TAKE 3

IGVC 2019 – Design Report University of Detroit Mercy





Team Captain: Samar Bayan Team Members: Nathaniel Maley Yuyi Li Mohamad Ali Mokhadder Ali Baholhavaeji Karthika Balan Carlos Carpenter Melvin P Manuel Ratheesh Ravindran Christopher Harness

Faculty Advisor Statement:

We certify that the engineering design in this vehicle undertaken by the student team, consisting of undergraduate students, is significant and qualifies for course credits in senior design and in the undergraduate program respectively.

awarded credit in a cenior	g of the vehicle (original or changes) by the current student team has been significant and equivalent to what might be design course.
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Introduction

The University of Detroit Mercy Senior Design 2019 team is entering the competition with TAKE 3, a successor to Detroit Mercy's 2018 design group's robot VERTIGO. TAKE 3 sets itself apart from its successor through improved software suite and redesigned hardware components. Modifications have been made to the prior mapping, IP, and goal selection algorithms. In addition, a new redistribution of hardware was made to support easy conversion between Tractor and Balanced modes of TAKE 3's base chassis, Segway RMP 220-v3. The result is more precise localization, maps, and navigation goals. This report highlights the significance of our hardware and software systems that go into the creation of TAKE 3.

Team Organization

The 2019 IGVC team is composed of five undergraudate Electrical Engineering and Robotics and Mechatronics Students, and 6 graduate Electrical Engineering Students. In addition, four of the seven Electrical Engineering students have a concentration of computer engineering. These diverse areas of study allow for a very versatile and dynamic group with all the skills necessary to make TAKE 3 a success.

The work distribution focused on channeling students towards their strengths. Having 11 members in the group gave the team flexibility with the number of people per module. A list of tasks was developed and broken down into modules, then team members were assigned to the tasks and Gantt charts were developed to organize team activities and track progress. The team met twice a week and reported progress though oral reports as well as documented logs on a web-based shared repository. The team facilitator managed the organization of the repository and insured proper submissions in timely manner. A team leader developed agendas for every meeting and reported with the facilitator to the team advisor. The team devoted approximately 12 hours a week for 20 weeks totaling 240 hours during the academic year in addition to 100 hours projected towards the end of the competition. Table 1 lists TAKE 3's team members and their corresponding tasks.

Team Members and Responsibilities					
Name	Responsibilities				
Ratheesh Ravindran	Image Processing				
Mohamad Ali Mokhadder	Navigation & Path Planning				
Ali Baholhavaeji	Localization				
Karthika Balan	Navigation & Path Planning				
Samar Bayan	Image Processing				
Yuyi Li	Image Processing				
Nathaniel Maley	Hardware				
Christopher Harness	Image Processing				
Yusuf Dilawar	Image Processing				

Carlos Carpenter	Inter-operability Profiles (IOP)		
Melvin P Manuel	Inter-operability Profiles (IOP)		

Table 1: Team Members and Responsibilities

Cost Analysis

The cost of TAKE 3 is broken down in Table 2 with an approximate total of \$44 K.

Vertigo Team Costs				
Component		l Unit Cost	Tea	am Cost
Segway RMP 220 v3 Chassis	\$	24,000.00	\$2	24,000.00
Caster Wheel for Tractor mode Capability	\$	247.00	\$	247.00
Auxillary Batteries	\$	729.00	\$	1,458.00
Battery Charger	\$	55.00	\$	55.00
Multisence S7 3D Camera	\$	6,800.00	\$	6,800.00
Velodyne VLP-16 LiDAR Puck	\$	8,000.00	\$	8,000.00
KVH CG-5100 IMU	\$	15,000.00	\$	-
ProPak6 Triple-Frequency GNSS Receiver	\$	22,070.00	\$	-
Sparton AHRS-8P IMU	\$	1,425.00	\$	-
NUC Computer	\$	1,100.00	\$	2,200.00
Mini Box intel Computers (3)	\$	1,000.00	\$	-
Router	\$	100.00	\$	-
Aluminum framing	\$	400.00	\$	400.00
aluminum sheeting	\$	384.00	\$	384.00
Shelving Unit	\$	400.00	\$	-
E-Stop Controller	\$	353.00	\$	353.00
Totals	\$	82,063.00	\$4	43,897.00

Table 2: TAKE 3 Cost Estimates (blank cells indicate donated items)

Power Budget & Distribution

TAKE 3 has two independent power systems: the first is built into the Segway and operates the RMP 220 base platform (motors) and its corresponding built-in computer, and the second is added to the chassis to power all the custom added sensors and computers.

The built-in Segway power system consists of three 72V, 380Wh batteries, two of which power the two Segway motors, and the third powers the onboard computer and motor controllers. The Segway reports a max range of 30 miles on a full charge, and a charging time of about 3 hours.

These specs were sufficient to drive the vehicle for a full day under IGVC conditions and rendered no serious challenges to the team.

The second added power system was designed around a quickly swappable 52V, 24A Panasonic GA 18560 battery. This battery provides 1300Wh when fully charged, which can operate TAKE 3's 169W load for about 7.5 hours. With two batteries in-house and a charging time of 4.5 hours, TAKE 3 can run for a full day with no power concerns, and a battery can be swapped and fully recharged before the replacement is exhausted.

Panasonic recommends the batteries to be charged to 80% of full capacity to extend their lifetime. This will result in a charging time of three hours and an operation time of six hours per TAKE 3's load. Again, this configuration is sufficient for IGVC conditions and provides plenty of recovery time in case there are gaps in charging.

TAKE 3's power budget was derived by summing the power consumption of all its accessory sensors and computing resources; it is presented in the Table 3.

POWER BUDGET							
		Normal Operating Conditions			Worst Cas	e Operating C	Conditions
Device	Quantity	Voltage (V)	Current (A)	Power (W)	Voltage (V)	Current (A)	Power (W)
Velodyne LiDAR	1	12	1	12	18	1.75	31.5
Carnegie Multisense							
S7	1	24	0.3	7.2	24	0.8	19.2
DVDO G3-Pro Air 3C							
Pro	1	3	1	3	5	1	5
Netgear NightHawk							
X6 Wireless Router	1	12	0.55	6.6	12	1	12
Microstrain 3DM-							
GX2 IMU	1	9	0.09	0.81	1	0.09	0.09
Spartan AHRS-8							
Digital Compass	1	5	0.064	0.32	5	0.064	0.32
Mini-Box Computer	2	12	2.5	30	12	3.75	45
Novatel Propak LB							
Plus GPS	1	12	0.31	3.72	12	0.4	4.8
Indicator LEDs	160	10	0.02	0.2	10	0.06	0.6
Wireless E-Stop	1	12	0.045	0.54	12	0.465	5.58
Total			5.879	64.39		9.379	124.09

Table 3: TAKE 3 Power Budget - Normal and Worst Case Operating Conditions

Software Strategy

The team is split up into four groups. The groups are as follows: JAUS, perception, localization, and navigation. The JAUS group worked separately from the group for most of the project. The JAUS module functions on its own without much reliance on the other functionality of TAKE 3. The perception group holds the responsibility of detecting obstacles and lane lines. As time went on, the perception group started working more closely with the navigation group to ensure that the robot does not cross lane lines and avoids obstacles. The localization group holds the responsibility of integrating the GPS, IMUs, and wheel odometry. Each data set is transformed into one coordinate system for uniformity, with Kalman filtering used for sensor fusion. The navigation group holds responsibility for the navigation stack. This consists of mapping, local and global path

planning, as well as goal selection, which is an algorithm used to determine where the end point should be for the global path. Having been given some code from last year's team, the team's primary focus was on integration of perception, localization, and navigation. To achieve this goal, the three aforementioned groups were required to work closely together.

Robot Design

TAKE 3 is a Segway RMP 220 Chassis capable of a 200lb payload before modifications. The chassis is equipped with reinforced aluminum framing to support our various sensory equipment. The system is run with two Intel Mini-Box Computers using the Ubuntu based ROS platform to run the robot. TAKE 3 uses a VLP-16 Velodyne LiDAR, and Multisense 3D Camera for lane line detection and obstacle avoidance. Two GNSS-502 Antennas are mounted on top of the vehicle and feed into the ProPak6 GNSS Receiver. The receiver works with the KVH IMU to give precise heading and global positioning.

Mechanical Improvements

The 2019 team changed some key components to the mechanical design from the 2018 team to make the robot more functional and efficient. The aluminum frame gives TAKE 3 a static camera configuration, simplifying data reception and analysis by comparison with the dynamic configuration of the gimbal which we used in 2018 design.

The team designed and welded a new shelving unit to house the IMU, batteries, and Intel Mini Box Computers. The new design makes the components easily accessible while providing more protection than the

previous design. The shelving also creates a static location for the IMU, giving TAKE 3 more accuracy in heading and positioning.

Navigation

TAKE 3's software is built on Robot Operating System (ROS), which provides a variety of advantages. Most importantly, ROS is a peer-to-peer networking framework that allows efficient communication between software modules; these modules can be located on different computer platforms, which allows us to distribute computation tasks and increase system speed. ROS also comes equipped with software modules that can be configured to work with different systems; specifically, ROS's navigation stack was used, and some sensors come with ROS-compatible modules. Overall, systems provided with the navigation stack did not work out of the box and needed to be configured in a manner specific to our system.

ROS allows efficient debugging and adjustment of modules, as the topics that modules use to communicate can be monitored, manually fed data, and easily adjusted. Systems such as IOP, image processing, goal selection, and navigation (Movebase) were all developed and debugged modularly, which simplified programming efforts.

Figure 1-TAKE 3 Final Design

ROS provides other advantages suited to our application. We used the ROS Gazebo simulator to test navigation systems before deploying them to the actual robot. We also use the ROS transform library, which allows rapid and efficient transformation of data. This refines the mapping process by combining all sensor data into a common frame of reference, eliminating the possibility of data skew.



Figure 2: Block Diagram of System Architecture

Odometry & Kalman Filter

TAKE 3's odometry consists of a built-in ROS Kalman filter, which is fed data from a variety of position sensors. Since the GPS and IMU already have a built-in Kalman filter, but we think this localization method in not enough to getting the accurate state of robot based on that. The Kalman filter is a mathematical method to estimating the state of robot based on series of measurements which are observed by sensors, also handles the fusion of these disparate data types (e.g. IMU, GPS, etc). We use the standard robot localization package to get more accurate state estimation which it has two stages:

At first, we estimate the local state of the robot based on two Sparton IMU sensors and odometry and feed these data to first Kalman filter to get the local estate estimating. Then the output of the first Kalman filter goes to the second Kalman filter with GPS data and one IMU data to do global estate estimation.

We also analyzed the use of other tools provided in the ROS navigation stack, including Adjustable Monte Carlo Localization (AMCL) and SLAM Gmapping. AMCL accepts a static map and a LiDAR scan and will attempt to approximate the robot's position within that map. Gmapping is similar but produces a ROS transform between the given static map and the robot's odometry frame within that map. The benefit of using such localization systems is that they reduce the uncertainty in the Kalman filter's final approximation by providing world-referenced position data.

We simulated both modules, but ultimately decided to use the Kalman filter to fuse our globallyreferenced and dead reckoning sensor data.



Figure 3- Localization scheme

Local & Global Planner

The global planner aims to implement A* algorithm with some modifications according to the desired application. Since there is no map available for the environment, and the robot has to navigate between plenty of obstacles placed on its left and right sides, the written algorithm aims to generate relatively close way-points. Moreover, the robot is navigating within a zig-zag path, so it is mainly unable to detect obstacles that are few meters away. Therefore, the algorithm will check for few closed nodes around the robot's current location then chooses the closest one to the goal but within 1 to 3 meters range from the robot. Note that closed nodes are the free cells that the robot may navigate through to reach its goal. In other words, the generated way-points are more considered as bread crumbs. The traditional A* algorithm plays the role of a global navigator where it generates a complete path from the original position till the destination based on a given map. The generated path is usually represented by a vector of waypoints (named closed nodes in A* algorithm) where the first element of the vector is the starting point, the last one is the desired destination and all points in between forms the optimal path generated. As long as the robot has no idea about the entire environment, it is inefficient to use a global planner because the generated path will be continuously updated. Thus, it will be time consuming to plan the whole entire path after each step. For this reason, the applied algorithm was not the traditional A* yet it is a modified version. In other words, the code generates only one breadcrumb at a time, then the robot will move toward that intermediate goal before another breadcrumb is produced. Modified A* stops iterating once the number of detected closed nodes is 10, then the closest one will be chosen as a breadcrumb. Using local navigator (to be discussed in the next paragraph) the robot approaches the breadcrumb, then modified A* will run again to find the next intermediate goal. The flow chart shown below summarizes the steps explained above.



Figure 4- Modified A* Global Planner

The aim of the local navigator is to adjust the linear and angular velocities of the robot so that it can approach its intermediate goal (generated bread-crumb) smoothly with minimum latency. It is simply based on measuring the distance to the obstacles detected by the lidar and getting the direction of the empty path between those obstacles. Since the robot is moving with forward and angular velocity, its axis is rotating thus the inclination angle between the robot and its goal is variable with time. Inclination angle or error must decrease as the robot approaches its target. The angular velocity must be proportional to error so that the robot will not oscillate too much before reaching its goal, however; the linear velocity must be inversely proportional to the distance separating the robot from its goal. The flowchart below describes the algorithm for smooth drive towards breadcrumbs.



Figure 5- Local Navigator

Goal Selection

The local and global planner generates the path to the goal given by the goal selection algorithm. The purpose of the goal selection algorithm is to keep the robot from losing its heading and turning around unexpectedly. The robot uses a system given by image processing to map right and left lane lines separately in a costmap. With this information, TAKE 3 knows how to keep its heading in the forward direction by comparing the history of the right/left lane line map to the instantaneous image processing data. The goal selection algorithm toggles between image processing goals (local goals that are generated in the presence of lane lines) and GPS goals (provided in the Auto-Nav and IOP challenges).

LiDAR

TAKE 3 uses a VLP-16 Velodyne LiDAR Puck equipped with a 360-degree horizontal field of view (FOV). The LiDAR on TAKE 3 is configured for a 270-degree FOV horizontally with a 30-degree vertical FOV. Raw data from the LiDAR comes as a 3D point-cloud. The 3D point-cloud is reduced to a 2D laser scan primarily to reduce computation. The data runs through two filters

during this reduction; a voxel grid filter and a radius outlier filter. The radius outlier filter only passes points with a given number of neighbors in a threshold distance. The voxel grid filter applies a 3D grid of cubes measuring 343 cubic centimeters (7cm x 7cm x 7cm) over the input data. The data is then downsized by passing the centroid of each voxel cube to the laser scan.

GPS & IMU

TAKE 3 uses a Novatel Propak6 GPS receiver coupled with two VEXXIS™ GNSS-502 Dual Band Antennas, making TAKE 3 capable of SPAN technology. This technology provides continuous 3D positioning, velocity, and altitude. When the GPS receiver and antennas are running, TAKE 3 can localize its position to within 4-30cm. The longitude and latitude are transformed into UTM – Universal Transverse Mercator – to generate a transform for the data, enabling exact positioning.

TAKE 3 also uses a KVH CG-5100 IMU for extra certainty in navigation. The IMU collects data on the positioning and heading of the robot, which integrates the data with that of the GPS for use in localization, goal selection, and navigation.

Mapping

TAKE 3 tracks its current and past environments using ROS implemented costmaps. These maps are marked using our LiDAR and camera sensory equipment. The sensors are also able to clear obstacles off the map, dependent on the situation. Obstacles identified by the sensors are marked with confidence values, such that TAKE 3 can choose to recognize an obstacle as being absent or present.

TAKE 3's LiDAR and camera each have their own individual costmaps. This is done to prevent one sensor from clearing objects that the other sensor sees. We then integrate the two maps together by converting the data types of the individual maps into one recognized universally by the ROS control system.

Image Processing

Image processing is used to detect lane lines. TAKE 3 uses a Multisense S7 3D Camera by Carnegie Robotics. The camera provides the distances of each pixel in a 3D point cloud format. The information below explains how the locations of the lane lines are extracted from the raw camera data.

Ground Plane Extraction

TAKE 3 creates a 2-D image of the ground plane from the 3-D point cloud data. Since the camera is mounted at an angle of 35-degrees below horizontal, the ground plane can be extracted by accepting points at a certain

distance along the camera's z-axis. Every point in the ground plane forms a triangle with the camera in which the hypotenuse is the camera's Z-axis, the base is the distance from the robot base, and height is the camera height. If the Z-value for a point in front of the robot is less than the expected





ground hypotenuse value, then this point is rejected from the ground plane image. We can adjust the threshold for the Z-value cutoff in case there are variations in the height of the ground plane.



Figure 7: Standard Image



Figure 8: Ground Plane Extracted Image

Metric Image

In the extracted image, the distance per pixel is not uniform due to the optical characteristics of the camera which causes difficulties during the sensor fusion operation with the LiDAR. Hence, a uniform number of meters per pixel in the ground plane image is important to accurately mark the locations of the lane lines and fuse then into the navigation maps.



Figure 9: Metric Image Re-shaping

Vision Code

TAKE 3 uses vision code to extract a binary (black and white) image from the metric image. This is the final output used for navigation. The vision code uses the openCV C++ library, which has a list of standard image functions and classes. The RGB planes are extracted from the metric image. Because grass typically has a higher red content than blue content, the red plane is subtracted from twice the blue plane. In doing so, the image is kept on a 0-255 grayscale interval, while creating a stark contrast between the grass and the white



Figure 10: Binary Black/White Image

lane lines. Intensity thresholding is applied to create the binary image. Canny edge and median blur functions are used to reduce computation for the Hough transform and increase accuracy through elimination of salt and pepper noise. The Hough transform is used last to pass only points that fall into a certain set of lines. These lines are the lane lines, thus giving the final output image.

Mapping Integration

TAKE 3 converts the binary image into a laser scan allowing us to mark and clear cost maps. Because the binary image came from the metric image – a geometrically uniform image in meters per pixel – we can derive the location of the lane line pixels in TAKE 3's X-Y coordinate system. The X and Y distances are converted to polar distances to match the laser scan data type. TAKE 3's system adds this layer to the overall cost map and gives authority to mark and clear only on this layer.

To avoid erasing lane lines, the scan is given a variable size based on the locations of lane lines in the binary image. The scan only has authority to mark and clear lines from a region determined by the minimum and maximum angles of the white pixels with respect to the robot, i.e. the areas that are known to be see clearly.



Figure 11: IP Costmap

E-Stop and Controller

The controller has both a joystick for manually controlling TAKE 3, and an E-stop button for stopping the robot in case of emergencies. What makes the controller unique is its versatility, the controller is custom made with multiple mode selections. There are two distinct ways to stop the vehicle when the E-stop button is pressed that are differentiated by a switch: Decelerate-to-zero (DTZ) and Emergency stop. DTZ mode stops the robot without turning it off, while E-stop mode shuts TAKE 3 down, making it ideal for extreme emergency situations. There is another mode select for switching TAKE 3 between autonomous mode and remote-control (RC) mode. Lastly, there are push buttons for the Segway's balance mode; pushing the correct button allows for TAKE 3 to operate with or without its caster wheel attached.

The controller's programming also manages the indicator LEDs, setting them to different colors to indicate connectivity between TAKE 3 and the controller, autonomous mode/remote-control mode, and when the robot has been E-stopped.

The controller is designed such that it does not need to be disassembled to access the internal Arduino for reprogramming; this allows limitless modes or functionalities to be edited, added, or removed.

The LED functionality can be reprogrammed to indicate current status, which carries the potential for real-time feedback from the robot. In this way, the team can potentially detect errors while testing or during the competition based on the color and status of the LEDs.

Hardware

TAKE 3 uses a custom RC controller based on two Arduino Unos for data processing and control. The Arduino has an operating voltage of 5V with a 16 MHz clock. To cut down on the amount of processing and potential delay resulting from the Arduino's multiple functions, an Arduino Pro-Mini powered by ATmega328 was used to control the LED programs for TAKE 3. This allowed for less than one millisecond delay in processing for both E-stop and joy-stick commands, and less than one millisecond delay for the corresponding LED indicators. Two Xbee Pro SB3 wireless transmitter/receivers create the serial point-to-point network connection to send data between TAKE 3 and the wireless controller.

IOP Challenge

One of the communication systems implemented in Take3 is the Joint Architecture for Unmanned Systems (JAUS) protocol. TAKE 3 uses the 2010 version of the Aerospace standards. These standards specify the unique structure, transportation, and responses for each message and service. The three types of communication that JAUS enables are discovery, navigation commands, and reporting.

For the implementation of JAUS, TAKE 3 uses the JAUS Toolset (JTS), an open source software which uses a Graphical User Interface (GUI) input system to auto-generate the code for JAUS functionality. This software relies on JR Middleware to create the network connection. However, JTS is built using a different build tool than the rest of the module systems on TAKE 3. Thus, there is no direct way to interface between the JAUS networking system and ROS.

The ROS/IOP bridge is an open-source tool that, as the name suggests, creates a bridge between an IOP communication node and a ROS robot control system. The bridge allows the JAUS components to access data about the robot's status, position, velocity, and more, to be used in communications with an external controller or other entity. Also JAUS commands can be send from the OCU/CVT for controlling the robot. The main goal of JAUS is to structure communication and inter-operation of unmanned systems within a network. A JAUS system is made up of subsystems connected to a common data network. A Subsystem typically represents a physical entity in the system network, such as an unmanned vehicle or operator control unit, in our case the Segway robot. The JAUS network is further subdivided into hierarchical layers. There are two types of communicating nodes, the sender and client (receiver). As such, the bridge contains methodology for creating plugins for both types of nodes. The message is parsed by the JAUS plug in and sent to the ROS/IOP bridge. The bridge then maps the data and commands contained therein out to the correct ROS topics and publishes them, thereby completing the bridge. The ROS/IOP-Bridge consists of a lot of independent components. All services of the ROS/IOP Bridge are implemented as a plugin. We can configure to use all services in one component or in a lot of independent components. The system that we have implemented includes only one component, which will contain all the services needed for the competition. Each service is defined in the SAE standards as a finite state machine, which is implemented by the ROS-IOP bridge as a plugin.

Innovations

Hardware

E-Stop Controller

Teams competing in IGVC are required to have a hard E-stop on the base of the robot, as well as a remote-control E-stop that the judges will hold during the competition. This is standard for all teams. TAKE 3 and Detroit Mercy took wireless e-stop a step further with a controller that is efficient, versatile, and easy to use. The controller has modes for E-Stop, remote control, balance mode, tractor mode, and DTZ/Emergency Shutdown functionalities. In addition, the controller also commands the LEDs on TAKE 3. Likewise, the software in the controller acts as a fail-safe: TAKE 3 will not run without a successful connection with the controller, and if the controller were to disconnect for any reason, the E-stop is immediately activated. Lastly, the controller is easy to reprogram for any additional use the robot might need.

Software

IOP System

The Interoperability Profiles (IOP) system designed for the challenge works as a self-contained software system. It reuses several functions from the main Segway control system by reading from the existing ROS topics, especially those broadcasting sensor and position data. It also incorporates several unique functions, which help to integrate with the robot control system.

The innovation in JAUS is in the way that the system was created. We used several opensource tools, which we modified to meet the specifications of our system. The most significant roadblock was with incorporating the ROS/IOP Bridge software. This software was designed for a different version of ROS and Ubuntu than those being used for the rest of the robot's systems. Therefore, we had to adjust the code to meet our specifications.

Right and Left Lane Differentiator

The right and left lane line differentiator is an algorithm that differentiates the two lane lines. This is important in determining if the robot is driving in the correct direction. The differentiator requires an initial condition of seeing at least one lane line and facing within 90-degrees of the forward direction. The algorithm uses connected component analysis to determine if the bottom of the lane is in the right or left half of the image. This, along with knowing the angle of the line, can determine if a lane line is a right or left lane. The differentiator converts the image to a laser scan and publishes it to a cost map for each lane. Having a cost map for each lane is important in correcting failures in the detection, because the histories of the lanes are known. Ultimately, this algorithm will be used to keep the robot driving forward throughout the course.

Failure Points and Resolutions

Hardware

Segway Electrostatic Battery Discharge

While working on TAKE 3, there were system faults appearing claiming a completely dead battery, manifested by a loud beeping and fault logs. The measured battery voltage proved this false by showing as fully charged. After contacting tech support for the Segway base, a software

reset tool was acquired which reset the faults, and if the problem occurs again we will understand it. We have learned to check the battery voltage, as well as the necessary measures to turn off the beeping noises to continue full functionality.

Velodyne LiDAR

The LiDAR has a minimum range of 40 cm. When an object is within 40 cm of the LiDAR, it will be cleared from the map, if clearing privileges are given. This is a problem, because the global planner hugs corners (typically from barrels marked in the map) to minimize the distance of the planned path. When the barrel is within 40 cm of the robot, the barrel will be erased from the map and a new path will be created. Without the object in the map, the path planner will tell the robot to drive through the barrel. One possible solution is to increase the inflation radius of objects in the map to greater than 40 cm. This, however, could block potential paths. A better solution in this situation is to simply remove the clearing privileges from the LiDAR.

With the clearing privileges removed from the LiDAR, the barrels will not be removed from the map, even when the LiDAR is not reporting that it sees them. Removing clearing privileges from a sensor is not typically a proper solution to integration problems between perception and mapping. This is because if the robot sees an obstacle while its localization has a failure, the robot will mark the obstacle in an incorrect location in the map. If clearing is removed, the obstacle will be reported in more than one location in the map, and the problem cannot be fixed. While it is a risky decision to remove clearing from the lidar, it was determined that the shape and size of the objects detected by the lidar would not cause considerable marking to the map with the minimal amount of odometry slippage that TAKE 3 has.

Software

Right and Left Lane Differentiator

The lane differentiator has a likelihood of failure when the robot is perpendicular to the lane. A failure here can cause the robot to think that it is facing the opposite direction, making the robot want to face backwards. This can be fixed by comparing the output of the differentiator with the history of the right and left lanes from their individual costmaps.

Testing and Debugging

TAKE 3 has gone through rigorous testing and debugging. One of the bonuses of using ROS is that it lends itself to debugging. ROS enables the printing of topics to verify the actual vs. expected values of our modular systems, allowing us to check values in real time and handle deficiencies in code proactively as they appear.

In relation to JAUS, setting up the multicast network for testing proved an area of difficulty. The university public wi-fi network, which we used for initial testing, did not allow multicast communication. We checked to ensure that the router on TAKE 3 would allow multicast communication, then we obtained a separate wi-fi router to test in a way that would not monopolize use of the robot.

Conclusion

The innovations and technology that surround TAKE 3 are already being widely used in the automotive industry. Our exposure during school gives us an advantage to other students, as it gives us a chance to practice lifelong learning. Lifelong learning is an important component of an

engineering education; being able to practice it in school makes us more adaptable to new technologies and processes. In this project we have encountered a variety of new technologies that have forced us to practice this skill. Additionally, to complete the competition tasks we have had to read and follow a variety of technical standards, another important transferrable skill that will follow us as we move on to industry jobs. The automotive industry, along with other industries that employ electrical and robotics engineers, is replete with standards and other sets of technical rules to be followed. By ensuring that we are able to understand and follow these rules, we prove that we can function successfully in a highly regulated industry.