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## RPI TECENICAL REPORT MP-57

PROCEDURES FOR THE INTERPRETATION AND USE OF ELEVATION SCANNING LASER/MULTI-SENSOR DATA FOR SHORT RANGE HAZARD DETECTION AND AVOIDANCE FOR AN AUTONOMOUS PLANETARY ROVER

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## ABSTRACT

To gather more information about the solar system, future missions to Mars should include visits to many remote locations on the planet surface for scientific experimentation. An autonomous roving science vehicle that relies on terrain data acquired by a hierarchy of sensors for navigation is one method of carrying out such a mission. Included in the hierarchy of sensors is a short range sensor with sufficient resolution to detect every possible obstacle and with the ability to make fast and reliable terrain characterizations. A multi-laser, multi-detector triangulation system is proposed as a short range sensor. The general system is studied to determine its perception capabilities and limitations. A specific rover and low resolution sensor system is then considered. After studying the data obtained, a hazard detection algorithm is developed that accounts for all possible terrains given the sensor resolution. Computer simulation of the rover on various terrains is then used to test the entire hazard detection system.

## I. INTRODUCTION

The use of remote sensors to explore the solar system has contributed much knowledge in the search for answers to numerous questions concerning, particularly, the origin of the universe and the existence of extraterrestrial life. Considerable effort has been focused on exploration of the planet Mars and successes achieved to date represent a great achievement. As with any good scientific investigation, however, more new questions have been raised than resolved. To answer these new questions, an extensive surface exploration of Mars should be undertaken. Regardless of the type of experiments to be performed, a thorough investigation of the planet surface should involve visiting many sites on a trajectory of several hundred kilometers.

One method of conducting widely separated experiments is to construct many sets of scientific equipment and to land one set at each site. While this plan is feasible, it requires much duplication of effort and hardware. Another alternative is a mobile science station that can visit every site. The time required to visit every site is now an important consideration. A vehicle that cannot deal with a wide variety of adverse cerrains will have few traversable trajectories available. A higiner mobility vehicle, on the other hand, can take advantage of more direct yet possible adverse terrains thereby minimizing travel time and maximizing science time. One suggested vehicle is a "tumbleweed" that is blown across the planet surface by wind. It is not likely, though, that the tumbleweed will reach all of the desired sites by chance. Furthermore, even such a high mobility vehicle can get permanently lodged in one location. For these reasons, the
vehicle to be used should be controllable.

Selection of a desired path for a roving vehicle should proceed on several levels. Obtaiaing an overall view the .ain to be traversed is a good first step. Without such information, the situation is similar to going on vecation without a road map. unnecessarily long route will probably be taken. To gather information over several hundred kilometers, an orbiting sensor is a good choice. Due to a resolution of only 100-200 meters, however, many smaller objects that may be hazardous to the rover are not detected. Therefore, shorter range, higher resolution sensors are required in addition to the orbiting sensor. Compared to the long range sensor, short range sensors will have a higher scanning frequency and will require a higher frequency of path selection decision. Given long round trip communications delay times of from nine to forty minutes and limaced "windows" during which information can be transmitted, direct earth control of the vehicle is not a routine matter. Most of the path selection decisions should originate on Mars. A manned mission to Mars is a possible solution but $s$ made difficult by the long duration of the mission, higher risk involved, additional payload required, and a necessary return trip. An autonomous roving vehicle with onboard short range sensors can provide a simpler alternative,

One possible strategy is to first plan a rough path to be rraversed that avoids major terrain features based on orbiter sensor data, Televisior cameras on board the vehicle enable controllers on earth to locate landmarks and to choose intermediate targets along the route, Finally, a short range sensor with appropriate software is employed to :eer the vehicle safetly from one target to the next.

## The focus of this paper is to investigate the short range sersor conceft. A general sensor scheme is proposed and its characteristics are analyzed. For a particular given sen:or, a terrain modelifg algorit'm is developed and is tested and evaluated by means of computer simulation.

There are several shorter range sensor systems currently '.'ng studied. Techniques such as TV lmaging and laser range finding are being developed for use over a range of about fifty meters. With these systeris the vehicle stops, a scan is taken, the data are processed, a path is selected, and the vehicle moves along the desired trajectory. The only problem is that to get sufficient resolution over a fifty meter range, a large quantity of data and, therefore, much data processing time are required. The resuit is a vehicle moving only on the order of 400 meters per day. With a mission covering a distance several orders of magnitude greater, time becomes $2 n$ important issue.

A better solution might be to maintain the same resolution while shortening the range to a few meters and, thus, decreasing the amount of data to be prosessed. With the increased scan and decision rates made possible by this short range system, vehicle speed can be greatlv increased. The only drawback is a very limited field of view The midrange sensors with a fifty meter range have a hagh probability of choosing as direct a path as possibi: over the fifty meter view. A short range sensor that can see only a few meters at a time will not necessarily choose the optimal trajectory. The short range system will prove to be superior if the effect of increased speed exceeds the effect of longer trajectories so that the overall vehicle displacement over time is increased. The incres.se in velifcle speed using a short range system is estimated to be at least one orier of magnitude by virtue of the enormous data reduction. The midrange sensis arf nu: expected to give a comparable reduction in path length. Even though a shurt ringe sensor is not likely to choose the optimal trajectory for a trip of several hundred kilometers, a fifty ueter
mid range scanner will probably not do much better.
Use of a short range sensor does not automatically rule out techniques such as IV or laser range finding. TV pictures do present a time problem, though, due to the extensive inige enhancement techniques that must be applied. Decision rates on the order of one per second are necessary for a short range system to be feasible, While laser range finding data can be processed quickly enough, the technique is more difficult to implement over short ranges because of the increased difficulty in measuring time of flight.

An easier technique for obtaining accurate measurements with a laser over short ranges is triangulaticn. The system consists of a laser located on a mast with a laser detector located at a known separation, Figure 1. When the laser is pulsed, a short segment of the beam intersects the detector field of view. Knowing the nointing angles and locations of both the laser and detector, the location of the line segment of intersection can be determined by trigonometry. If a laser pulse strikes terrain lying within the detector field of view, scattered light is sensed. The terrain is then known to lie somewhere along the line segnent of intersection. If no return signal is received, the terrain is assumed to lie elsewhere. By proper choice of pointing angles and derector size, a systen can be developed in which a return indicates the presence of safe terrain while no return means a hazardous path. Rotating the mast enables the scanning of sever.: $1:$ fmuths in search of safe paths, Figure 2. Such a one laser, one detector system scanning fifteen azimuths at ten degree spacings has been tested at R.P.I. ${ }^{1,2}$ The vehicle operated subject to assumptions that only terrains involving gradients of less than




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FIGURE 2
thirteen degrees and step obstacles with less than twelve inches differential were safe. All other terrains resulted in no return signal and were assumed hazardous. The surprisingly gocd performance obtained represents a remarkable achievement since this is the only hazard detection system to be successfully field tested. While past investigations have demonstrated that the laser/ detector triangulation system can work, it is not clear that other systems will not perform as well. More sophisticated systems have a better chance of choosing more direct trajectories. If the paths chosen by the single laser/detector system are too erratic, the higher vehicle speed may not be enough to offset the eifect of longer trajectories. The path selection problems are due to uncertainty. Any terrain is characterized by truly safe paths and unsafe paths, Figure 3. The job of the hazard detector is to choose the most direct path from the safe paths available. As with any real system, $t^{\prime}$ re is always some uncertainty added. The one laser, one d. :oryor system bases its decisions on very little information and as a result, many terrain classifications are unsure. With the success or failure of the mission dependent on avoiding dangerous situations, 311 unsure terrain must be classified as unsafe. The result is a smaller number oí safe paths available and a reduced ability to select a more direct path. A more accurate system is characterized by less uncertainty and more available paths. Clearly, if the laser/detector short range triangulation hazard detector is to be competitive, the amount of uncertainty associated with each scan should be reduced.

The main source of uncertainty in the single laser/detector system Is the low quality and accuracy of data. Additional laser firings and shorter

Real Terrain


Single Laser-Detector Systen


Ideal Sensor System


Multiple Iaser-Detector System


FIGURE 3
$\cdots$ Terrain Classifisation
line segments of intersection would give a better terrain ricture. A multi-laser, multi-detector system provides the extra data desired, Figure 4. Other sources of uncertainty, such as instrumentation error, are neglected because their effect on the lengths and positions of the line segments is assumed small. The multi-laser/detector hazard detection system is the short range sensor proposed for an autonomous Martian rover.

III. THBORY

Figure 5 illustrates the generalized multi-laser, multi-detector system with a boulder in the field of view. The data obtained are shown as the darkened line segments. Note that while it is obvious from the data that a bump accurs in the terrain, the actual contour is not clear and a wide variety of terrain features is possible, Figure 6, In general, a single scan does not necessarily produce data that define unambiguously a particular obstacle. It is helpful to learn what the perception capabilin ties of the system are and which parameters can improve perception.

Probably the greatest limitation on the perception results from the discreteness of system. In order to perceive desired features or terrain fluctuations, there is a minimum data spacing or rate of sampling that must be observed. There is a direct analogy tc sampling of electronic signals. In theory, a signal that has a finite bandwidth can be uniquely reconstructed after sampling if the sample rate is at least twice the highest frequency contained in the signal. What this means for terrain sensing is that if high frequency or highly fluctuating terrain is ignored, the original terrain can be uniquely reconstructed from sensor data by choosing the sampling rate sufficiently high. To some degree the assumption that rapidly fluctuating terrain does $\mathrm{n}^{+}$exist is a valid one since features such as spikes, poles, or tree trunks are not likely to be found on Mars.

Unfortunately, the theory discussed above assumes that the signal is well known at the sample points. This is not necessarily true for the triangulation sensor. The sample points are really line segments, Even if high frequency terrain perturbations are ignored and if the proper sampling rate is observed, the original terrain cannot be uniquely reconstructed from the data, Figure 7. The most that can be hoped for is to define an envelope



FIGURE 6
Example of ambiguity associated with sensor data

## NMMN



FIGURE 7
Example of ambiguity resulting from uncertainty at sample points
in which the terrain lies，While it is not trivial in practice，in theory it is possible to determine the entire range of terrains possible givon a set of data assuming that high frequenciesare ignored．The size of the envelope can be reduced by observing that extreme terrain fluctuations are not possible between data points since other adjacent detector would have sensed the terrain．Therefore，given the sensor data and some a priori knowledge of the character of Martian terrain，an envelope can， $1=$ theory， be constructer in which the terrain is known to lie．While this is not as good as completely spectfying the terrain location，at least the zossible terrains are bounded and the uncertainty in terrain location is $=\equiv=10 d$ ， One of the keys to better perception is to reduce the $5 x=0.0$ the terrain envelope．A logical conclusion may be that more lasw piases
 Doubling the number of laser pulses does improve the situation $2=\sim$ ニニitional
 seen by the same detector，the pattern of line segments is very sーニニニッred， Figure 8，As the number of laser pulses approaches infinity，$t=\underset{=1}{E=3}$ actually become a series of contiguous quadrilaterals resemblint grams，Figure 9．What is of interest is that the parallelograms defined，Each parallelogram is foined to its neighbors only at
 field of view and two laser pulses．If the locations of all comeroser
 ，on the parallelograms are uniquely defined．The coriclusion ix $=:=0$ of the information obtained from an arbitrarily large number of $5=5$ is totally represented by the location of the $n+1$ vertices wher $=\sim=\sim$


number of detectors, Stace all yertices occur at detector houndaries, the laser need only be fired at those points where the terrain intersects detector boundaries. Knowing the laser and detector pointing angles, the envelope can be constructed, is zontinuous laser that scans for this occurrence will do the job.

The above findings provide an easy method for obtaining an accurate terrain envelope. Frrthermore, it can be concluded that each additional laser contributes a decreasing marginal increase to the amount of information. There is an upper limit to the amount of information and, hence, a laser limit to the uncertainty associated with a given number of detectors and an arbitrary number of lasers. The amount of information available from a finite detector system can be maximized without increasing the amount of data that must be processed.

Just as decreasing the laser spacing gave improved perception, similar benefits should be obtained by decreasing individual detector sizes, Smaller detector fields of view shorten the line segments of fintersection and decrease the uncertainty associated with each measurement, Wherever data are taken, the location of the terrain is known more accurately. The effect is to decrease the size of the terrain envelope, In the limit as the detector fields become infinitessimally narrow, the line segments of interaction are reduced to points, Figure 10 . This is the case of ideal sempling where the terrain is uniquely reconstructed from the data when the proper sample rate is observed.

From the above analysis it is now known that laser density determiner che sampling rate and, thus, the number of data points while the detector deri.. , As the data accuracy, It is possible, although not ain $; \quad . \quad$ fine an envelope that bounds the set of possible terrains

given a set of data. It has been shown that an accurate envelupe can easily be generated by using a continuous laser that scans the terrain for detector boundaries. Furthermore, this system represents the optimum usage of the laser since the maximum amount of information is extracted with a minimum of data. The use of very narrow individual detector fields with a sufficient number of laser firings yields data from which the actual terrain can be uniquely reconstructed.

While the conclusions drawn look promising, there are practical considerations that cannot be overlooked. The main thrust of the analysis has been to account for all of the possible terrains that may have given rise to a set of data sinee, for safety's sake, even the improbable terrains cannot be ove, looked. A terrain envelope performs this function and, in theory, one can always be generated. Except for a few spectal cases, though, nc method has been developed for generating these terrain bounds in the general case. Even in the two special cases of narricw laser spacIng and narrow detector fields, the validity of the expected terrain envelope breaks down. In these cases the measurements being made are so fine that the assumption that measurement error can be neglected no longer holds. The message here is merely a reminder that what can be done in theory is not always true in practice.

There are other considerations to be made when specifying a sensor system. The overall detector field of view is of critical importance. One reason for a short range sensor is to maintain high resolution with a small number of individual detectors. Yet, there is a lower limit to the size of the field. The field of view mus: be wide enough to see sufficiently large sections of obstacles so that meaningful decisions can be made. Another constraint is that, particularly under large vehicle pitch conditions, rapidly rising or falling terrain may fall outside of the sensor


#### Abstract

fiald of view, Figure 11. The location of the lasers and detectors is also a factor, With the laser and detector clusters restricted to belig on t'ie same mast, better results are obtained with a larger separacion' stween the aser and detector clusters. Increasing their separation increases the angles between laser beans and detector fields and decreases :he lengths of the line segments of intersection, Figure 12. There are practical limits to the degree of separation. The laser height cannot exceed the mast height while the detectors must be high enough to clear t!e ground,




FIGURE 11
Swall sensor field loses sight of terrain


## IV. THE ELEVATION SCANNING LASER/MULTI-DETECTOR CONCEPT

Having gained some knowledge about general multi-laser, multidetector sensors, attention is now focused on the specific case of the elevation scanner laser system under development at Rensselaer. The geomatry is identical to the general case but there are some very important parameter constraints. For simulation purposes, the lasers are placed at the top of the mast at a height of 2.0 meters and the detectors are located at a height of 1.0 meter, locations which compare well to actual vehicle dimensions. The lowest laser and lowest detector are aimed to intersect level ground at 1,0 meter. This distance is chosen because obstacles closer than 1.0 meter cannot be safely avoided without a backup maneuver. The laser firings can be variably spaced but must have an average separation of at least one degree, Only twenty detectors with equal fields of view are available and this represents a major prablem. A tradeoff must be made between resolution and overall field of view. The vehicle encounters terrains varying in slope from $-30^{\circ}$ to $30^{\circ}$ and will need a $60^{\circ}$ field of view to deal with the most extreme situations, Using proper optics the $60^{\circ}$ field can be obtained but the individual detector fuelds must be $3^{\boldsymbol{p}}$ each. The resolution possible from such a system is not sufficient to detect certain obstacles, More will be said on this issue later. As a compromise, $2^{\circ}$ detector fields with a $40^{\circ}$ overall field of view. are chosen for simulation. To maintain a "square" array where laser and detector densities are equal, there will also be 20 laser firings of $2^{\circ}$ incraments. A base design has now been developed with which experiments can be undertaken, Figure 13.

The remaining task is to develop an algorithm for interpreting

sensor data that, given the uncertainty inherent in the system, accounts for all possible terrain features. Unfortunately, none of the earlier findings can be applied in this case. The $2^{\circ}$ individual detector fields result in such large line segments of intersection that ideal sampling and unique reconstruction of the terrain is not possible. In many cases, the lengths of the line segments equal or exceed the dimensions of obstacles. A terrain envelope can be easily generated using a continuous laser but this is not possible given the hardware on the R.P.I. rover. A continuous laser cannot even be reasonably approximated by the particular pulsed laser being used. The reason is one of insufficient power dissipating capability. To achieve an acceptable signal to noise ratio of laser to ambient light each laser pulse must be of a certain minimum power. This power level is large enough compared to the laser's power rating that the maximum allowable pulse rate must be kept low. Clearly, another method must be developed for bounding the terrain. Furthermore, the method must be kept simple given the additional constraint of iimited computer support available to the R.P.I. vebicle.

## v. DATA PROCESSOR FOR HAZARD DEIECTION

A method for processing laser data is suggested by analyzing the raw data. In the typical return matrix from a single azimuth scan, Figure 14, each column represents the result of the firing of a single laser. The position of the number " 1 " in a column indicates the detector that received a return after the laser pulse. The number "2" is used instead of " 1 " if the return fell on level ground in the context of a vehicle fixed coordinate system. The " 3 "'s are inserted as a reference line indicating the returns that would have been received if the terrain had been level. By taking the difference in position between the actual data and base terrain data, the measurements representing the terrain in a given azimuth can be reduced to a diagonalized return, ${ }^{3}$ Figure 14. This set of data gives an indication of the level of the terrain above or below level ground. Application of the diagonalized return concept to the R.P.I. system defined earlier is somewhat misleading. Notice that the line segments of intersection fall into curved bands, Figure 15. The diagonalized return concept would be more useful if the data fell into straight, horizontal bands that corresponded directly to the diagonalized return levels. Such a system can be achieved by proper choice of laser and detector pointing angles, The particular detector system to be used, however, is csi:strained to uniform spacing, A good approximation to horizuinti eveis: is possible with evenly spaced detectors by aiming laser pulses at the center of intersection of the individual detector fields with level ground, This modified arrangement which replaces the original system is defined as a quasi-linearized array, Figure 16. Notice that the uppermost laser pulses are of little use since they intersect the detector field at too great a


FIGURE 14
Typical sensor return matrix and corresponding diagonalized return


range. For this reason, these higher elevation angles are umitted leaving 1j laser pulses and 20 detectors, Figure 17.

Scanning an arbitrary taxala with the quasi-lnnearized array reveals that the zerrain data are actually quantized by the roughiy horizor.tal discrimination levels, Figure 18. Regardless of the contour of the terrain, each set of data maps the terrain into a set of steps. Any one of the possible patterns can be completely ani uniquely described by the location and magnitude of the reps.

Give: the well defined patterns, a possible data interpreting scheme might be to associate a particular pattern with a particula terrain feature. There are some problems tc be dealt with if this idea is attempt d . First, there is not a one-to-one correspundence between terrains and patterns. With an infinite number of possible ter-ains but only a finite number of possible patterns, each fittern represerts an infinite number of terrains, Even though the set of patterns ; finite, there ase very many of them. Attempting to match up ? given data set with cne of a large set of st :ed patterns can be a great bookioeping and seurching task. Finally, the a-lsilinearized array is by no mears iniform. The patterr associated with a particular terrain feature varies with the relative position of the feature within the array. In spite of these difficulties, pattern recognition may still be a viable solution and is left open to other investigations.

Before attempting to derive an algorithm to process sensor data for hazard detection, a clear and concise definition of a hazard must be developed. A generalized definition is desired for two reasons. First. the simpler the definition is, the simpler is the task of analyzing the data for hazards, Second, the definition cannut be so specific trat it requires more information than is available from the data. To obtain some


crite for identifying obstacles, the vehicle's mobility characteristics must be considered. From tests it is known that the vehicle can climb a maximum of a $30^{\circ}$ slope and descend a $-30^{\circ}$ slope. The step climbing ability is limited to a height equal to the wheel radius of 0.25 meters. It is assumed that the same limit applies to negative steps. Restating the above more simply, any feature whose vertical height exceeds 0,25 meters and whose slope magnitude exceeds $30^{\circ}$ is a hazard. These criteria have been established for a vehicle on level ground but do not necessarily hold If the vehicle is pitched. When the vehicle's inpath slope exceeds $20^{\circ}$, additional positive obstacles cannot be tolerated, An analogous rule applies to the negative case. The crude rules thus presented form an initial point for investigation. The criteria ..se simply defined and apply to all terrain features,

Having defined what an obstacle is, all that remains is to extract the desired information from the available data. The test for critical height can be easily done since the array is organized into essentially horizantal height levels, There are, of course, restrictions to be placed on the site of the quantization bands. If the levels are chosen to be so large that they exceed the critical step height, then significant terrain features cannot be detected, By choosing the levels to be sufficiently small, a critical altitude change is revealed as a step change in the data. Thus, a relatively simple way of testing for possible obstacles. is to scan the diagonalized return for level changes.

There is, however, some ambiguity possible, Figure 19, While the figure shows two sbstacles of different heights, the diagnolized return is

and indicat ss two ohjects of identical hetght, The coasseness of the quantization does not allow a better distinction. If the larger object is a hazard but the smaller object is not, then a serious problem exists. Any terrain perturbation, regardless of size, that crosses the boundary between two quantization and causes a level change in the diagonalized return is an unsure case and must be classified as unsafe. Since the probability of any arbitrary but safe terrain crossing a quantization level is quite high, almost all terrains would be viewed as hazardous. To remedy this situation the restriction must be imposed on the width of the quantization levels that a 0.25 meter terrain rise will result in at least two level changes in the data. Unfortunately, it is difficult to meet this condition with the twenty detector array, Due to the geometry of the array the lasers, detector $f f^{-1} A s$, and discriminacion levels diverge as the distance from the vehicle increases, Figure 20. To impose the condition that the width of two adjacent discrimination levels be no more than 0.25 meters even in the most distant areas of the array would require very narrow individual detector fields and thus, a prohibitively small overall field of view, The only alternative is to compromise by satisfying the condition only in the area near the vehicle, Figure 21 . The restriction creates a very myopic vehicle that will accurately detect only large obstacles at large distances. Perception improves as the vehicle approaches until a range is reached where perception accuracy reaches desired levels. To insure that every terrain section is examined in the accurate area, the rover displacement between successive scans must be made sufficiently small.

The inherent discreteness of the array has so far been only a source of problems. Yet, there is one benefit derived from discrete data, The quantization of data "filters out" small terrain perturbations in much


the same way that digital communication systems are useful in reducing noise in a signal. Only those features that are large enough to be possible hazards will show up in the data. Any terrain that falls entirely within a single discrimination level is safe and need not be scanned for hazards. This is a big benefit since obstacles are automatically revealed without computational effort.

Having addressed the problem of locating critical heights, the next step is to find a method for determining slopes. The magnitude of a slope is reiated to the spacing of level changes in the data, A rapid succession of level jumps suggests a steep slope while widely separated jumps mean a much gentler terrain. Unfortunately, the data levels have finite width so that a diagonalized return does not uniquely specify a single slope but, rather, a small range of possible slopes, Figure 22 , In order to exactly specify this range of slopes, the upper and lower bounds must be computed. The upper bound is useful because it represents the absolute worst possibility. This is important for a Martian vehicle since no risk can be taken. The lower bound is useful for resolving some ambiguous cases and, thus, providing for better decistons.

Suppose, for instance, that the range of slopes calculated for a particular terrain feature is $25^{\circ}$ to $35^{\circ}$. A possible hazard exists since the maximum slope exceeds $30^{\circ}$. However, there is also the possibility that the terrain is safe since the slope could be as low as $25^{\circ}$, This case is ambiguous. Suppose now that slopes of $35^{\circ}$ to $45^{\circ}$ are estimated for another feature. The upper bound of $45^{\circ}$ again indicates a possible hazard, In this case, though, the mintmum slope also exceeds $30^{\circ}$ indicating that a hazard definitely exists, A simple procedure for calculating the maximum and minimum slope estimates is presented here. Figures 23 and 24 have been included to aid in the explanation of the procedure.


FIGURE 22
Range of slopes possible with given sensor data


FIGURE 23
Technique for determining maximum slopes
43.


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FIGURE 24

## Maximum Slope

1. Determine that a possible height differential of 0.25 meters exists.
2. Determine the coordinates of the lower endpoint of the first line segment after the first jump and the higher endpoint of the last line segment before the last jump.
3. Compute the slope using these two points.
4. For negative features, select the same line segments but use the opposite endpoints.
5. If multiple or consecutive jumps occur, select the lower endpoint of the line segment before the jump and the higher endpoint of the line segment after the jymp.

## Minimum Slofe

1. Determine that the least possible height differential exceeds 0.25 meters. This is done to make sure that any slope calculated rises above 0.25 meters. Otherwise, the slope is not hazardous regardless of how steep it is.
2. Determine the coordinates of the higher endpoint before the first fump and the lower endpoint after the last jump.
3. Compute the slope between the two points.
4. For negative features, select the lower endpoint before the first jump and the higher endpoint after the last jump.
5. The procedure does not change for multiple or consecutive jumps.

These methods yield the least and greatest slopes possible that intersect every line segment in the area in question. They are also easily implemented. The locations and magnitudes of level jumps are known from the dia: nalized return. The endpoints of all line segments in the array can be computed by geometry and stored for easy access when needed. Since the slope calculations involve just two points, the arithmetic is minimal.

So far, the step and slope criteria for obstacle detection have
been considered. The remaining case is the decreased climbing capability
when the vehicle pitch exceeds $\pm 20^{\circ}$. This test is easily done because the vehicle attitude is readily available from onboard gyros. If the vehicle pitch exceeds $20^{\circ}$, then any positive jumps are assumed impassable. Similarly, negative iumps are impassable if the pitch is less than $\mathbf{- 2 0}$. All of the obstacle criteria defined earlier have now been treated. There is, however, another possibility to be considered. It is possible that a laser shot will not be seen by any detector. This can occur if the scattered light is blocked by an obstacle before it reaches a detector. In this case, no data is received and it nust be assumed that a deep crevasse exists. Fortunately, missed returns provide some information. Based on the number of consecutive missed returns the size of the hole can be estimated. If the hole is large enough for a wheel to fall into, then the path is unsafe, Figure 25. However, just because the gap is small, safety is not guaranteed. Several missing returns can also signnify a sharp, hazardous drop. To account for this possibility, the difference in terrain height before and after the missed returns is computed, Figure 26. Of course, when the missed returns occur at either end of a scan, the height of the terrain is not known on both sides of the missed data. If the closest laser shots are not seen, then the vehicle is close to potential obstacle but can no longer see the entire feature. To deal with this case, it is assumed that the whole feature was seen in a previous scan. Since past scans id not detect an obstacle, the terrain is considered safe in spite of missed returns. Missed returns can also occur at the far end of the scan. This possibility raises another important issue, Ofteu a possible obstacle is detected at a distance but there is insuffic: . Information to make a definite decision. In the case of missed returns, a single one at the far end of a scan may signal the leading edge of a crevasse or fust a small, traversable depression, An example of another ambiguous case occurs shen a distant object is determined to have a range



FIGURE 26
Interpretation of small number of missed returns
of slopes from $25^{\circ}$ to $35^{\circ}$. In both of the above cases, caution should be exercised since the terrain is potentially hazardous. To turn away fmacdiately, however, is not a good idea because many false alarms can occur. This is particularly true due to the poor accuracy of the data at long distances. The obvious solution is to get closer and to take a better look. The R.P.I. vehicle has a scan rate fast enough to give five different views of the same terrain as the vehicle approaches. Taking five scans increases the chances of resolving the ambiguity. Naturally, there is a limit as to how closely the vehicle can safely approach an obstacle. In this system, the limit is set at 1.4 meters. The strategy is, therefore, to approach an obstacle until either a definite decision is made or until the obstacle is wit: in 1.4 meters in range.

Until now, all of the obstacle detection has been done in the vehicle frame of reference. The reason for doing the analysis this way is simplicity. The coordinate transformations required to convert the data from the vehicle to the planet frame require additional calculational effort and time. After that has been done, the benefits of the horizental quantization levels are lost. However, the step and slope climbing ability are related to gravitation and only have meaning in the planet frame. The solution is to convert all of the computer terrain slopes $t$, the lanet frame by simply adding in the vehicle attitude. This is much simpler and faster than doing the transformation before the slopes are computed. The hazard detection algorithm is now complete and a general flow chart appears in Figure 27.


## V. SIMULATION PACKAGE

The multi-laser/detector triangulation sensor and the accompanying hezard detection algorithm are to be tested using the R.P.I. dynamic simulator ${ }^{4}$. The dynamic simulator is the result of several years of effort and accurately represents the scanning, decision making, and motion of the actual vehicle on specified terrain surfaces. The user can choose from among a number of arailable general terrain surfaces including slopes, hills, and sine waves. Discrete obstacles such as boulders, craters, and steps may be added to the general terrain surface. There is also the provision for simulating rubble and small rocks on the surface as a noise function.

The user may also choose from a variety of sensors and is free to specify the placement, size, and geometry of each. There is a choice of data processors and path selection algorithms to interface with the various sensors. The measurements made by the sensors can also be contaminated by noise if so desired. The user can also control the physical dimensions and dynamics of the vehicle.

After the user specifies the initial and target locations, the simulation package takes over. Sensor scans are taken at user prescribed intervals after vehicle attitude information from the gyro subroutine adjusts the sensor position position. A terrain model is developed and the best path is selected based on the vehicle's position relative to the target and the surrounding hazards. Control then passes to the motion routine and the vehicle is moved at a rate and for a duration given by the user. The cycle then repeats after this point.

The simulation terminates when either the target is reached, the allotted time is exceeded, or the vehicle finds no safe paths available. At this time, the performance is evaluated be ${ }^{-}$on path length, trip
duration, and the number of close encounters with hazards, finally, maps are printed out showing the terrain and the vehicle trajectory.
VI. SIMULATIONS AND RESULTS

Four groups of simulations have been conducted each designed to test the sensor's ability to detect various obstacles under various conditions. These are summarized in the table below.

## Simulations Performed

## I. Vertical Steps

A. 0.2 meters high
B. 0.3 meters high
C. 0.4 meters high
II. Smooth Slopes
A. Twenty degrees
B. Twenty-five degree magnitude

1. Positive slope with 15 laser, 20 detector system
2. Positive slope, same sensor but field of view aimed closer
3. Negative slope, original system
C. Thirty degree slopes
4. Original 15 laser, 30 detector system
5. 25 laser, 30 detector system
6. 32 laser, 40 detector system
III. Sine Waves
A. 0.25 meter amplitude, 6,0 meter period
B. $\quad 0.3$ meter amplitude, 6.0 meter period
C. 0.4 meter amplitude, 6.0 meter period
IV. Boulder-and Crater Field

In the first group of simulations, vertical steps of various sizes are placed in the vehicle's path. The purpuse of these tests is to determine the vehicle's ability to detect changes in terrain elevation. When a change in height of 0.25 meters is detected, the slope of the leading edge of the step is computed. No calculations are done for small steps. The smallest step size worth considering is 0.2 meters. According to the
algorithm, an object must create a change of two levels in the diagonalized return before it can be considered hazardous. A drawing of the $15 \times 20$ array shows that this does occur when a 0.2 meter step is within a 1.4 meter range, Figure 28. The simulaticn verifies this. Initially, only zeroes and ones appear in the diagonalized return and the terrain is considered safe. It is not until the scan at one meter range that returns occur in the second level. A slope of $98^{\circ}$ is calculated for the leading edge and the feature is declared hazardous, Figure 29. Ideally, the path should have been declared safe since the obstacle is below the 0.25 meter threshold. It is the discreteness of the data, not a defect in the data processor that prevents making a more accurate decision. Given the data received the step could have been as high as 0.35 meters or as little as 0.1 meters. Since the error is due to the finite width of the quantization level:, making them smaller is the best way to improve performance. Steps of greater height are also considered. Obviously, these will be intected as hazardous. What is of interest is to note at what range the ciecrsion to avoid the obstacle is made. Due to myopia, the vehicle will see only larger objects at a distance and the smaller ones up close. For a 0.3 meter step, a possible obstacle is detected at 2.2 meters range. However, the data are not good enough at that point to know for certain that a hazard exists. The criterion for such a decision is that the minimum possible change in elevation be at least 0.25 meters. This never happens, though, and the vehicle continues to approach until it reaches 1.0 meters range. Even though the ambiguity is not resolved, the vehicle must turn due to the close proximity of the obstacle.

The last run in this group is a 0.4 meter step. A possible hazard is detected at 2.5 meters range. Again, the data is not good enough to



FIGURE 29
make a decision. The vehicle proceeds until at 1.9 meters range a minimum step of 0.25 meters is detected. The calculated maximum and minimum slopes are $105^{\circ}$ and $60^{\circ}$, respectively. Steps whose heights exceed 0.4 meters will be declared definitely hazardous on the first scan. This would occur at about 2.5 meters in range. In summary, anv obstacle at least 0.2 meters in height is avoided by the vehicle. This includes features from 0.2 to 0.25 meters high that in reality are safe. Large obstacles are seen and avoided at greater ranges than smaller obstacles. The limitations stem from the quantization error inherent in the system. The best way to improve performance is to increase detector density by adding more detectors and reducing the field of view of theindividal detector.

The next group of simulations tests the ability of the algorithm to estimate the magnitudes of smooth slopes. The first run at $20^{\circ}$ is used to demonstrate the performance for an easy case. The vehicle begins on horizontal ground and approaches the slope head on. At 1.8 meters enough information is available to estimate a maximum slope of $39^{\circ}$. As the vehicle approaches, the estimates improve until at the 1.0 meters range the slope is estimated to lie between $17^{\circ}$ and $25^{\circ}$, Figure 30 . The slope calculations for the off center azimuths indicate less severe slopes since those gradients are not as steep. For instance, in the $30^{\circ}$ azimuth the true gradient is $17.3^{\circ}$ and the processor computes the maximum slope as $18^{\circ}$. Also of note is that the minimum slope calculations are generally more accurate than maximum slopes. The reason is that minimum slopes must be computed over longer ranges. This helps average nut some statistically bad data.

As the vehicle continues to approach the $20^{\circ}$ slope, the estimates do not improve appreciably. The maximum slope never exceeds $25^{\circ}$, though,

and the vehicle is allowed to climb. As this occurs, a steadily decreasing slope magnitude is perceived. After correcting the slope estimates for the vehicle attitude, they are similar to the estimates obtained on level ground, demonstrating that accurate estimates can be made when the vehicle has an arbitrary pitch. When the vehicle pitch approaches that of the slope, the perceived slope is so small that no potential hazards are detected, Figure 31. In this situation there is less that can be said about the terrain inclination ${ }_{8}:$ The reason is that the change in terrain elevation seen by the vehicle is less than 0.25 meters. Therefore, no slope estimate is obtained. In this case, the vehicle assumes that if its inpath slope is less than $20^{\circ}$, any terrain that lies ahead is safe. The vehicle then proceeds to climb completely onto the slope at which point it sees flat terrain everywhere. In this case, the assumption made as to the safety of the terrain is correct. The program ignores other possibilities, though, where the logic breaks down. Suppose the vehicle is traveling across the face of a $30^{\circ}$ slope, Figure 32. The inpath slope is $0^{\circ}$ while the crosspath slope is $30^{\circ}$. If a small terrain feature appears on the vehicle's high side, the path in which it lies is considered safe. The reason is that as in the case above, the vehicle inpath pitch is less than $20^{\circ}$. However, if the vehicle cnooses that path, it will be traveling up a gradient in excess of $20^{\circ}$. The vehicle can tolerate absolutely no obstacles in this case and the terrain is actually hazardous. A simpler example involves the vehicle climbing a $40^{\circ}$ slope at a $45^{\circ}$ angle. The inpath gradient is $28^{\circ}$ and safe. Suppose the vehicle wants to make a $-45^{\circ}$ turn. The scan shows flat terrain and the turn is allowed. In reality, the path chosen has a $40^{\circ}$ gradient. To correct this problem, the vehicle roll must also be accounted for. From pitch and roll information the plane
59.



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FIGURE 31
Diagonalized return for $20^{\circ}$ slope, vehicle pitch $10.4^{\circ}$


FIGURE 32
Examples demonstrating importance of incorporating roll infremation in patch inclinalion estimates
in which the vehicle lies can be determined. It is then an easy matter to find the inclination in any azimuthal direction by taking the direc tional derivative. In this way, the estimated slopes can be pr. 4 :, 7 corrected for vehicle attitude.

To test the vehicle under more challenging conditions, a $25^{\circ}$ slope is attempted. The first slope estimates of $41^{\circ}$ occur at 1.8 meters range. As with the $20^{\circ}$ slope, the estimates improve until the vehicle is within 1.0 meters. At this range the computed slope range is from $21^{\circ}$ to $34^{\circ}$. This range of slopes is greater than for the $20^{\circ}$ case. Steeper slopes are generally predicted less accurately than gentler ones. The reason can be determined by observing the data for both cases, Figure 33. With steep slopes the level jumps occur closer together and allow a greater range of slopes. Furthermore, negative slopes are estimated more accurately the nositive ones. This happens because negative slopes are nearly parallel to the laser shots, Figure 34 , and allow only a very small slope variation. A test on a $-25^{\circ}$ slope reveals that the slope estimates have a much smalles variance than in the $25^{\circ}$ case. This does sot mean that, in general, negative terrain is perceived more accurately than positive terrain. This is only a special case of smooth slopes, The data in the lower area of the terrain is very poor and should result in poor estimates for arbitrary negative obstacles.

Returning to the $25^{\circ}$ simulation, the computed slope range of $21^{\circ}$ to $34^{\circ}$ in the center azimuth forcos the vehicle to change course, It must imb the slope at an angle so that the gradient is less steep, Once the vehicle is completely on the slope, it sees level ground and is able to resume its original heading, Figure 35. It is disappointing to see that a


FICunE 33
Example of greater deviations possible with ateeper slopes


FIGURE 34
Slope estimates experience smootret deviations in the negative case
$25^{\circ}$ slope cannot be climbed head on. This would be possible if the data were made more accurate. One way of doing this is to move the field of view in closer. This concentrates the same number of laser shots into a smaller area. To test the effectiveness of such a modification anocher $25^{\circ}$ rus is done with the $15 \times 20$ array. This time the field of view is moved in so that the first laser pulse strikes level ground at 0.6 meter as opposed to 1.0 meter previously. As a result, the vehicle must now appruach even closer to get a good view. With the modified array, no slopes are calculated until the vehicle is within 1,0 meters. At this point the initial estimate is $34^{\circ}$ as compared to $41^{\circ}$ for the initial estimate for the first case. However, that $41^{\circ}$ estimate was made at 1.8 meters range. The previous system predicted a slope of $21^{\circ}$ to $34^{\circ}$ at the 1.0 meter range. This is the same maximum slopa predicted by the modified array at the same distance. The estimate from the modified array improve as it approaches while the original system gave practically the same estimates. At 0.6 meter the computed slope is $31^{\circ}$. This is $3^{\circ}$ better than the original system. Hence, the vehicle must still turn since the $30^{\circ}$ threshold is exceeded. The conclusion is that, at least for st oth slopes, no noticeable improvement results from moving the field of $v .-$ in closer.

Returning to the original $15 \times 20$ system, a $30^{\circ}$ slope is now attempted, Of course, there is no way that the vehicle can climb this slope head on, The simulation is, therefore, done with the vehicle attempting tu climb at a $40^{\circ}$ angle. Systems with 25 lasers $\times 30$ detectors and 32 lasers $\times 40$ detectors are also tested on the same path as the $15 \times 20$. This is done to determine the effectiveness of increasing data density and accuracy, The results are sh swo in Table 1 .


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TABLE 1

## COMPARISON OF DIFFERENT SENSOR SYSTEMS VIEWING $30^{\circ}$ SLOPE



Two sets of slope estimates are given of each location for each system. The maximum estimates are on the upper line and the minimum estimate: are on the lower line.

Multiple estimates are given of each location because the estimates correspond to different sections of the same terrain. They ara arranged in increasing order of range.

The effect of additional lasers and detectors is quite noticeable. With all three systems, the maximum slope estimates always overestimate the true slope while the minimum estimates always underestimate. The difference is that the amount of variation in the estimates is greatly decreased when greater data density is used. Unfortunately, even the finest array gives some statistically bad data, even at close range. However, the probability of receiving statistically bad data is much lower than with the $15 \times 20$. A common way of dealing with statistically bad data is to use filtering or smoothing techniques. This usually requires many measurements of the same signal or object. Even though the laser scanner does not generate a large quantity of data, some first approximations can be made. The higher order systems, such as $25 \times 30$ make several slope estimates over a short section of terrain. If the terrain is assumed not to vary greatly over small distances, then some smoothing can be done. As an example, Figure 36 lists slope estimates obtained from a $25 \times 30$ sensor system scanning a $30^{\circ}$ slope. All of the estimates are based on data from a single scan. Each estimate represents only a small section of the slope and the estimates are printed in order of increasing rancs. In the $-60^{\circ}$ azimuth, the $39^{\circ}$ estimate is the closest and is based on the most accurate data. However, it is also the least accurate estimate of the five calculated in that azimuth. The other four estimates are very consistent and cast doubt on the validity of the $39^{\circ}$ estimate. Furthermore, the range of the $39^{\circ}$ estimate partially overlaps the range of the adjacent $32^{\circ}$ estimate. For these reasons, the inconsistent estimate should not be counted as heavily.

In this example, the slope is still hazardous even if the $39^{\circ}$
estimate is ignored, There may be cther situations in which this technique would have a greater effect,


A technique that will gield better results is to filter data from several scans. Table 2 shows maximum and minimum slope estimates obtained over six consecutive scans made by a $32 \times 40$ system as the vehicle approached a $30^{\circ}$ slope. Along with each estimate is the location relative to the vehicle of the terrain associated with that estimate. In Table 3, the estimates are regrouped by location relative to the planet. Note that there are several slope estimate. at each location. Each maximum slope estimate always overestimates the slope while each minimum always underestimates. Clearly, the best estimates are the least maxima and the greatest minima for each location. These estimates have been selected and placed in Table 4. These "filtered" estimates are a much more accurate representation of the terrain than any set of estimates from a single scan.

The benefits obtained from this technique must be weighed against the computational effort required for implementation. The coordinate transformation of the estimate locations from vehicle to planet frame is easy with a straight trajectory as in this example but is much more complicated otherwise. furthermore, in this example the estimate locations were given as points. In reality, the estimates are taken over finite ranges of the terrain and often the ranges from consecutive estimates will overlap, Finally, in the example above, only data from the center azimuth are considered since those scans overlap when the vehicle is on a straight trajectory. The data obtained from off center azimuths do not overlap as much as in the $0^{\circ}$ case. A more sophisticated algorithm would be needed to deal with this. Even with a complex algorithm, the best results will be obtained in the center nf scan. The implications are that the resulting system will have better central visiion than peripheral vision, Actually,

TABLE 2
$32 \times 40$ system, $30^{\circ}$ slope
Maximum, binimum slope estimates and corresponding range estimates

Vehicle Location

| Planet Frame <br> (meters) |  | Slop <br> Range estimates, <br> estimated (degrees) <br> vehicle frame |  |  |  |  |  |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :---: |
| -1.8 | Max | 43 | 40 | 40 | 84 | 38 |  |
|  | Range | 1.9 | 2.1 | 7.2 | 3.4 | 2.6 |  |
|  |  |  |  |  |  |  |  |
|  | Min | 26 | 29 | 25 | 25 |  |  |
|  | Range | 1.8 | 2.0 | 2.1 | 2.3 |  |  |

$-1.5$
$\begin{array}{llllll}47! & 44 & 33 & 32 & 40 & 83\end{array}$
$\begin{array}{llllll}1.6 & 1.8 & 1.8 & 2.0 & 2.2 & 2.5\end{array}$
$28 \quad 27 \quad 25 \quad 27$
$\begin{array}{llll}1.6 & 1.7 & 1.8 & 1.9\end{array}$
$-1.2$
$\begin{array}{lllllll}33 & 32 & \because & 34 & 34 & 34 & 41\end{array}$ $\begin{array}{lllllll}1.3 & 1.5 & 1.6 & 1.7 & 1.9 & 2.0 & 2.2\end{array}$
$\begin{array}{llllll}27 & 27 & 26 & 26 & 27 & 26\end{array}$
$\begin{array}{llllll}1.3 & 1.4 & 1.5 & 1.6 & 1.8 & 2.0\end{array}$
$\begin{array}{lllllllll}-1.0 & 38 & 30 & 32 & 37 & 36 & 31 & 36\end{array}$
$\begin{array}{lllllll}1.1 & 1.1 & 1.3 & 1.4 & 1.5 & 1.6 & 1.8\end{array}$
$\begin{array}{llllll}28 & 28 & 27 & 27 & 28 & 27\end{array}$
$\begin{array}{llllll}1.0 & 1.1 & 1.2 & 1.3 & 1.5 & 1.5\end{array}$
$-0.7$
$\begin{array}{llllll}34 & 33 & 36 & 35 & 32 & 33\end{array}$
$\begin{array}{llllll}1.0 & 1.1 & 1.2 & 1.3 & 1.4 & 1.5\end{array}$
$\begin{array}{lllll}30 & 28 & 30 & 29 & 28\end{array}$
$\begin{array}{lllll}1.0 & 1.0 & 1.1 & 1.2 & 1.3\end{array}$
$-0.4$
$\begin{array}{lllll}33 & 33 & 31 & 32 & 31 \\ 0.9 & 1.0 & 1.1 & 1.2 & 1.3\end{array}$
$2928 \quad 28$
$0.9 \quad 1,0 \quad 1.0$

TABLE 3
$32 \times 40$ system, $30^{\circ}$ slope
Slope estimates from six consecutive scans grouped by planet frame location

| Terrain Location <br> Planet frame <br> (meters) |  | Maximum Slope Estimates (degrees) |  |  |  |  | Minimum Slope <br> Estimates (degrees) |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 0.0 |  |  |  |  |  |  | 26 | 28 |  |  |  |
| 0.1 | 43 | 47 | 33 | 38 | 30 |  | 28 | 27 | 28 |  |  |
| 0.2 |  |  |  |  |  |  | 29 | 27 | 27 | 27 |  |
| 0.3 | 40 | 44 | 33 | 32 | 32 | 34 | 25 | 25 | 26 | 27 | 30 |
| 0.4 | 40 | 35 | 37 | 33 |  |  | 27 | 26 | 30 |  |  |
| 0.5 | 32 | 34 | 36 | 36 | 33 |  | 25 | 28 | 27 | 29 | 29 |
| 0.6 | 84 | 31 | 35 | 33 |  |  | 27 | 28 | 28 | 28 |  |
| 0.7 | 40 | 34 | 32 | 31 |  |  |  |  |  |  |  |
| 0.8 | 38 | 34 | 36 | 33 | 32 |  | 26 |  |  |  |  |
| 0.9 | 31 |  |  |  |  |  |  |  |  |  |  |
| 1.0 | 83 | 41 |  |  |  |  |  |  |  |  |  |

TABLE 4
$32 \times 40$ system, $30^{\circ}$ slope
Best slope estimates obtained from six consecutive scans

| Terrain Location <br> Planet Frame <br> (meters) | Least Maximum Slope Estimate (degrees) | Greatest Minimum Slope Estimate (degrees) |
| :---: | :---: | :---: |
| 0.6 | - | 28 |
| 0.1 | 30 | 28 |
| 0.2 | - | 29 |
| 0.3 | 32 | 30 |
| 0.4 | 33 | 30 |
| 0.5 | 32 | 29 |
| 0.6 | 31 | 28 |
| 0.7 | 31 | - |
| 0.8 | 32 | 26 |
| 0.9 | 31 | m |
| 1.0 | 41 | - |

this is not necessarily a problem at all since the human eye behaves In the same way.

All of the drawbacks listed above can be overcome at the ce.t of increased computer time. Only further study will tell whether or not the additional computing is justified. In theory, the idea has merit because it makes use of all of the data available. With a system as crude as the one studied here, discarding even a small amount of data can substantially degrade results. The technique has the further advantage that the estimates are done recursively. This greatly reduces the effort required as opposed to storing all of the data in a constantly updated map. Much more storage and calculating time are required with the map method.

Returning to the comparison of the three simulations on $30^{\circ}$ slopes, increased data density does provide better estimates. Better estimates result in more accurate distinctions between safe terrain and hazards and allow the vehicle to travel a more direct course towards its target. In the simulations, the trajectory followed for the $32 \times 40$ is steeper than with the $15 \times 20$, Figures 37 and 38. Furthermore, the trajectory for the $32 \times 40$ could have been steeper still if the vehicle had not been constrained to a heading of $40^{\circ}$. The performance of the $15 \times 20$ system is nonetheless, admirable. Slopes from $-20^{\circ}$ to $+20^{\circ}$ are negotiated with absolutely no problems. More severe slopes can aiso be handled but the vehicle must approach at an angle. The amount of uncertainty inc eases with increasing slopes so that steering commands be,ome more erratic.

Another group of simulations places the vehicle on sinusoidal terrains of varying amplitude. The purpose is to test the ability to deal with a variety of constantly varying terrain situations. While the results



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look good, they are more difficult to interpret quantitatiyely than the previous simulations. This is primarily due to the fact that it is difficult to imagine what the vehicle sees during a scan under arbitrary pitch and roll conditions. Also, an error has been found in the simulation subroutine that adjusts the sensor attitude for vehicle roll. Subsequen simulations will be affected to some degree by this defect.

Three simulations are performed on a sine wave with 6.0 meter period and amplitudes of $0.25,0.30$, and 0.40 meters. These are chosen to duplicate simulations performed earlier to test the previous one, two, and three laser/detector systems. 4 In all cases the new $15 \times 20$ system performed as well or better than previous systems, as would be expected. The run for 0.25 meters amplitude poses no problem at all as the vehicle is able to travel a direct heading. The maximum computed slopes are ${\mathbf{~} 22^{\circ}}^{\circ}$ which ci mpares to true maximum lor 31 slopes of $\pm 14,5^{\circ}$.

The 0,3 meter af.plitude case reveals an additional problem, The primary source of trouble is a small field of view. As the vehicle scans down from a crest into a trough, the closest laser shots are not seen because the terrain falls below the detector field, Figure 39, The algorithm is set to identify four or more consecutive missed returns as an obsacle. It is total lack of data, not porr data, thet causes the vehicle to turn. Laser shots are also missed at the far end wht. Lerrain rises rbove the detector field. This occurs especially when the vehicle is pitci:ed dosmward.

The roblems with missed returns become more pronounced in the 0.4 meter case. I: addition to iaser shots hitting outside of the detector field, the terrain falls off so quickly nor (maximum of $\pm 22,5^{\circ}$ ) that some 1-ser shots aimed into troughs are blocked by the slope, Figure 40 . While the number of missed returns due to this is not surficient to be hazardous, greater amplitudes will no doubt be fmpassable as this effect will worsen,



Furthermore, the maximum slope estimates at close range are approaching $30^{\circ}$ and are just barely acceptable. In spite of this, both the 0.3 and 0.4 meter cases are traversed with nearly straight trajectories, Figure 41. Since the trajectories are almost straight, the vehicle roll is is small and the error in correcting the scanner alcitude is,likewise, small. The simulation results are, therefore, reliable. The last simulation tests the $15 \times 30$ system on a field of shallow boulders and craters, Figure 42. Mosi of them are small e. jugh that they can be safely traversed. They are merely added $t$ : terrain uneven and cause variable pitch and roll situatin. hazardous boulders and craters are interspersed and they bo detected from among the other features.

The vehicle performs well and is able to traverse the boulder/ crater field without mishap. On the first two scans hazards are detected. This comes as a surprise since the terrain was intended to be safe at that point. What is seen is a 0,15 meter deep crater adjacent to a 0.15 meter boulder. The features are safe when considered separately. When put together, though, the combination yields an average slope of $20^{\circ}$ over a 0,25 meter rise with a maximum instantaneous slope of $37^{\circ}$. This compares to predicted slopes of $34^{\circ}-36^{\circ}$. It is possible that the algorithm made the correct, though unexpected, decision. A sharp turn follows and the vehicle makes a successful attempt to enter the field at another point. It proceeds rn a direct $=0 u r s e$ toward the target passing alongside a 0.25 meter deep crater. Contrary to expectations, the crater is not considered hazardous by the vehicle. Again there is a problem as to how the feature should be interpreted. With a depth of 0.25 meters, a maximum local slope of $48^{\circ}$, and an average slope of $18.5^{\prime \prime}$ from edge to center, it is not clear


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Venicle srajectory $15 \times 20$ syeres. Boulder. Crater tield
whether op not the crater is hazardous by the given algorithm. This is especially true since a local slope cannot be detexmined by the scanner, The decision is made more difficult considering the vehicle"s distance from the crater and its pitch and roll situation. The fact that the vehicle has some roll casts some doubt as to the validity of the results, Interestingly, the crater is definitely not hazardous. cause of its small size compared to the vehicle dimensions. If the vehicle were to pass through the crater, it would experience a maximum pitch of $7^{\circ}$ and a maximum roll of $9.5^{\circ}$. Of course the present algorithm is not expected to take these considerations into account, A more sophisticated algorithm is required that estimates the vehicle's expected pitch and roll tnstead of or in addition to the terrain slope.

The question as to whether or not the $15 \times 20$ system considered the 0.25 meter crater hazardous has not been answered yet, The only way to get the answer is to repeat the simulation after correcting the program for the roll adjustment error, What this simulation does demonstrate, though, is that the question of defining a hazard is by no means trivial, More work must be done to extend the simple rules developed here.

In all of the simulations, the $15 \times 20$ system with the hazard detection algorithm is able to steer clear of all obstacles, Unfortunately, some safe terrain is also avoided. This is necessary stnce all ambigl . . cases must be treated with caution. It was seen that one of the best ways. of improving the performance is to increase data density and accuracy. by going to higher order scanning systems, The number of bad estimates is. reduced thereby minimizing the variation of computed slopes, Just as important is the accuracy of the data interpreter. The algorithm developed


#### Abstract

is quite reliable since it makes use of all of the data available in a scan and attempts to account for all possibilit:'rg, however, the decisions made are based on some oversimplified assumpti.as and techniques and could stand some improvement. It must be stressed that this is not disastrous since the assumptions lend to make the algorithm more conservative in decision making. While the algorithm does use all of the data taken in a scan, it ignores data taken from previous, overlapping scans. No doubt the additional information would be helpful but the tradeoff is a much more time consuming program,

Even with the difficulties mentioned above, there is absolute confidence in the slope estimates. Not once did the true terrain slope lie outside the bounds predicted by the maximum and minimum slope calculations. Sometimes this is hard to ascertain since the estimates are only averages taken over a finite range interval in which the terrain slope is varying, Furthermore, the maximum slope estimates are usually computed over a shorter range than the minimum slopes, As a result, it is sometimes difficult to attach physical meaning to the estimated range of slopes when the upper and lower bounds do not represent the same terrain, Figure 43. The estimates are still no less correct in spite of this,

Another otential problem demonstrated by the simulations is inadequate field of view. With a small fleld of view, there will be times when absolutely no information is received about a section of terrain. This is particularly dangerous when the area in question is very close to the vehicle. The only decision a cautious vehicle can make when no information is given is to choose another path, This is an unnecessary hindrance to the vehicle, The additional lasers and detectors needed to cover the entire field of view would be a worthwile modification, Assuming the




FIGURE 43
Example of calculated slofes demonstrating difficulty. in attaching physical significance to estimates
vehicle will never have a pitch in excess of $+30^{\circ}$, a $60^{\circ}$ field of view is recommended. At $2^{\circ}$ per detector, this requires an additional ten detectors. Given a fixed number of detectors, the fiell of view problem is a tradeoff between adequate peripheral vision and curate center vision. Expanding the field of view without adding additional detectors degrades the data overall. One solution might be to concentrate most of the laser shots and detectors directly in front of the vehicle. This would insure that one very accurate scan can be obtained. To deal with the peripheral vision problem, additional lasers and detectors with greater spacing could cover the fringe areas. These would be used mainly to signal a major shift in the terrain. If the terrain rises or falls greatly or if the vehicle is pitched, the scanner can be rotated so that it again focuses on the ground immediately in front of the vehicle. Naturally, this would require a degree of sophistication that has not yet been reached with the R.P.I. rover. However, the idea is by no means unrealistic.

There is one great Iimitation with the hazard detection algorithm that has not ever been considered in this paper. This data processor was designed to interpret data in a single azimuthal scan. Unfortunately, gradients occur in all directions and these are totally overlooked, -It ing entirely possible to have safe slopes in the inpath direction but hazardous slopes in the crosspath sense, Figure 44. This happens, for instance, when the vehicle taveis across the face of a slope. One way of dealing with the problem would be to try to estimate crosspath information from inpath slopes. As an example, suppose a $20^{\circ}$ slope is indicated one azimuth while an adjacent one is estimated at $-20^{\circ}$. It is apparent that while both paths are safe in the inpath sense, a rarp change occurs in the crosspath direction. This technique, though, is qualitative at best and gives much more
86.

## Vehicle Attitude Problem side view



Crosspath Hazard Problem
front view


FIGURE 44
Vehicle Attitude and Cross:
Considerations
emphasis to the inpath interpretation. There is no reason to believe that the inpath slope is any more important than the crosspath slope. To give equal treatment to the crosspath case, the algorithm used for the inpath data can also be applied to data perpendicular to the path. The main problem with this solution is that except for very close ranges, the data density in the crosspath sense is less than in che azimuths. Terrain interpretation would be difficult. This can be remedied by taking the azimuth shots at increments smaller than $10^{\circ}$. To maintain the same $150^{\circ}$ sweep, this requires more azimuths, more data, and more processing time. Another alternative is to make the azimuth shots closer together but to use the same number. This concentrates them in a much smaller area. The center vision is again improved at the expense of peripheral vision. It would appear that the tradeoff is worth it since the terrain immediately in front of the vehicle is of the greatest concern. If a sharp turn is required, a scan in the new direction can be made before any action is taken.

The hazard detection algorithm presented $i$. does have some weak areas that require further study. Most importait among these are the need for crosspath slope information and a more refined conception of what an obstacle is. The main purpose of this investigation, however, was to determine whether or not a multi-laser/detector triangulation hazard sensing system is feasible. This has been accomplished. Despite resolution problems, reasonably accurate terrain characterizations have been made using the R.P.I. system. More importantly, reliable bounds have been placed on the terrain estimates. It is for this reason that the R.P.I. rover never found itself in a dangerous situation and this is an important consideration when pianning a Mars mission. Without the
hardware constraints limiting the R.P.I. system, facreased resolution and performance are possible with a triangulation sensor. Increased laser density increases the quantity of data and finer individual detector fields improve the accuracy. The amount of information available from the data, however, is a function of the spacing of buth the lasers and detectors. Furthermore, it is, in theory, possible to extract the maximum amount of irformation available from a given sensor and to easily construct an accurate enveiope enclosing the terzain. Even with the limited data and large uncertainties characteristic of the R.P.I. system a short range multi-laser/detector triangulation sensor, by virtue of its speed, can compete with larger range sensors. This work s.rves as a basis for further investigation that should further strengthen thr case for the short range sensor.
ViII. CONCLUSIONS

Computer simulations nave shown that a 15 laser, 20 detector : inggulation sensor can be used for hazard detection on an autonomous riving vehicle. An algorithm has been developed that, regardless of the type of terrain, collates terrain slope estimates allowing the vehicle to locate and avoid obstacles. Because of ambiguity inherent in the system maximum and minimum slope estimates are computed to account for all terrain possibilities. The price to be paid is that if the range of slopes is large, a conservatively biased vehicle declares some safe paths as hazardous.


#### Abstract

It is possible to improve perception by increasing laser density. While this does not improve accuracy, the extra laser pulses increase the amount of data. Making detector fields smaller increases data accuracy without adding to the total number of data points. Computer simulations have shown that the combined effect of decreased laser spacing and reduced detector fields is to reduce the ambiguity in the system and increase the number of safe paths available.

It is evident from simulations that a $40^{\circ}$ overall field of view is insufficient to detent all safe terrains. Theoretical considerations suggest a $60^{\circ}$ field would substantially improve the terrain interpretations.


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## APPENDIX A -- CODE LISTINGS

A. Subroutine DIAGNL - This subroutine diagonalizes the laser/sensor returns.
B. Subroutine MODEL 2 - This subroutine processes the laser/sensor returns in accordance with the terrain modeling rules

```
SUBROUTINE DIAGNL
COMMON/SENSR/BEMRNG,ASMUTH,LANGLE,RTN,KOUNT,ISCAN
COMMON/SENX/HITLAS.HITSEN.NUMLAS,NUMSEN.NUMAZ.INTDAT.
E NMDTPR,LASAGL,SENGLE,SCON,DATA,DIAG,POS
INTEGER DIAG (50,50), BTN (50)
INTEGER*2 DATA(50.j0)
REAL ASMUTH (50). POS (50,51,2). LASAGL (50).SENGLE(50)
HEAL BEMRNG(50), LANGLE(50)
DO 100 J=1,NUMAZ
DO }100\textrm{I}=1\mathrm{ , NUMLaS
    DIAG (J,I)=DATA (J,I) - I
    IF(DATA (J,I) . EQ.0) DIAG (J,I) =1000
    URITE (6,200)
    FORMAT('1',10X,'DIAGCNALIZEE RETORN')
    WRITE (6,300) (J,N=1, NUMA Z)
    300. FORMAT('0'.'AZIMUTH'.3X.50(1X,I2,1X))
    WRITE (6.400)
500
600
400 FOAMAT('0','LASER')
    DO }500\textrm{I}=1\mathrm{ ,NUMLAS
    K=NOMLAS-I+1
    WRITE (6,600)K, (DIAG (J,K),J=1,NUMAZ)
    FORMAT(: 1.3X,I2.5%.50(I3.1X))
    RETUSN
    END
```

```
    SUEROUTINE MODEL2
    COMMON/CHOOSE/NMMOD,NMSEN,NMPSA,NGTRN,INTVDB,INTSEN.
    & INTHOD,INTPSA,INTTRN,INTGYR
        COMMON/TIEUP/THETNU,ALPHA,SLPIN,SLPCRS,SLPCHS,TALLOW
        COMMON/DYNMIC/EWNMAX,UPMAX,VEHLEN,VEHKID,CRSMAX,VELMAX.
    E DT,TUZN1,STBMAX
    COMMON/SENSR/BEMRNG,ASMUTH,LANGIE,RTN,KOUNT,ISCAN
    COMMOY/SLOPE/SLPMAX,SLPMIN,ISTOP1, ISTCE2
    COMMON/DETECT/SENTIM,SCNTIM.:IMSTP.SENSTP,SENHI,SENLIM.
    & SENLEN,SENHID,NUMBER,IJK
    COMMON/SENX/HITLAS,HITSEN,NUMLAS,NUMSEN,NUMAZ,INTDAT,NMDTPR.
    \varepsilon LASAGL,SENGLE,SCON,DATA, LIAG,DOS
    REAL GANGE(50),SLPMAX(50,50,3), SLPMIN(50,50,3), AZMOTH(50).
    E ASMUTH(50), POS (50,51,2),LASAGL(50),SENGLE(50)
    REAL BEMRNG(50). LANGIE(50)
    INTEGER HAZARD(50),DELTA,JOMP(50,2),SEN1,SEN2,SEN3,SEN4
    INTEGER RTN (50), DIAG (50,50)
    INTEGER*2 DATA (50,50)
    IP(IJK.GT.0) GO TO 7C
    RSAD(5,30) MXMISS,ZMAX,RNGMIN
    FORMST(I2,8X,2F10.5)
    CONVE }F=180./3.1415
    DO 5O I=1,NOMAZ
    AZMUTH(I)=ASMUTH(I) *CONVRT
    RETURN
    CALL DIAGNL
    DO ONE LOOP PER AZIMUTH
    DO 4000 J=1. NUMAZ
    I=1
    CHECK FIRST RETURNSPOR MISSING LATA
C
    100
    200
    300 ISTRT=I
C LCOK FOR MISSING RETURNS
    400 IF(I.EQ.NUMLAS) GO TC 800
    I= I+1
    IF(DIAG(J.I).NE.1000) GO TO 400
C IMISS1 - FIZST LASER WITH MISSING RETURN
    IMISS 1=I
C MAX. NO. OF MISSES EXCEECED?
    500 IP(I-IMISSI.GE.MXMISS) GO TO }70
    I=I+1
C LOOK FOR NEXT HIT
    IP(I.GI.NUMLAS) GO TO 900
    IF(DIAG(J,I).EQ.1000) GO TO 500
    IMISS2 - LAST LASEÁ WITH MISSED RETURN
    IMISS2=I-1
C COMPARE DIAGONALIZED RETUANS BEFORE ANL AFTER MISSES:
C INSERI LOWER DIAGOIALIZEL RETURN AND CORRESPONDING SENSOR
C THAT WOULD HAVE SELN LASER AT AIL MISSES
    DELTA=MINO(DIAG(J,I),DIAG(J.IMISS1-1))
    DO 600 K=IMISSI,IMISS2
```

```
        DIAG (J,K)=DELTA
        DATA(J,K)=K+DELTA
C IF tOO MANY MISSES. STORE GANGE OF LAST HIT.
    700
    HAZARD INDICATES TYPE OF OBSTACIE ANE IS USED LATER.
    IHIT=IMISS1-1
        ISAW=DATA(J.IHIT)
        RANGE(J) = POS(IHIT.ISAN. 2)
        HAZARD(J)=2
        GO TO 4000
        ISTOP= NUMLAS
        GO TO 1000
        ORIGINAL PAGE IS
        OF POOR NTIIIIY
C REMEMBER LASER THAT EADE LAST HIT
        900
        1000
C
C
    1 1 0 0
    1200
    1300 DO 1400 K=ISTRT.ISTOP
        IF(DIAG(J,K).LE.DIAG (J.ISTRT)) GO TO 1400
        ISAN=DATA(J,K)
        RANGE (J) = POS (K,ISAW, 2)
        HAZARD(J)=4
        GO TO 4000
    1400
C
    1500
C
C IN JOMP
    JUMP(N,1) - LASER GEFORE N-TH JUMP
    JOMP(N,2) - SILE OE N-T! JOMP
    ISTOP1=ISTOP-1
    IP(ISTOP1.EQ.O)GOTO 1010
    DO 1600 I=ISTRT,ISTOP1
    DELTA=DIAG(J.I+1)-DIAG(J,I)
    IE(DELTA.EQ.O) GO LO 1600
    I2OUNT=ICOUNT+1
    J0:P(ICOUNT, 1)=I
    JUMP(ICOUNT, 2) = DELTA
    1600
C
    1610
C
    1 6 5 0
    1700
C N INDICATES SIGN UR JUMP, IF JUMP<O ,N=-1
    y=1
    LAS 1=JUMP(K,1)
```

```
        SSN1=1FIX(DATA(J.LAS1)*.S-N/2.)
        21=POS(LAS1.SEN1.1)
        IF(N*JUMP(K,2).GT.1) GO TO 1900
    1800
C
    1900
    2000
C
    2100 IF(LAS2-LAS1.LE.2) GC TO 2150
        LAS3=LAS 1+1
        LAS4=LAS2-1
        SEV 3= IFIX(DATA (J.LAS 3) +.5-N/2.)
        SEN4=IFIX(DATA(J.LAS4)+.5+N/2.)
        Z3=POS (LAS3.SEN3.1)
        Z4=POS (L,S4,SEN4.1)
        R3=POS(LAS3,SEN3,2)
        R4=POS(LAS4,SEN4, 2)
        DELZ=24-23
        GO TO 2170
        [3=POS(LAS1,SEN1,2)
        R4=POS (LAS2,SEN2,2)
        2170 DELR=R4-R3
        JCOUNR=JCOUNT+1
        SLPMAX(J,JCOUNT,1)=ATAN2(DELZ,DELR)
        SLPMAX(J,JCOUNT, 2) =R 3
        SLPMAX(J.JCOUNT,3) =R4
        R=K+1
        GO TO 1700
        STORE NUMPER OE SLOPES CALCULATED
        ISTOP1=JCOUNT
        IF(ISTOP1.NE.0) GO TO 2250
        HASARD(J) =9
        GO TO 4000
        2250 K=1
        JCOUNr=0
2300 IF(K.GT.ICOUNT) GO TO 28CO
        I NCR=0
        N=1
        IP(JUNP(K,2).LT.0) N=-1
        LAS1=JJMP(K,1)
        SEN1=IPIX(DAPA(J.LAS 1) +.5+N/2.)
        21=POS(LAS1,SEN1,1)
        IP(N*JUMP(K,2).GT.1) GO TO 2500
        2400 INCR=INCR+1
        IF(K+INCR.GT.I:`UNT) GO TO 2800
        IF(N*JUMP(K+INCR,2).LT.0)GO TO 2600
2500 LAS2=JUMP(K+INCR,1)+1
    SEN2=IFIX(DATA(J,LAS2)+.5-N/2.)
    Z2=POS(LAS2.SEN2,1)
    DELZ=22-21
    IF(ABS(DELT).GT. 2MAX)SO TO 2700
    GO TO 2400
```

C SEARCII FOR OBSTACLES
C FIRST CHECK SLPMAX FOR CLOSEST SLOPEDCRITICAL VALUE
DO 2900 I=1. ISTOP1
IF(SLPMAX(J,I,1).LI.UPMAX.AND.SIMMAX(J,I, 1).GT.DWNMAX)GO TO 2900
IF(SLPMAX(J,I,2) .GI. RNGMIN) GO TO }300
RANGE (J) =SLPMAX(J,I, 2)
HAZARD(J)=5
GO TO3200
CONTINUE
HAZARD(J)=8
GO TO }320
IF(ISTOP2.EQ.0) GO To 3100
DO 3100 I=1.ISTOP2
IF(SLPMIN(J,I, 1).LT. OPMAX.AND.SIPMIN(J,I, 1).GI.DWNMAX)GO TO 3100
RANGE(J) =SLPMIN(J,I, 2)
HAZABD(J)=6
GO TO 3200
CONTINUE
HAZARD(J) =7
IF(INTMOD.NE.1)GO [O 4000
WRITE (0,3300) azMUTi:(1)
3300 FORMAT('-'.'AZIMUT! ANSLE ',F4.'.' EEGEEES')
IP(ASTOP1.LT.1) GO 2O 3020
DO 3320 I=1. ISTOP1
3320
3400 FORMAT('0'.'MAX SLOPE(DEG)'.SO(2X.FU.O!)
3 5 0 0
3600
3620
3640 SLPMIN(J,I,1)=SLPMIN(N,I, I)*CONVET
3700
3800 FO:MAT(' '.MIN RANGE(M) ', 2X,50(2X,F5.1))
WRITE(G,3900) (SLPMIN(J,I,31,I=1,ISTOP2)
3900
4000
C
4100 FORMA[('1'.'AZIMURH'.46X.'TERRAIN CHARICTERIEARION'.4JX.'RTN')
SLPMAX(J,I,1)=SLPMA.N(J,I,1)*CONVRT
HRITE(6,3400) (SLRM:X(J, [, 1),I=1,ISTOP1)
WRITE(6,3500) (SLRMAX(J,I,2),I=1.ISTCP!)
FOHMA:(" '.'MIN RANGE(N) ', 2X.50(2X,F゙S.ll)
WRITE (0.3600) (SLPMIX(J,I,3).I=1,ISTOE1)
FOHMAT(' '.MAX RANGE(M) '.2X.50(2X,F5.i))
IF(ISTUIN.[t.1) GO TO 4000
DO 3640 I= 1,ISIOP2
WRITE(6,3700) (SLPMLN(J,I,1),I=1,ISTUR2)
FO|MAT('O0, MIN SLOEE(DEG) !.50(2X,F4.U))
WRITE(0,3800) (SLP.._N(J,I,2),I=1,ISTOP2)
FOKMAT(' ',MAX aAMgE(Y) ', X,50(2X,F5.1))
CONTINU::
OUTPUE DATA ON AJInOTHS
WRITE(0.4100)
DO 2OUCO J=1,NUMAZ
I=|AZARD(J)

```

GO TO (5000, 6000, 7000,80C0.9000, 10000,11000,12000).I
RTN \((J)=1\)
WRITE 6.5500\() \mathrm{J}\)
 .'IS PASSABLE.'.61X.'1')
GO TO 20000
\(\operatorname{RTN}(J)=0\)
HRITE (6.6500) J, MxuISS. BANGE (J)
FORMAT('00.3X,I2.5X,I1, MISSING RETURNS DETECTED BEGINNING". 'at a range op ',p3.1.' meters. terrain IS not passable.'. 23x, \(0^{\prime \prime}\)
GO TO 20000
\(\operatorname{RTN}(J)=0\)
WRITE \((6,7500) \mathrm{J}\), RANGE (J)
 - meter range wita vehicle pitch below a sape level. not'. - pasSable.'. \(10 \times 0^{\prime \prime} 0^{\prime \prime}\)

GO TO 20000
\(\operatorname{RTN}(J)=0\)
WRITE \((6,8500)\) J. RANGE (J)
 'meter range hitg vehicle pitce above a safe level. not', - PASSABLE.'. 10x, \({ }^{\prime \prime}{ }^{\prime \prime}\)

GO TO 20000
\(\operatorname{RTN}(\mathrm{J})=0\)
WRITE \((6,9500) \mathrm{J}\), Range ( J\()\)
 - WITH dossibly hazabdous slope. terain will ee avoided.'. 16x. \(10{ }^{1} 1\)
GO TO 20000
RTN ( J ) \(=0\)
RRITE \((6,10500) \mathrm{J}, \operatorname{RANGE(J)}\)
format('0., 3X, I2,5z, obstacle detected at •f3.1.' meter range ', 'hith definitely hazabdous slope. terrain is not passable.'. \(14 \times .0^{\prime \prime}\)
GO TO 20000
\(\operatorname{RTN}(J)=1\)
GRITE (6, 11500) J
format ('0'. \(3 \times, 12.5 \times\). POSSIBLE OBSTACLE DETECTED BOT NOT CLOSE '. 'enough to necessitate avoidance.'.38x.'1')
GO TO 20000
\(\operatorname{RTN}(J)=1\)
WRITE \((6,12500) \mathrm{J}\)
 ,' TERAAIN IS PASSARLE.'.52X.'1')
continde
betorn
END
sobroutine pitch(J)
COMMOY/SENSB/BEMRNG, ASMOTH, LANGLE, ETN, KOUNT, ISCAN
COMMON/TIEUP/THETNU,ALPHA,SLPIN,SLPCRS,SLPCHS, TALLOW
COMMON/SLORE/SLPMAX, SLPMIN, ISTOP1, ISTOP2
REAL SLPMAX(50,50,3), SLPMIN (50.50, 3), ASMUTR (50)
real bemang (50), langle (50)
INTEGEK RTN (50)
teis subroutine corrects calculated obstacle slopes to account
foa vehicle attitude.
Delta \(=\) asmuta ( J ) +alpha
thetazatan (Tan(SLPIN) *COS (DELTA) -TAN(SLPCRS) *SIN(DELTA))
```

        DO 100 I= 1,ISTOP1
        SLPMAX(J,I,1)=SLPMAX(U,I, I) +THETA
    100
        CONTINUE
        DO 200 I= 1,ISTOP2
        SLPMIN(J,I,1)=SLPMIN(J,I,1) +THETA
        CONTINUE
        RETURN
        END
    ```

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