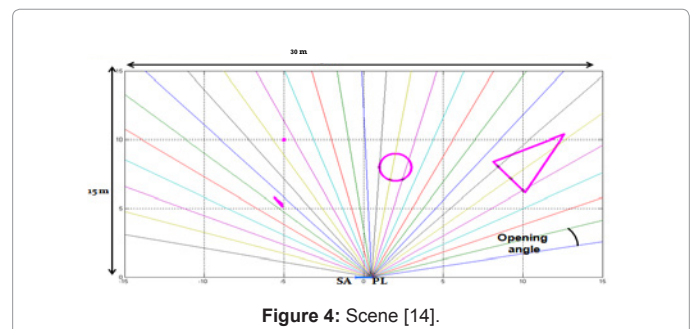


Figure 4: Scene [14].

Where α are increments of $\Delta\alpha$ and $\Delta\alpha$ is the increment in the circle obstacle discretization.

$$Circle_n = C + re^{j \times \alpha} \quad (10)$$

Where $Circle_n$ is the obstacle drawn for n circles.

Parameters and obtained frames analysis

Simulation was 101 different scenes, which varied next parameters [22]:

1. Opening angle (0.6 degrees to 15 degrees with $\Delta\theta = 0.1455^\circ$ for each step)
2. Number of obstacles (1 to 13)
3. Obstacle position (Random)
4. Obstacle dimensions (Random)

Figure 5 show us some scenes of simulation, each scene was scanned with 100 different step angles and varying parameters thus obtaining 10,100 frames to analyze, the notes were:

1. Smaller obstacles and near positioned, were difficult to detected, need angle around 0.6 to 0.7455.
2. On Figure 5a with 8 small obstacles and opening angles 15° and 14.8545° detected one obstacle in minor time.

3. On Figure 5c uses 12 obstacles; scanning time was long in the most of angles.

4. On Figure 5e have a triangle with a wall behavior, in this phenomenon all the angles take more time versus others frames

In simulation data were enough to apply statistical methods (Simulation Graphic Shows: Detected Point Versus Opening Angles) for 101 experiments (Figure 6). The Figure 6 shows values of optimized scanning angles using: small angular increment of scanning step in simulation; the quantity of obstacles in scene 1-13 and laser length was 30 m.

Every scene was analyzed using parameter Z which called advance (The Joint Quantity of All Detected Points on Surfaces of All Obstacles inside the Scene) that was calculated:

$$Z \equiv P / t \quad (11)$$

Where P is number of points detected on obstacles, t is the scanning time in seconds.

Figure 7 represent simplified by maximum and minimum method, where Z parameter is under hypothesis "Find three optimum aperture angles based on the data obtained from 101 scenes using Method connection means and the detected points vs. time", which show us the optimum angles in Figure 8.

In all possible values of scanning step there are three values of angle where the scanning velocity grow continuously but the quantity of

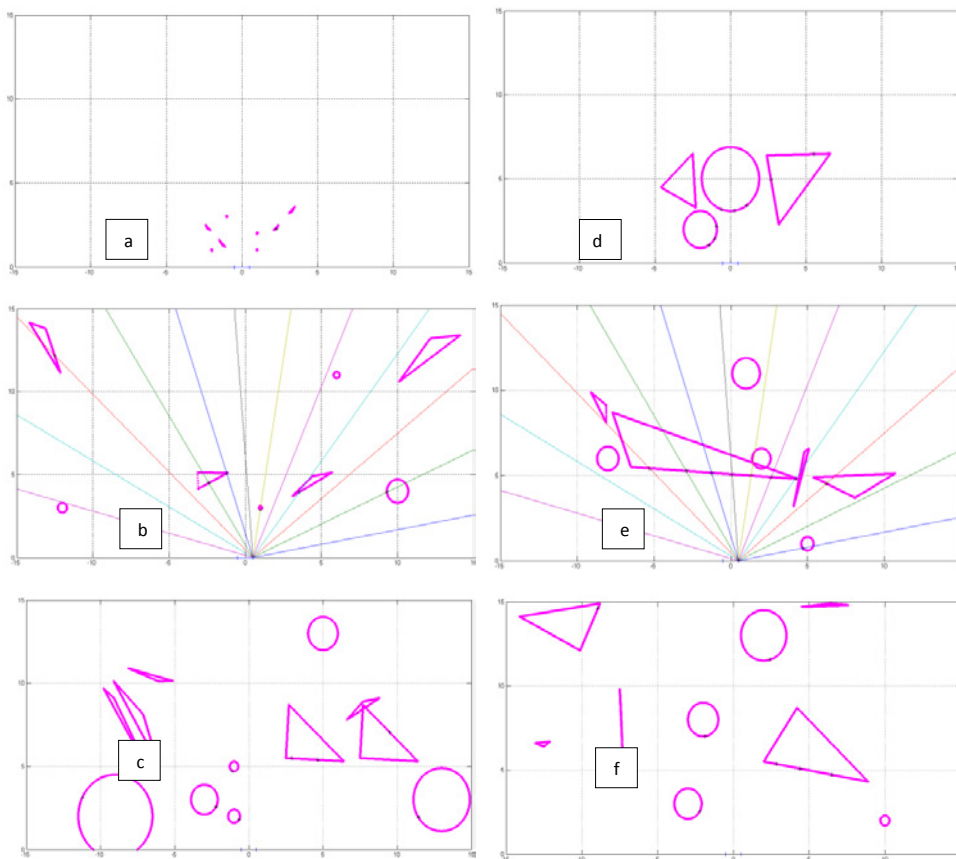


Figure 5: Frames from simulation in MATLAB.

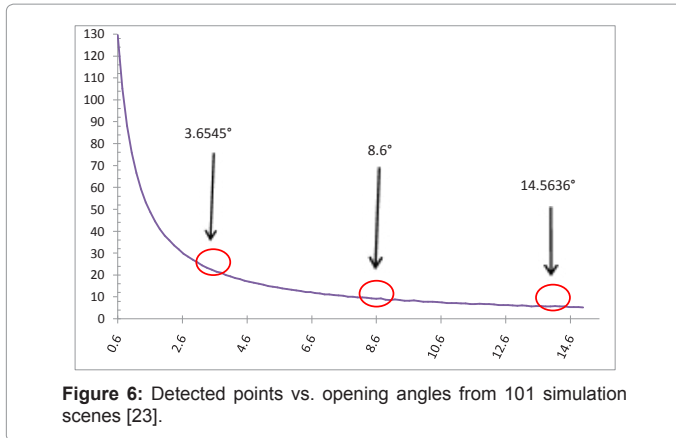


Figure 6: Detected points vs. opening angles from 101 simulation scenes [23].

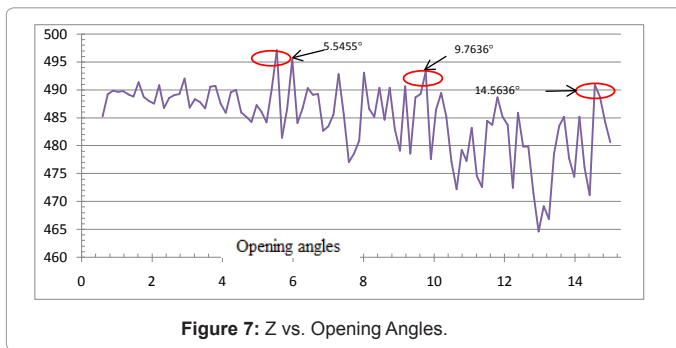


Figure 7: Z vs. Opening Angles.

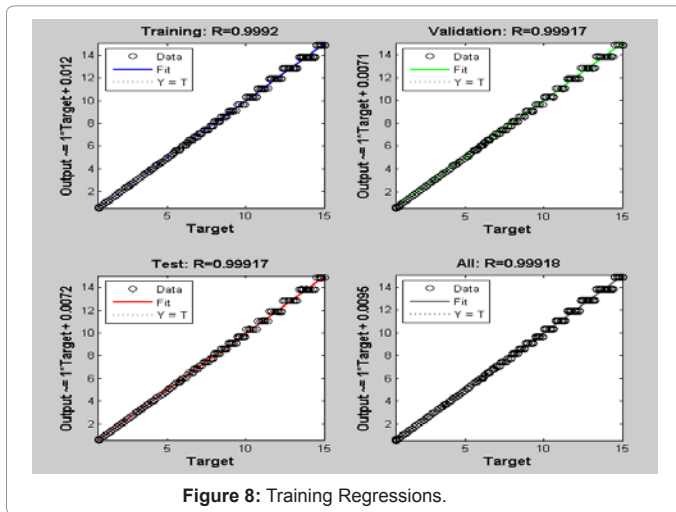


Figure 8: Training Regressions.

information about obstacle is changing with discrete level. Such angles are angles around 1.8-4° (Blue Rows of Table 2), 8.2–8.7° and 14-14.6° (Green Row of Table 2). Such situation is very convenient for us in this particular task. Practically, it means the next.

The biggest angle (around 14°) permits us to scan the scene the fastest possible way, is the biggest angle for all provided 10100 simulations still not miss anyone of simulated obstacles. The angles of bigger value are already not detecting some obstacles. The next selected angle of 8.6° is appropriated for fastest rough scanning, Table 1 shows analyzed data. Angle around 8.6° seems to be the best candidates for optimal scanning, 8.6° gives 4.42 times faster than 1.9091° but still 1.72 times worse than 14.5636°. But at the same time is still 2.32 times

faster than 3.65454°, but slower 1.11 times that 9.7636° see Table 2 for comparing angles.

The last extreme experimentally obtained value of step angle around 1.8-4° provides the slowest scanning; practically it is applicable only for precise 3D measurement on the edge of obstacle under highest interest for collision prevention in the further task of robot trajectory planning.

Levenberg-Marquardt Method application for rectifying of optimal angles

According to Rodriguez-Quinonez et al. [19] the most proper tool in this case is the Levenberg-Marquardt Algorithm, which is in the least squares curve fitting problem: given a set of “m” empirical datum pairs of variables, (x_i, y_i) , optimize the parameters β of the model curve $f(x, \beta)$ so that the sum of the squares of the deviations becomes minimal [3,21].

$$S(\beta) = \sum_{i=1}^m [y_i - f(x_i, \beta)]^2 \quad (12)$$

Levenberg-Marquardt is an iterative procedure. Marquardt’s contribution [19,23] is to replace Levenberg equation by a “damped version”,

$$(\mathbf{J}^T \mathbf{J} + \lambda \mathbf{I}) \delta = \mathbf{J}^T [\mathbf{y} - \mathbf{f}(\beta)] \quad (13)$$

Marquardt [19,23] replace the identity matrix, \mathbf{I} , with the diagonal of $\mathbf{J}^T \mathbf{J}$, resulting in the Levenberg-Marquardt algorithm:

$$(\mathbf{J}^T \mathbf{J} + \lambda \text{diag}(\mathbf{J}^T \mathbf{J})) \delta = \mathbf{J}^T [\mathbf{y} - \mathbf{f}(\beta)] \quad (14)$$

where \mathbf{J} is the Jacobian matrix [19,22] whose i^{th} row equals J_i , and where f and y are vectors with i^{th} component $f(x_i, \beta)$ and y_i , respectively. This is a set of linear equations which can be solved for δ . \mathbf{I} is the identity matrix, giving as the increment, δ , to the estimated parameter vector, β . The (non-negative) damping factor, λ , is adjusted at each iteration.

Opening Angles	Detected Points'	Scanning time (sec)	Resolution	Conclusions comparing with 8.6°
1.9091°	41	0.084	High	8.6° detecting 32 points less than 1.9091° but was 4.42 times faster
3.6545°	21	0.044	High	8.6° detecting 12 points less than 3.65451° but was 2.32 times faster
5.5455°	14	0.029	Medium	8.6° detecting 5 points less than 5.5455° but was 1.53 times faster
9.7636°	8	0.017	Low	8.6° detecting 1 point more than 9.7636° but was 1.11 times more slow
14.5636°	5	0.019	Low	8.6° detecting 4 points more than 14.5636° but was 1.72 times more slow

Table 1: Comparison with scanning step angle of 8.6°.

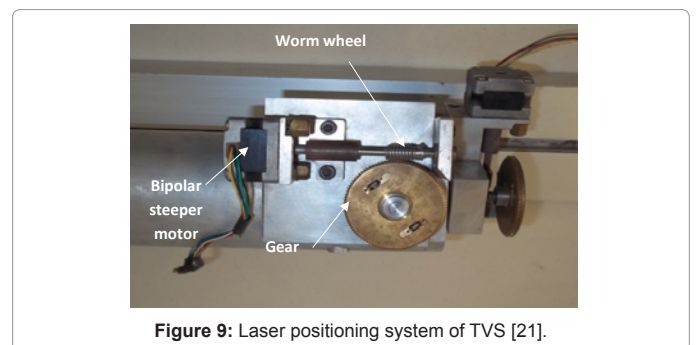


Figure 9: Laser positioning system of TVS [21].

The LMA has been performed on system results. All information from simulation, was concentrated in a matrix of [3×10100], the data was used for preparing the Neural Network (NN) with of the LMA, using the next percentages: the training was 50/100, validation 25/100 (these are used to measure network generalization), testing 25/100 (when generalization stops improving). (NN) regression algorithm is effective for next angle prediction; it can verify which predicted values adjust to real values with 2% error (see Figure 8). Network is adjusted by error, and trained to acquire the constants of the scene components, based in the training it will decide which angle is the best for the next scanning.

Gear box or MET

MET is aimed to change adequately and in shortest time the step angle. Actually system [14,20] has an anti-backlash gear 48 pitch 0.104" face with 20° pressure angle, 96 teeth, P.D 2.0000, gears 303 stainless steel, 2024-T4 Aluminum, hubs 303 stainless steel, the worm wheel has 48 pitch 20° pressure angle, right hand double thread, the material it makes bronze ASTM B21 alloy 464, 12 VDC bipolar stepper motor with step angle 1.8°, holding torque 2100 g-cm, 200 steps per revolution [14,20]. The current design of laser positioning system of TVS is represented in Figure 9.

We were proposed an ideal electro-mechanical system [20] with variable opening angle as a Dual Clutch Transmission (DCT) [24-27]. However, as the present advanced research shows in a difference to [20] that optimal scanning angles are not so fixed, but the DCT permits due to its design only fixed transmission ratio, so we must to apply another principle than in [20]. The design needs especially three angles for three

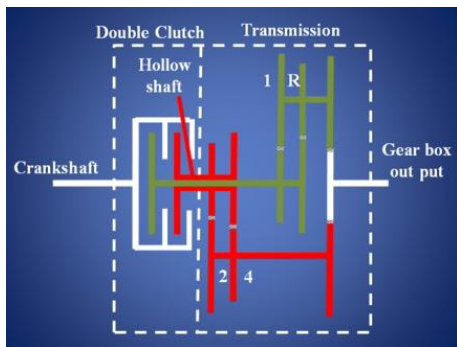


Figure 10: DCT proposed [21].

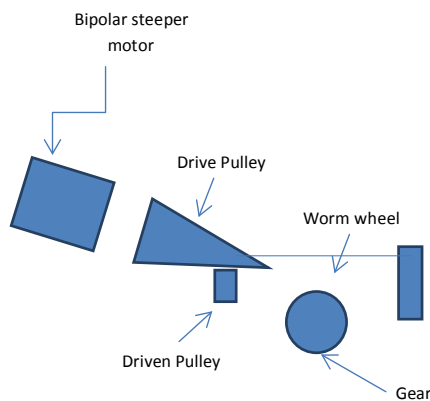


Figure 11: MET proposed [32].

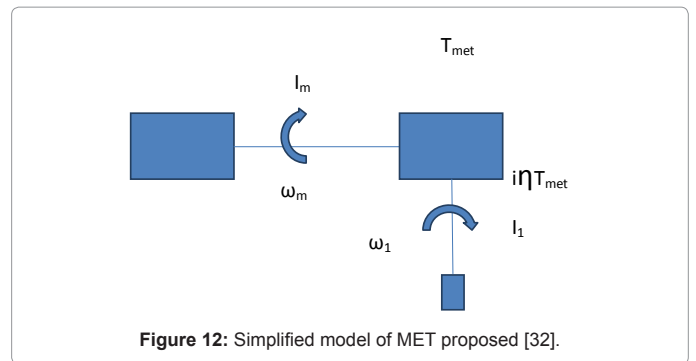


Figure 12: Simplified model of MET proposed [32].

	Optimize time	Using energy	Mechanical maintenance	Weight	Jumps during shifts.	Friction
CVT	More	Less	Less	Less	Lees	Influence of contact-zone between chain links on the torque capacity and dynamic performance [29]
DCT	Less	More	More	More	More	To avoid friction in this type of transmission it need hydraulic pressure [27]

Table 2: Comparison CVT vs. DCT [21].

transmissions ratios supposedly to be that angles simulation brings us. For proposed DCT introduced on Figure 12 [20], the available space for TVS MET is around 0.115 m width and 0.03 m high; this space is not enough for proposed design on, for maintenance this kind proposed of transmission need lubricant for decrease friction, DCT need vary velocity in three rigorously fixed speed changes and reverse, see Figure 10.

Analyzing Continuously Variable Transmission (CVT) vs. DCT (refer Table 2), for TVS application CVT is the better option, CVT is better in optimized time, minor energy use, low maintenance, less weight, and it avoids jumps during shift. For reduced friction CVT is better than DCT. Below will be explained DCT design for actual TVS.

Continue Variable Transmission uses two slots or pulleys that fit their axial widths simultaneously in opposite directions to change the transmission ratio This design has more than a century of use in the automobiles; snow blowers and lawn mowers [21,28], and this long experience in different applications with distinct conditions of use shows that CVT has an advantage: it is strongly reliable and low dependent on surrounding constrains. It is a good property to novel design with unknown practical restrictions. In Table 2 we analyzed conditions and parameters for CVT versus DCT [21].

Based on our data, previous design in research [20], we can decide the next. CVT has various sub-designs. We choose principle of CVT operation mostly matched to our case, for still more economy solution we propose the design without one cone and clutch. The design will put on novel MET type "Cone mesh CVT", in this design use one cone (which be called Drive Pulley) and one shaft (which be call Driven Pulley), for reference see Figure 11 [21].

MET proposed components and give it a representation, in Figure 12 can see simplified model of proposed MET.

We can obtain the dynamic equations of MET from Figure 12 [29]:

$$I_m \omega_m + B_m \omega_m = T_m \tag{15}$$

$$I_1\omega_1 + B_1\omega_1 = i\eta T_{met} - T_1 \quad (16)$$

Where I_m is the equivalent rotary inertia of motor, I_1 is the equivalent rotary inertia of the active pulley of MET, ω_m is the angular velocity of motor, ω_1 is the angular velocity of the active pulley of MET, B_m , B_1 , represent the equivalent damping coefficient of each axis respectively; i is the speed ratio of MET and η is the transmission efficiency. T_m is the output torque of motor; T_1 is the output torque of laser ray drive (equivalent torque of all units sum of laser positioning system).

$$T_1 = F_1 r = r \left[G + \left(m + \frac{\sum I_2}{r^2} + \frac{I_f i^2 \eta}{r^2} \right) \right] \quad (17)$$

In this equation, r is the radius of laser ray drive; G is the weight of all units sum of MET; I_j is the rotary inertia of laser ray drive; m is the mass of rotary parts [29].

According to our design presented on [20,21], the MET must to provide the scanning step angle value changes as soon as possible, and as evident from equation 15 and 16 there are no too much options to provide it. The majority of parameters in equations 15 and 16 are physical constants, even ω_m is the constant in our offer because of the simplest and reliable design of TVS, and so, only ω_1 is the unique candidate under consideration for its variable control function programming. In our opinion, this is another small advantage of our proposition: the simplest controls always are more robust. In fact, we propose to program the changes by system processor of the position of active pulley (see Figure 12) to satisfy the condition that the number value of ratio of controlled ω_1 and constant ω_m will deal exactly value of one angle of 3.65°, 8.6° and 14.6°, as necessary at the moment to provide scan[21].

Conclusion

This paper can offers the solution capable to increase velocity of obstacle detection to mobile robot navigation. Computational experiments in the present paper are obtained values of three particular angles which permit among others the minimal losses of information about scanned objects. That application permits accelerate the process of automatic search of the obstacles within the scene under interest, with posterior precise measurements only those edges of obstacles mostly close to future trajectory of the robot. Multivariable simulation of scenes in this software warranties the high reliability of such optimized scanning angles and the enhanced resolution of fast scanning.

Comparing two kinds of micro transmission design, dual transmission clutch and the conic, in a difference to our previous solution in [20]. Clearly shown by various simulation results that transmission conical more of a practice limits is best matched our practical application requirements.

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