Constrained Motion Planning System for MRI-Guided, Needle-Based, Robotic Interventions

Christopher Bove
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Constrained Motion Planning System for MRI-Guided, Needle-Based, Robotic Interventions

by

Christopher Bove

A Thesis

Submitted to the Faculty

of the

WORCESTER POLYTECHNIC INSTITUTE

In partial fulfillment of the requirements for the

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in

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by

Christopher Bove

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Abstract

In needle-based surgical interventions, accurate alignment and insertion of the tool is paramount for providing proper treatment at a target site while minimizing healthy tissue damage. While manually-aligned interventions are well-established, robotics platforms promise to reduce procedure time, increase precision, and improve patient comfort and survival rates. Conducting interventions in an MRI scanner can provide real-time, closed-loop feedback for a robotics platform, improving its accuracy, yet the tight environment potentially impairs motion, and perceiving this limitation when planning a procedure can be challenging. This project developed a surgical workflow and software system for evaluating the workspace and planning the motions of a robotics platform within the confines of an MRI scanner. 3D Slicer, a medical imaging visualization and processing platform, provided a familiar and intuitive interface for operators to quickly plan procedures with the robotics platform over OpenIGTLink. Robotics tools such as ROS and MoveIt! were utilized to analyze the workspace of the robot within the patient and formulate the motion planning solution for positioning of the robot during surgical procedures. For this study, a 7 DOF robot arm designed for ultrasonic ablation of brain tumors was the targeted platform. The realized system successfully yielded prototype capabilities on the neurobot for conducting workspace analysis and motion planning, integrated systems using OpenIGTLink, provided an opportunity to evaluate current software packages, and informed future work towards production-grade medical software for MRI-guided, needle-based robotic interventions.
Acknowledgements

I would like to express my gratitude to Professor Fischer for his assistance scoping the project, guiding the feature development, and offering alternative solutions when I hit some tough walls to climb. Professor Michalson and Professor Li provided excellent feedback as I worked through and presented the thesis formally, helping refine the end product in a flurry of activity in the closing months. I also need to extend my thanks to the other graduate students I worked with in AIM Lab, Paulo Carvalho, Radian Gondokaryono, Katie Gandomi, Christopher Nycz, and Joe Schornak, for helping me work alongside them and making the project much more entertaining.

A tremendous thanks goes out to my parents for supporting my foray into robotics during my childhood. From tearing apart VCR’s to cluttering the basement with model airplanes, dissected Furby’s, and walking mechanical ostriches controlled by RockBand guitars, I owe you both a lot in the patience department, especially in your tireless support of helping me start FIRST teams in high school which inspired an education at WPI that culminated in my Master’s degree. Thanks for cheering me on when things seemed sometimes hopeless on the path to my first job and always encouraging me to hold fast to the hope that is found in Christ.

My time at WPI wouldn’t have been the same without my sister, Eva, who decided to follow me up here for her studies, and I’m so appreciative of the wonderful friendship we’ve experienced over our lifetime. I’d like to thank my grandmother, Betty, for all her support of my education and being the closest relative from which I could steal a meal and Rita’s Water Ice dessert from (spoiled). My thanks to Cara for always encouraging me to do my best on this project and tremendously improving my cooking skill throughout the year.
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<td>Automation and Interventional Medicine Research Laboratory</td>
</tr>
<tr>
<td>WPI</td>
<td>Worcester Polytechnic Institute</td>
</tr>
<tr>
<td>DOF</td>
<td>Degrees of Freedom</td>
</tr>
<tr>
<td>FK</td>
<td>Forward Kinematics</td>
</tr>
<tr>
<td>HIFU</td>
<td>High Intensity Focused Ultrasound</td>
</tr>
<tr>
<td>IK</td>
<td>Inverse Kinematics</td>
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<tr>
<td>KDL</td>
<td>Kinematics and Dynamics Library</td>
</tr>
<tr>
<td>LTS</td>
<td>Long Term Support</td>
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<td>MOCAP</td>
<td>Motion Capture</td>
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<td>MRI</td>
<td>Magnetic Resonance Imaging</td>
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<tr>
<td>MRTI</td>
<td>Magnetic Resonance Thermal Imaging</td>
</tr>
<tr>
<td>Neurobot</td>
<td>Neuroablation Robot</td>
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<tr>
<td>OMPL</td>
<td>Open Motion Planning Library</td>
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<tr>
<td>OpenRAVE</td>
<td>Open Robotics Automation Virtual Environment</td>
</tr>
<tr>
<td>PCL</td>
<td>Point Cloud Library</td>
</tr>
<tr>
<td>Pose</td>
<td>Position and Orientation</td>
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<td>RCM</td>
<td>Remote Center of Motion</td>
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<td>ROS</td>
<td>Robot Operating System</td>
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<td>RVIZ</td>
<td>ROS Robot Visualizer</td>
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<td>SRDF</td>
<td>Semantic Robot Description Format</td>
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<td>STL</td>
<td>STereoLithography - 3D triangle mesh file</td>
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<td>TF</td>
<td>ROS Transform Library</td>
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<td>URDF</td>
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<td>Virtual Machine</td>
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Chapter 1

Introduction

Since only, perhaps, a couple of decades ago, the fields of medicine and robotics remained relatively separate practices, only associated through the common technology utilized by both of them. Yet in responding to increasing needs of healthcare providers for faster, more cost-effective, safer, and more accurate treatments, the solution space has moved towards robotics systems, where, in the industry and beyond, robotics and autonomous systems have been improving performance in procedures which were very labor and time intensive while maintaining very high precision and repeatability requirements. In some respects, robotics and medical problems are similar, and the expansive development in both fields has illuminated the strong benefit in joining those two realms of expertise to better solve the challenging issues facing them today.

One of the ways in which robotics has forayed into the medical world is through assisting surgical interventions. As discussed in the following sections, there are a variety of spaces in which robotic medical devices can benefit and improve treatment outcomes compared to non-robotic clinical procedures. This becomes evident when examining the potential for robotics to improve surgeons’ perceptive and manip-
ulative capability in minimally invasive surgical procedures. The contributions of this work, discussed in the final section of this chapter, were to design and build a research-grade software system capable of interfacing with medical technology that provides computation and display of a robot’s workspace to better inform treatment plans and provide movement to treatment sites while avoiding collisions with the patient. While this system was tailored to a particular robotic platform for neurosurgical cranial procedures, it provided some research frameworks and methodologies adaptable for similar robotics platforms conducting needle-based interventions under the constraints and guidance of imaging systems such as MRI. Procedures that could benefit from this work would be those using rigid-body robotic manipulators to perform tool insertions into a patient, as specified by a human operator, during treatments that occur within an MRI scanner or under other environmental constraints. This could include prostate biopsies, deep brain stimulation, tumor ablation, tumor biopsies, or using robotic manipulators for laparoscopic surgery.

1.1 Clinical Problem

One of the goals during surgical procedures is to minimize healthy tissue damage while remaining effective at performing the procedure. Minimally invasive surgical techniques can mitigate the risk of infection, blood loss, tissue damage, cosmetic scarring, and improve recovery time by reducing the size of the incision made in the skin and exposure to the environment. [1] However, this usually comes at the cost of diminished perception, manipulative ability, and surgeon comfort, and the associated cost increases that can come with these issues. While the quality of the surgery is better than traditional procedures, this raises questions as to the broader societal implications of minimally invasive surgical techniques. [2] In regaining these
losses, other technologies have been developed to compensate for these drawbacks and make minimally invasive surgeries more desirable for both surgeons and patients, to the extent that some robotics interventions can offer advantages over conventional laparoscopic surgeries, such as Intuitive’s da Vinci Surgical System used for general laparoscopic surgery.\[3\] Providing feedback for the surgeon, such as through the use of imaging devices, and more comfortable system operation are keys for improving the performance and utilization of minimally invasive surgical platforms. Surgeon comfort/ergonomics and perception are just some of the factors influencing the usage of such systems; others include speed, cost, accuracy, outcomes, reliability, workflow, and convenience. For a system to be successfully adopted and employed in practice, it needs to balance these factors in a way that makes it more enticing and decidedly of greater advantage than existing alternatives.

Several interventions for brain cancers and malevolent tumors involve removing as much of the tumor as possible before sometimes utilizing further treatments, such as radiation or chemotherapy, to destroy the remaining unhealthy tissue. Of course, radiation used to kill cancer cells can also harm surrounding healthy tissue, and care must be taken to minimize that exposure. A different approach in contrast to surgical removal is to kill unhealthy tissue through thermal ablation, which heats up and holds cells at a temperature to a point at which they die, and ultrasound is one method of conducting thermal ablations.\[4\] The ultrasonic source can be located internally or externally to the patient, which is further discussed in subsection 2.1.1 and High Intensity Focused Ultrasound (HIFU) externally delivered has been used to successfully treat brain tumors while in MRI scanners.\[5\] However, internally delivering high intensity ultrasound treatment through a probe or catheter offers some advantages over externally applied treatments, and the use and development of devices such as ultrasonic probes to deliver this interstitial ultrasound treatment
is a current area of research. [6] [7]

The challenge with ultrasonic ablation treatment methods utilizing probes is how to guide the probe to the treatment location effectively. While traditional procedures could involve significant removal of the skull to access the tumor area, ultrasonic probes on the other hand are small enough to allow for minimally invasive stereotactic procedures, in which only a small burr hole is drilled in the skull and 3D imaging is used for instrument guidance, which can reduce healthy tissue trauma, improve accuracy, and speed recovery. [8] Recent research has shown that MRI-guided ultrasonic ablation is a viable approach for treatment of brain cancers. [5] By conducting the experiment within an MRI scanner, the temperature change of tissue during thermal ablation can be monitored using a special type of MRI scan (MRTI), discussed further in section 2.1.2, to ensure that the malevolent cells have received sufficient temperature levels to terminate them.

As stated before, one challenge with minimally invasive surgery is that the tooltip cannot be directly observed by the surgeon, which complicates positioning the instrument and reaching the target area successfully. Rather, the tool can only be tracked through some form of imaging feedback or by tracking the movements of the base or handle located outside of the patient. While a camera could be inserted for some surgical procedures, most needle-based interventions are anatomically prohibitive of providing empty space around the treatment location for a camera to see anything useful. CT scans and X-Ray offer alternative means for viewing and guiding instruments inside the body, but the health risks of harmful radiation exposure to the patient and surgeon during those imaging processes necessitates their use only when such risks outweigh the potentially life-threatening conditions being treated. Additionally, both CT and X-Ray imaging techniques are not very suitable for imaging soft tissue. However, MR or ultrasound imaging techniques do not emit
harmful radiation and are good at imaging soft tissue, yet for the case of MRI and CT scans, the procedure must be conducted in an image scanner for the image feedback in contrast to X-Ray and ultrasound which have less constraining, less bulky, and physically more flexible imaging instruments. For procedures in MRI scanners, use of non-ferrous tools and MRI-compatible electronics are necessary. However, the physical properties of an MRI bore make it very challenging ergonomically for a surgeon to perform the procedure with real-time image feedback for guidance. There are some MRI scanners which allow access in the center, but these are still challenging to work with. [9]

Alternatively (or in addition to having imaging feedback), the tool can be guided using a mechanical device. Frames can be rigidly attached to the skull to precisely insert the instrument to the desired location [8], but significant time is required to setup these frames, register their coordinate system with the patient, and verify correct alignment. Additionally, there is some healthy tissue damage during the attachment of such a stereotactic frame to the head. An example of this type of frame is shown in Figure 1.1. Once again, using non-ferrous metals in the stereotactic frame or in other similar guidance systems can allow surgeries to be conducted in an MRI scanner for imaging feedback or to be utilized for validating the tool insertion and setup.

1.2 Robotic Solution

A potential alternative to manually aligning stereotactic frames for tool insertion is to achieve the same motion control with a robotic system. Such a system could obtain the tool location using the forward kinematics of the robot, registration of the robot to the imaging system, and MR images of the probe to precisely position the
Figure 1.1: A photograph of an example stereotactic frame that could be used for interventions and treatments during minimally invasive neurosurgery. [10]

Ablation tool at the treatment site. This promises significant speed and reliability improvements over manual tool alignment. Past research at WPI AIM Laboratory produced a five DOF neuroablation robot, called the neurobot, which was MRI compatible, lacking ferrous material, and did not distort the image significantly while operating its piezoelectric motors to move. [11] Kinematically, the system produced Cartesian translation from the base and orientation through a remote center of motion, emulating the workspace of stereotactic frames. Newer iterations of this system provide two extra degrees of freedom for the robot to perform the needle insertion along the insertion axis and to rotate the ultrasonic probe to enable more precise coverage of tumors once the probe is inserted. [12] Figure 1.2 shows the latest version of the robot as a SolidWorks model.

1.3 Problem Description

As discussed earlier in section 1.1 some of the primary challenges when working with minimally invasive surgical techniques are perception and maneuverability limit-
tions. For the neurobot, these problems manifest themselves in two primary man-
ners: 1) it is difficult for operators to plan procedures with the robot without know-
ing what it can reach and 2) the robot needs to be able to move to the procedure
location without striking obstacles. In other words, it is necessary to know where
the robot can reach and how to move it there, and there are some complications of
solving these seemingly simple problems.

1.3.1 Differentiation from Related Work

Other work has focused on motion planning in MRI machines as a control problem
by steering needles in soft tissue, [13] modeling needles as serial chains for conduct-
ing motion planning in task space, [14], and conducting workspace design analysis
to optimize the robot’s designed workspace within the MRI bore. [15] [16] These
works are associated with the topics of this one but do not provide a solution for
analyzing the workspace in real-time by checking collisions with the scanner based
on operative registrations. Motion planning techniques in the literature specialize in
trajectory following for needles typically used in MRI as these holonomic problems
are challenging to solve and the needle dynamics are complicated, but conventional rigid body manipulators do not appear to have been studied for workspace analysis and use in MRI scanners. Thus, this work seeks to develop a generic framework and approach for conducting workspace analysis and motion planning for a manipulator confined to an MRI scanner, yet apply this framework specifically in an implementation targeted for the neurobot while allowing for the possibility of adaptation to other platforms.

1.3.2 Workflow Overview

Current neuroablation procedures using the neurobot have a generally straightforward workflow for performing ablations, which is outlined below:

1. Take preoperative images of the patient head.
2. Attach robot to scanner bed.
3. Register robot in MRI scanner. Retract scanner bed from bore.
4. Drill burr hole in skull for access to target site.
5. Secure patient on scanner bed and insert into bore.
7. Surgeon selects entry and target points.
8. Robot moves to entry point.
9. Cannula then probe inserted into patient with the robot.
10. Scanner bed inserted back into scanner.
11. Ablation treatment performed, repositioning and repeating as necessary

In the future, Step 4 of the workflow could be performed using the robot itself later in the workflow by equipping it with a drill, while in the current iteration, the surgeon completes the drilling. It would be ideal for the burr hole location to
be chosen so as to optimize the robot’s manipulability around the target zone, and having the robotic system inform this selection process and perform the drilling can further improve the workflow and procedure outcomes.

1.3.3 Potential Problems in the Workflow

Once the neurobot is attached to the patient table on the MRI scanner (Step 2), it faces several constraints that can interfere with achieving the desired needle insertion into the patient’s head to place the ultrasonic probe at the treatment location (Step 9). Some of these are inherent due to the robot design itself while others are environmental constraints. A few of the robot’s joint limits change based on the motions of other joints (i.e. the amount of axial travel depends on the current height due to a trapezoidal/scissor lift mechanism). This complicates the modeling of the robot’s kinematics in conventional robotics software packages and means that extra precautions must be taken when planning its motions and ensuring treatment locations are reachable.

Concerning the environment, there are two primary objects the robot and its end effector and attached instrument can unintentionally collide with and must avoid: the patient and the scanner. When the patient is secured on the scanner bed (Step 5), the robot must move to the entry point in the skull without striking the patient’s head (8). This is aided by removal of the ultrasonic probe and cannula, as shown in Figure 1.3.

Once the robot is in place above the burr hole, the cannula is inserted into the skull and attached to the robot, and the ultrasonic ablation probe is attached to the robot and inserted into the brain by the robot’s needle driver joint (Step 9). This final robot configuration must then be able to enter the bore of the MRI scanner without the ultrasonic probe or robot collide with the scanner as the scanner bed
is inserted back into the scanner bore (Step 10). Therefore, a successful insertion solution must abide by the robot’s kinematic limitations, avoid collisions with the patient’s head and skull during movement along a trajectory from the starting configuration to the final configuration (with the cannula and probe removed), has a final configuration (with cannula and probe attached) that does not collide with the MRI bore, and reaches the target destination successfully.

To meet this need currently, the neuroablation robot requires the coordination of two operators to drive the robot joint-by-joint to prevent collision with the patient head and align the system to the desired insertion point. This process is time consuming and requires knowledge of the robot’s kinematic limitations, increasing the difficulty for medical professionals to perform this procedure themselves. Further, the workspace of the robot is not visualized before the alignment is attempted. Given the placed burr hole in the skull, it would be very useful to know where the robot can reach and administer treatment, but this ability has not yet been realized. This limitation can lead to costly situations in which time is spent attempting to
align and reach a position that is outside the robot’s workspace, in which case the patient may need to be repositioned and the workflow repeated.

### 1.4 Contributions

This project entails developing a research-grade software system which integrates existing medical platforms, provides a mechanism for evaluating the neurobot’s reachable workspace, and generates collision-free trajectories to enable faster, less tiresome, and more intuitive surgical procedures using the neurobot in ultrasonic ablation with flexibility to be adapted on similar platforms. The specific contributions of this work include:

- Designing and implementing a software system for conducting motion planning
- Modeling the neurobot in and interfacing it to the ROS environment
- Generating the collision-aware workspace of the robot
- Computing the collision-aware treatment workspace given an entry point
- Computing collision-free trajectories to control the robot’s motion
- Utilizing ROS-OpenIGTLink-Bridge to transfer data between the robot, software system and slicer
- Employing 3D Slicer to visualize and interact with the generated data
- Integrating a workflow to segment patient body MRI into a collision object
- Evaluate system performance on the neurobot
- Demonstrate the system in animal trials

Known techniques from both medical and robotics fields are applied into this integrated system to produce a significant contribution for the problem described in [section 1.3](#). This work does not introduce a novel approach for motion planning or way of interfacing medical devices. Rather, it is an exercise in using current
state of the art software within a new research-grade system to produce useful functionality with reduced development overhead. The utilized toolsets are also evaluated to perhaps illuminate some problem areas or technical stumbling blocks for other researchers wishing to accomplish similar objectives. In the end, this work advances the capabilities of the neurobot, making it easier to operate in the hope of improving surgical procedures and the outcome for patients suffering from cranial diseases.
Chapter 2

Background and Related Work

As a project that spans several technical fields, there are a number of topics that need further examination to better understand the current state-of-the-art and how this project relates to those developments. In this section, current medical technology, neuroablation procedures, the Neurobot platform, robotics tools including ROS, and motion planning concepts will be explored before the system design is discussed.

2.1 Medical Technology

The medical field has several key technologies that are closely entwined with the contributions of this project. This includes ultrasonic ablation, MRI, and medical imaging software tools.

2.1.1 Ultrasonic Ablation

The process of ablating, or removing, tissue can be achieved through different means, with heat being one option in a process known as thermal ablation. Ultrasonic thermal ablation involves utilizing ultrasonic energy to heat up surrounding tissue
to the point at which it dies while leaving healthy tissue at safe temperatures due to the narrow field of applied ultrasound. Once tissue has been exposed to high enough temperatures for a certain length of time, as shown in Figure 2.1, tissue will experience necrosis, or death.

![Figure 2.1: Graph of tissue reaction to heating at particular durations of time. From [17].](image)

By modulating the length of time and exposure to temperatures, a very precise treatment (assuming sufficient monitoring and control) can be achieved which leaves healthy tissue unexposed to dangerous temperature levels for long lengths of time while fully destroying unhealthy tissue. The key to attaining this ideal treatment is through the monitoring process, which can be attained through MRTI, discussed in section 2.1.2.

As previously mentioned, ultrasonic energy can be administered externally or internally to the patient. Outside of the patient, the ultrasonic beam is focused to a particular depth such that the focused energy of the ultrasound traveling through the body heats the target tissue to the point of necrosis, and this technology is
known as High Intensity Focused Ultrasound (HIFU). Inside of the patient, ultrasonic treatment can be delivered through an element placed on a catheter or needle, referred to as interstitial ultrasonic ablation, and this method can provide conformal tumor coverage. By conducting a series of interstitial ablations, using different ultrasonic probes with limited fields of effect, or by rotating a probe with a non-symmetric ultrasonic beam, arbitrary tumor shapes and sizes can be effectively ablated. Planning and conducting these probe rotations and insertions to achieve proper treatment coverage is a goal of current research, and MRI provides a key mechanism for enabling that monitoring capability. Ultrasonic probes are typically MRI-compatible and cooled with water, which can increase their size due to piping, as shown in Figure 2.2.

Figure 2.2: A SolidWorks render of the ultrasonic probe used in thermal ablation on the neurobot.
2.1.2 Magnetic Resonance Imaging

Magnetic resonance imaging (MRI) is a non-invasive imaging technique which utilizes a very strong magnet and radio frequency to image soft tissues in the body, providing 3D images when slices are combined together. Essentially, hydrogen protons are spin-aligned to the magnetic field in the scanner, and radio waves are used to temporarily realign the protons while the radio is on. When the radio is turned off, the protons realign with the permanent magnetic field, releasing energy as they do so, and sensors in the scanner can resolve this energy release into 3D positions within the scanner. Based on a number of factors, differences in tissues (usually molecular density including water) can be detected and seen in the MR images, as seen in the image of the pig brain in Figure 2.3.

Figure 2.3: Example MRI image of a pig head, showing the brain and surrounding tissue.

Because MRI’s soft tissue contrast is much higher than X-Ray or CT scans and there is no ionizing radiation emitted (which can cause harmful DNA mutations), it is a preferred imaging method, yet more expensive, for cases where repeated
images need to be taken to guide procedures. However, the tight confines within an MRI scanner (most are between 55-70cm in diameter [20]) makes scanning patients sometimes challenging and also hinders efforts to conduct medical interventions within the scanner.

The scanner located at the UMASS Memorial Medical Center, the primary location for current tests with the Neurobot, is a Phillips Achieva 3T scanner with a 60 cm bore.

Frame Conventions

MRI scanners typically follow the ISO Standard for frame conventions, which describe the orientation of images and the patient position within, as Figure 2.4 illustrates. [21] While there are some specialized naming conventions for the patient (anterior and posterior, left and right, and superior and inferior), these are generally straightforward frame definitions. Within the scanner itself, due to the cylindrical nature of the bore, the terminology of lateral and axial motions describe movements in the x direction and z direction respectively. The neurobot utilizes the RAS convention as its coordinate frame, where right is positive x, anterior is positive y, and superior is positive z as shown in Figure 2.4

Magnetic Resonance Thermal Imaging

MRTI is a scanning technique for MRI machines which effectively allows for real-time monitoring of changes in tissue temperature, which is of use particularly during ablation of tumors in the body. [23] This analysis can be conducted on a number of simultaneous or sequentially captured 2D slices in the MRI (either in parallel or orthogonal 2D planes) and serves great utility for monitoring the progress of thermal ablation procedures in that the ablation process can be terminated once the tumor
tissue has received adequate heat so as to minimize damage to surrounding, healthy tissue. 
Figure 2.5 shows an example of MRTI and tissue necrosis analysis on an ablation procedure.

Figure 2.5: A) Image showing temperature change as obtained using MRTI techniques during an ablation. B) The computed zone of tissue necrosis in the brain around the tumor. From [24]
2.1.3 Medical Imaging Software

While commercial MRI scanners typically ship with the manufacturer’s own software solution for running the scanner and collecting and viewing images, there is a growing interest in providing open-source methods for analyzing MRI’s without proprietary software and creating open standards for interfacing with devices. These two desires have been addressed by the development of OpenIGTLink and 3D Slicer, open-source software which has greatly assisted recent medical research.

OpenIGTLink

OpenIGTLink is a communications protocol which standardizes the interfaces of scanners, surgical systems, and analytical tools on computers. [25] [26] Essentially, this could do for the medical research world what ROS did for much of robotics research (see section 2.4). The specifications for the OpenIGTLink protocol allow equipment from different manufacturers to be compatible (or at least allow compatible drivers to be written easily) to reduce the development overhead of researchers using these systems for planning, image collection, and image analysis. It supports real-time communications with data types specifically designed for and needed by medical imaging systems.

3D Slicer

3D Slicer is an open-source computer application for viewing and processing medical images. [27] [28] It includes a large assortment of downloadable extensions from a repository contributed to by a growing community, and it is relatively straightforward to write custom extensions or modules in C++ or Python. The main goal of Slicer is to fulfill a need of researchers and the general public to have a free tool for accessing medical images and to serve as a platform for conducting experimental image
processing and research for image-guided interventions. It includes an OpenIGTLink extension, which effectively allows it to interface with any OpenIGTLink-supported devices. The interface yields itself easily for interacting with imaging data and includes many processing techniques useful during pre-operative planning and during surgical procedures. Figure 2.6 shows a screenshot of the 3D Slicer application.

Figure 2.6: Screenshot of 3D Slicer application being used to view MR images of a pig head with a volume rendering using ray casting in the top right corner.

2.2 AIM Lab Neuroablation Robot Platform

2.2.1 System Overview

The neurobot is a seven DOF robot with piezoelectric motors featuring encoder feedback shaped in a form factor capable of fitting within an MRI bore, as shown in Figure 2.7. As stated previously, the current iteration has stemmed from a series of developments in WPI's AIM Laboratory. Electrical components are designed to
limit the noise introduced in the scanner environment, which can degrade the quality of MRI. This is especially apparent with the choice of motors; since most motors rely on magnetic fields to operate, traditional brushed or brushless motors cannot be used inside an MIR machine. Alternatives are available, such as pneumatic steppers and piezoelectric motors, which utilize piezo elements which flex under electrical voltages to achieve motion to drive shafts. [29] The mechanical components are iron-free and MRI compatible, which allows the robot to operate in the MRI freely, and it can undergo proper sterilization and covering procedures during surgery.

Figure 2.7: Rendering of neurobot SolidWorks model in cut-away MRI scanner bore.

The robot is connected via a copper wire tether to a control box which resides in the MRI scanner room. Motor control boards and an embedded microprocessor with an FPGA provide the low-level functionality for the robot, including position control, safety features, and communications. WPI custom built most of the components in the interface box, and the system setup in the MRI console room is shown in [Figure 2.8]. The interface box is connected to a pneumatic safety foot pedal, which enables robot motion when the operator presses it, a fiber optic cable for communication, and a power cable to a standard 120 volt wall outlet. Only the fiber cable leaves the MRI scanner room through a barrier wall into the operator control room. Here, a computer can be connected to the robot to control it, and other computers...
interface with the MRI scanner to configure scans and obtain images.

Figure 2.8: Photograph with annotations showing the interfaces between the robot, control computer, and MRI scanner in the console room. The computer monitor shows a camera feed from the scanner room on the other side of the wall. From [30]

2.2.2 Coordinate Frames

Most of the neurobot frames follow similar conventions as those used in the scanner itself. Figure 2.9 show the primary coordinate frames of the robot: the tool frame and the base frame, which is defined by the z-frame attached to its base. The z-frame is a specially designed fixture for image-based registration in an MRI scanner. The structure holds vials of MRI-visible liquid at geometrically aligned locations in which performing image analysis on scans of the z-frame allow the centroid of the frame to be identified relatively quickly and accurately. [31] This z-frame then serves as the registration between the scanner frame and the robot. The optical frame of the scanner is simply a translational offset from the base frame. Since the robot is always situated in the scanner in the same orientation, movements in the base frame x direction are considered lateral and movements in the z direction are considered axial.

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2.2.3 Kinematics and Joint Information

The neurobot is essentially a Cartesian platform base with a remote center of motion wrist which also includes a degree of freedom for driving the needle along its axis, as illustrated in Figure 2.10 totaling seven active DOF. This makes the system under constrained in some configurations. Cartesian motions of the base are achieved through screw drives, with lateral translation being a simple screw driven joint. Vertical and axial motion is obtained through a differentially driven pair of lead screws attached to a parallelogram mechanism. When both screws are driven in the same direction, this achieves axial translation. When screws are driven in different directions, the ‘legs’ of the parallelogram are brought closer together or further apart, which achieves a change in height.

Rotational motion of the upper assembly of the neurobot is attained through
(a) Diagram showing remote center of motion pitch joint, with yaw joint along same axis.

(b) Parallelogram structure composing cartesian base on the neurobot for both vertical and axial insertion joint and rotation on the probe.

(c) Needle driver joint showing insertion joint and rotation on the probe.

Figure 2.10: Neurobot pitch, parallelogram, and needle driver mechanisms. All images from [12]
revolute joints. Changes in yaw are rendered by a single point of actuation to rotate the entire upper assembly. To achieve pitch changes about the remote center of motion, a linkage translates rotations at the base near the yaw joint to that about the RCM. The needle driver is a linear screw prismatic joint which can translate the needle through the hollow cannula a total travel depth of 4mm. Finally, a gearing at the top of the needle allows the needle to rotate continuously, useful for non-symmetrically beamed ultrasonic probes, in order to aim the treatment appropriately. Table 2.1 summarizes some more information considering the joints on the neurobot, while [12] contains a detailed and complete explanation of the kinematics of the neurobot.

Table 2.1: Joint motions and limits on the neurobot, with frames as defined in Figure 2.9

<table>
<thead>
<tr>
<th>Axis</th>
<th>Motion</th>
<th>Min Value</th>
<th>Max Value</th>
<th>Units</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>tx</td>
<td>-37.5</td>
<td>0</td>
<td>mm</td>
</tr>
<tr>
<td>2</td>
<td>ty</td>
<td>0</td>
<td>44.23</td>
<td>mm</td>
</tr>
<tr>
<td>3</td>
<td>tz</td>
<td>-86</td>
<td>0</td>
<td>mm</td>
</tr>
<tr>
<td>3</td>
<td>tz2</td>
<td>-143</td>
<td>57</td>
<td>mm</td>
</tr>
<tr>
<td>4</td>
<td>Rx</td>
<td>-90</td>
<td>0</td>
<td>deg</td>
</tr>
<tr>
<td>5</td>
<td>Ry</td>
<td>-37.2</td>
<td>30.6</td>
<td>deg</td>
</tr>
<tr>
<td>6</td>
<td>Rz</td>
<td>-Continuous</td>
<td></td>
<td>deg</td>
</tr>
<tr>
<td>7</td>
<td>P</td>
<td>-40</td>
<td>0</td>
<td>mm</td>
</tr>
</tbody>
</table>

2.2.4 Software Control System

Within the interface control box of the neurobot is an embedded system with a microprocessor running Linux and an FPGA board. This board runs the joint-level position control and calibration routines for the robot, reading in encoder data and interfacing with the piezo driver cards to properly drive the axis to the proper set points. The software computes forward and inverse kinematics for the robot and can publish and receive data over OpenIGTLink.
A web interface is used to control the robot over the fiber optic network interface. Joint level commands can be inputted using this interface and sent to the robot, and the IK solution can be used to specify a desired target position for the robot and drive the joints to that state. To resolve the desired position and orientation of the end effector for IK computation, the entry point in the skull and the treatment location are specified using the interface. From there, target joint values are set and the operator can enable or disable particular joints and press the foot pedal to begin motion.

2.3 Neuroablation Procedures

Setup for a neuroablation procedure is fairly similar to other types of surgical procedures involving the brain and use of an MRI scanner. Preoperative images are usually taken of the patient to identify where the tumor (or planned treatment area) is located with respect to the patient’s anatomy. Current tests with the neurobot are being conducted with live swine subject which are kept under general anesthesia. The surgeon places a burr hole, a small hole in the skull created using a special drill that stops at the brain matter, in the skull at a location they deem to be suitable through which to insert the ultrasonic probe and to be able to reach the treatment site.

The robot is rigidly bolted to the MRI scanner bed after the scanner bed has been homed, and the z-frame registration is performed after the robot has homed itself and calibrated its joint encoders. Images are taken of the z-frame and then inputted into a computer running TheraVision which computes the registration using the fiducial vials in the z-frame. This registration is manually typed into the robot control software interface. The ultrasonic probe and cannula are removed
from the robot, then the bed is backed out of the scanner and the patient is placed and strapped down. Once the patient has been imaged, planning for the neurobot ablation can begin.

As seen in Figure 2.11, there are four different offsets used by the control software to configure the parameters for solving IK for entering the patient head and reaching the treatment location with the ultrasonic element on the probe. These offsets are user adjustable for tailoring the solution for insertion to give a safety cushion around the patient and to simplify the problem of solving the IK for the robot.

![Insertion at Home](image1.png)  
![Insertion at Target](image2.png)

(a) A: Treatment to tip offset. D: Robot to treatment at home.  
(b) B: Robot to entry. C: Cannula to element.

Figure 2.11: Treatment offset definitions used by the robot control software. Courtesy of Christopher Nycz

Without feedback for illustrating the workspace of the robot given the entry point on the patient, it is possible that when the surgeon selects an entry and target point on the patient using 3D Slicer that these do not resolve into a viable pose for the robot to reach the target. In the current procedure, the entry and target are input into the IK solver on the robot controller, and if it is unreachable, a new
target is selected, the offsets can be adjusted, or the entry location can be varied slightly. If the pose is reachable, it does not guarantee that the robot or attached ablation probe will avoid collisions with the MRI scanner once the bed is inserted back into the bore. If this occurs, then the patient may need to be repositioned so that the entry point is in a more accessible location to reach the target zone with the neurobot.

2.4 ROS Architecture and Tools

The Robot Operating System (ROS) is a software platform officially supported on Ubuntu Linux which seeks to make developing robotics systems easier by standardizing sensor drivers and data types and providing common tools on which to build software for robotics systems. [32] ROS is not an operating system by any means, but rather provides a node-based communication framework with standardized messages which are published and subscribed to using different topics. This allows a distributed system to be built relatively easily, with nodes unconcerned with what computer or robot is publishing data, and it also makes a system more abstract and modular. As long as, for instance, a sensor driver exists for publishing the standard ROS point cloud message for a certain piece of hardware, that hardware can be used as easily as any other and software becomes more hardware agnostic. This reduces the amount of work the programmer must complete to get a new system running and greatly simplifies the ability for simulations to be performed with different platforms.

Apart from the libraries used to achieve node-to-node communication and standard message datatypes and services, ROS provides several tools to simplify the coding and testing of complex machines. Robots can be modeled through the Uni-
versal Robot Description File (URDF), a type of XML document that describes the kinematic structure of a robot (including joint data such as limits, efforts, and speeds) as well as how to assemble visual and collision geometry. The URDF supports fixed, floating, planar, prismatic, continuous, and revolute joint types. There are methods for exporting 3D CAD drawings of robotics into the URDF format with some amount of work, such as the SolidWorks to URDF Plugin. With the various drivers in place for real robot, the joint state of the internal model can be updated, and other packages can use that information, together with the Transforms Library (TF) to relate various parts of the robot to its environment, for the purposes of motion planning or mapping. Finally, various visualization tools such as Robot Visualizer (RViz) and RQT (ROS Qt-based GUI for plotting data streamed over the ROS network) further simplify testing and operation of robotics platforms.

2.5 ROS OpenIGTLink Bridge

In the effort of enabling greater collaboration and reuse of software tools located in both ROS and resources such as Slicer accessible through OpenIGTLink, recent research has developed a package that bridges the two message-passing systems together, known as the ROS OpenIGTLink Bridge. This exists as an open-source ROS package, where a node is responsible for initiating an OpenIGTLink connection as either a client or server and interfaces with ROS over standard topics. Published topics expose some of the datatypes received over OpenIGTLink (postfixed with ‘IN’), and subscribed topics are those messages that are pushed over OpenIGTLink (postfixed with ‘OUT’). The following OpenIGTLink message types were implemented in this package: point, transform, polydata (for triangle meshes, line strips, polygons), string, image, and point cloud (as fiducial points).
2.6 Motion Planning

2.6.1 Motion Planning Overview

Motion planning is the process of generating valid instructions to get an object in one configuration to another configuration without violating any constraints in the workspace. In the context of robotics platforms, motion planning usually manifests itself in determining how to move a robotic manipulator from a starting configuration to an ending configuration without colliding with itself or objects in the environment, usually with the secondary goal of generating a motion which can be considered to be optimal against some form of costing.

When discussing motion planning on robotic manipulators, there are two spaces to consider: configuration space and task space. Configuration space represents the valid n-dimensional set of joint states that can be reached, where n would be the number of joints on the system and a configuration would be specified as the joint angles or positions of the manipulator. Task space is the world in which motion of the end effector/body is effected and described, and for the neurobot and people, this is a 3D Cartesian space with an additional 3 degrees of freedom of rotation and notated as SE(3) space. In most motion planning approaches, the configuration space is explored to join the starting configuration and ending configuration together in a path of valid configurations that span the configuration space. How this search is conducted and how configurations are sampled during that exploration are some of the key differentiators between motion planning implementations.

2.6.2 Motion Planning Algorithms

Classic motion planning problems have developed several solutions that adapt well to the problem of moving the neurobot to the entry location on the patient’s head.
Sampling-based approaches, such as Rapidly-exploring Random Trees (RRT) and Probabilistic Road Maps (PRM) work well for finding solutions to problems with high dimensional configuration space, as is the case in this seven DOF system. However, they most often fail to find optimal solutions and can be shown to most often find non-optimal solutions, prompting developments to ensure optimality of solutions such as the development of RRT*. Path length optimality is of some concern in this situation in order to reduce procedure time (due to the time it takes for the robot to move), but feasibility and search speed are of more importance. The relatively slow speed of the neurobot’s joints increase the benefit in using an optimal planner, yet the cost of additional planning time for the optimal solution could exceed the time it takes for a suboptimal path to be planned and executed.

Motion planning to the entry point would be completed after the robot is placed in the scanner, the patient is positioned on the bed, a planning set of MR images is acquired, and the surgeon selects an ablation solution. This means that the motion planning time directly affects the procedure time; motion planning cannot happen in parallel to other tasks. Thus it is important for the planning software to operate quickly to generate a feasible path to the desired final configuration.

In addition, uniform sampling methods to explore configuration space are often very slow to find paths through narrow passages, which could be a problem for this project as the needle needs to be guided through a small opening in the skull. However, this problem can be removed by separately considering the problem of moving to the entry point first, treating the robot as a 5 DOF system without the ultrasonic probe, and then adding the probe and simply ensuring the robot will reach the target once the probe is inserted. Nonetheless, there are several methods for biasing task space sampling to increase configuration sampling within the narrow passages, such as using workspace features that prove to be important key passageways for
finding a solution. \cite{37} Such changes to the sampling be necessary in order to speed up the search time for finding viable solutions or for better determining there is no valid solution for the given motion planning problem.

To that end, a well-developed motion planning algorithm, Kinodynamic Motion Planning by Interior-Exterior Cell Exploration (KPIECE), achieves a fast and robust solution for the motion planning problem by coupling, as the name suggests, dynamic-aware state space searching with optimizations on searching through less-explored areas of configuration space. \cite{38} KPIECE achieves noticeable computational and success rate gains compared to other algorithms such as RRT. \cite{38} In addition, an implementation of KPIECE can also allow the dynamics of the robot to be considered in finding an optimal trajectory. However, due to the high-gearing on the neurobot and the slow speed of the motors, dynamic considerations can be safely omitted for this project.

A customized implementation of KPIECE known as LBKPIECE utilizes lazy collision checking (only nodes are checked for collisions - edges are ignored until completing the final solution path) and bi-directional search (searching from both the starting configuration and ending configuration) to significantly speed searches beyond those possible with KPIECE. Utilizing OMPL (which is discussed in the next section), a team of researchers analyzed the different motion planners, and LBKPIECE was shown to be very fast compared to the others but suffered from slightly longer than optimal path lengths. \cite{39}
2.6.3 Motion Planning Libraries

Open Motion Planning Library

Some of the same authors of KPIECE developed the Open Motion Planning Library (OMPL), which provides several motion planning algorithm implementations that can be instantiated easily with minimal interfacing to reduce development time. Once methods are provided for collision checking and describing the configuration and task space, many of OMPL’s motion planning implementations can be used relatively easily. The source packages can be installed on Linux, Mac, and Windows, but other dependencies are needed to enable collision checking and allow the system to compile properly on some platforms.

MoveIt!

MoveIt! is the ROS package for motion planning, which utilizes OMPL and wraps it in a ROS-compatible layer to support motion planning objectives for robots in the ROS environment. By default, MoveIt! utilizes the Orocos Kinematics and Dynamics Library (KDL), which can describe develop numerical IK solutions for a number of different kinematic systems. KDL also provides collision checking with the robot itself and the surrounding environment, which, in ROS, can be specified as several different types of geometric primitives, octomaps, STL meshes, or inferred through real-time depth maps.

MoveIt! includes a setup assistant for taking a preexisting URDF and turning it into a MoveIt! motion planning package. This package includes several YAML files which describe the OMPL planning and KDL parameters and a Semantic Robot Description Format (SRDF), which contains extended information beyond the URDF about the robot for the purpose of describing motion planning groups, end effectors,
and collision checking optimizations. A MoveIt! motion planning package generated for a robot can then be used to instantiate instances for planning motions through Python and C++ API’s and through a plugin available in RViz for manually planning and executing motions, an example of which can be seen in Figure 2.12.

![Figure 2.12: Screenshot captured by the author of an RViz instance using MoveIt! to plan on a PR2. The orange limbs are the desired state.](image)

**OpenRAVE**

Open Robotics Automation Virtual Environment (OpenRAVE) is a research-grade software environment which includes methods of modeling different robots and environment objects through a simulation engine to serve as a platform for experimenting with and validating motion planning algorithms. OpenRAVE also includes the IKFast module, which can generate very quick C++ code that analytically solves the IK for robotic manipulators described in the OpenRAVE-supported robot description format. C++ and Python API’s are available for interact-
ing with OpenRAVE, and it can be installed on Ubuntu and Windows with Visual Studio. There is decidedly a smaller community supporting OpenRAVE than its ROS counterpart, MoveIt!, but OpenRAVE can be used in ROS installations with some additional setup to allow access to IKFast and other tools. Figure 2.13 shows a screenshot of an OpenRAVE environment being used for planning on a PR2 humanoid.

Figure 2.13: Screenshot captured by the author of an OpenRAVE instance with a PR2 in a simulated environment completing a motion planning problem.
Chapter 3

System Design

This chapter is concerned with the first step of this project: defining the objectives and requirements of the system, deciding how to integrate the system in surgical procedures, and designing the software infrastructure to support the functions that need to be fulfilled. The system is designed to be generic and adaptable to different procedures, but focuses on specific implementation on the neurobot.

3.1 System Objectives

This project has the following objectives to accomplish in order to successfully fulfill the user need:

- Interfaces with existing medical platforms including the robot control system and 3D Slicer.
- Accurately assesses robot workspace considering the environment and entry point into the body.
- Generates collision-free trajectories from the robot’s starting configuration to the entry point in the body.
- Segments the patient into a 3D volume for collision avoidance.
• Provides a mechanism for assisting surgeons with selecting an optimal insertion point in the body to most readily reach the treatment location.

3.2 System Requirements

There are some additional considerations that can impact how the system can be designed. These have been organized into the below list of requirements the system must abide by.

The system shall...

• Operate on Windows, Mac, or Linux computers.
• Be easy to install and use by a non-technical user.
• Be packaged for portability and distribution.
• Be readily accessible to new developers in the lab.
• Communicate with the robot using OpenIGTLink.
• Utilize 3D Slicer as the user interaction interface.
• Communicate with 3D Slicer using OpenIGTLink.

3.3 Surgical Workflow

During procedures with the neurobot, there is an established workflow as discussed in section 2.3. The software system being developed in this project will redefine that workflow, and it is important to consider how the system can best modify that workflow to effectively make use of its new features. Figure 3.1 and Figure 3.2 functionally illustrate the workflow during a procedure making use of this system. A high-level overview of the workflow is as follows:

1. Register the robot in the scanner and send offsets for planning.
2. Send the patient head as a model to the software system.

3. Software system sends a workspace of reachable entry points to 3D Slicer.

4. Surgeon selects entry point. System computes the ultrasonic element workspace given that entry.

5. Target point is selected. System computes motion plan from current configuration to entry.

6. Operator examines plan and sends to robot controller.

7. Robot executes motion. Cannula and probe are attached to robot.

8. Robot inserts probe and ablation(s) are performed.

9. Cannula and probe are removed from patient. Robot is removed from scanner.

In some cases, not all of the steps above would be performed during procedures. A procedure could start at Step 4, assuming the robot registration had already been performed, and skip the entry workspace check if the operators were familiarized with the reachable entry workspace of the robot. In future iterations of the design, the entry point could also be automatically selected based on MR image analysis to locate the burr hole.

### 3.4 Software Design

Based on the objectives and requirements for the system at consideration, the following approach has been formulated. The motion planning is completed using MoveIt!, utilizing ROS as the system for managing the robot configuration and model. This choice is driven by the capability of OMPL, the streamlined collision
Figure 3.1: Diagram of the surgical system workflow when the hole is drilled before the patient is placed on the MRI bed. Page 1 showing procedure setup, workspace visualization, and trajectory generation.
Figure 3.2: Diagram of the surgical system workflow when the hole is drilled before the patient is placed on the MRI bed. Page 2 showing needle insertion procedure and extraction.
checking with the environment in MoveIt!, strong community support and knowledge of ROS and MoveIt!, and the ease of installing and maintaining the ROS software environment. Initially, development began with the hope of installing OMPL on a Windows machine since this operating system was most commonly used in the hospital and this could allow OMPL to be used directly from within a Slicer module built in C++. Unfortunately, source installation of OMPL was necessary, unsupported by the OMPL developers, and several other dependencies for running OMPL were equally as challenging to install. After some time was spent trying to overcome these barriers (some resources and instructions for the progress made on this endeavor can be found in section B.6), it was decided that even if a solution could be found, the complexity of the install process was too great and increased the risk of the software failing to be easy to install and configure on new systems. Thus it was decided that having a separate computer (or virtual machine) running Ubuntu was acceptable given the ease of setting up OMPL through ROS on that platform.

ROS and MoveIt! are installable on Ubuntu by a single scripted command and officially supported on Long Term Support (LTS) Ubuntu releases, increasing the potential lifespan of the software system before needing upgrades. This is further discussed in section 4.1. The motion planning code would then run on an Ubuntu computer linked to the robot and Slicer over OpenIGTLink. Alternatively, a Virtual Machine (VM) running Ubuntu can be hosted on a Windows machine to reduce the number of computers needed to run the neurobot during procedures.

Figure 3.3 shows the software modules and data interfaces between the ROS packages for workspace analysis and motion planning, the Slicer module for interacting with the workspace data, and the robot controller. A key component for managing the interface between ROS and Slicer is the ROS-OpenIGTLink-Bridge package, which connects to OpenIGTLink interfaces and exposes that data on the
ROS network via custom messages, as discussed in section 2.5. This allows a clear separation between OpenIGTLink-connected software and ROS connected software.

Figure 3.3: Diagram of the software architecture showing interfaces between ROS packages and 3D Slicer. The groupings in the green boxes on the ROS computer represent ROS packages where each blue box is a node; for the systems on the right, these represent different functions.
Chapter 4

System Implementation

This chapter covers the development setup of the system, implementation of key interfaces, and the packaging of the software for use in the field.

4.1 Development Configuration

The system was designed to run on a Ubuntu 16.04 computer with ROS Kinetic installed. Typically, ROS releases are tied to specific Ubuntu releases, and LTS versions are supported with security patches and bug fixes for five years after release with new releases every two years. Usually, ROS versions are generally API-compatible between sequential versions, but changes to the API structure may throw warnings during compilation that those API functions are being changed or removed in the next ROS version. Over a greater time period, ROS 1 will generally remain consistent across yearly revisions, but ROS 2.0 is not expected to maintain API compatibility. Installing 3D Slicer on this system was optional and trivial, but at least one computer had to have Slicer installed for conducting the procedure. The most recent stable Slicer program was used, which was 4.8.1. Beyond the simple one-line installation of ROS and MoveIt!, OpenIGTLink had to be downloaded and
built from source using a script. With those simple dependencies installed, software was ready to be built.

A standard catkin build tools workspace was created for building and sourcing the ROS packages for the software platform. A git repository contained the code for the project and could be cloned and built in the catkin workspace alongside other ROS packages. Nodes were written in C++ and contained in the package structure shown in Figure 3.3.

4.2 Bridging with OpenIGTLink

The ROS-OpenIGTLink-Bridge package simplified the task of interfacing with OpenIGTLink, yet it did not complete all of work that was necessary to complete the integration of the two systems. The ROS-OpenIGTLink-Bridge defined its own messages, so packages needing to subscribe to topics using those message types had to specifically include the bridge package to build. This ran contrary to the goal of having the ROS-based workspace analysis and motion planning be agnostic to OpenIGTLink and able to operate entirely within the ROS environment. Consequently, nodes were written which were parameterized to allow conversions from the OpenIGTLink messages to standard ROS messages, and the following sections document some of the key interfaces of the system. Additionally, these nodes also performed unit conversion from ROS (using meters) to Slicer (using millimeters). Potentially, it would be useful to push these conversion nodes to the bridge project as this would allow for better compatibility with ROS systems which already exist and do not need to be rewritten to use the bridge data types.

An advantage of the ROS communication paradigm was that two OpenIGTLink connections could be made, one to Slicer and one to the robot, yet both would be
exposed to the same topics on ROS. In other words, the ROS system did not care (and could not tell) how many different OpenIGTLink connections provided and sent the information to them. This provided additional flexibility in allowing the target and entry points to be defined by the robot control software as well as from an operator using Slicer.

4.2.1 Transferring Frames

The key transform necessary for the software system was the registration transform from the optical frame of the scanner to the z-frame (also the base frame) of the robot. This transform was sent through OpenIGTLink by the robot control software, and a node was written to take the IGTLink transform published by the ROS-openIGTLink-bridge package and publish it as a transform to the TF2 package. This transform would then be used to transform incoming entry and target points into the robot’s frame of reference for the purposes of workspace analysis and motion planning. Figure 4.1 shows the TF tree in ROS.

4.2.2 Transferring Offsets and Points

Additional nodes were written to convert the IGTLink points into geometry point stamped messages so that those points could be transformed into the coordinate frame in which subscribing nodes would be doing processing. Unfortunately, points sent over OpenIGTLink did not have transform data associated with them. In other words, all points (as is the case with Slicer) are specified in the optical scanner frame. This necessitated writing the nodes such that this assumption could be changed or specified as needed.

The offsets discussed in section 2.3 were transferred as OpenIGTLink string types since the bridge did not include the OpenIGTLink double or float types. Nodes
Figure 4.1: TF tree of the system showing the transforms defined in ROS.
then converted these strings into messages that were simply published as floats. Alternatively, and better following ROS practices, these offsets could be updated with a dynamic reconfigure server instead of being published on separate topics.

4.2.3 Sending Workspace Volume

The workspace was generated as a point cloud by the workspace analysis ROS package, so a node had to convert the point cloud into a format for OpenIGTLink. While there was a type for fiducial points, publishing over around 700 points caused Slicer to hang and run very slowly once all the points were loaded. Thus a different format was needed.

When viewing the MRI scan planes, Slicer overlays polydata meshes on whatever slice was being viewed. This was ideal for visualizing the workspace as it pertained to the procedure planning. Polydata could contain points, polygons, vertices, strips, and lines. Seeing only points representing the workspace reduced the understanding of the workspace boundary, and the points would also hinder the view of the interior of the workspace. However, meshes (consisting of polygons) were very easy to see from different slice perspectives and allowed a more intuitive interpretation for what was reachable within the boundary of the mesh.

A node was written which converted the point clouds to polydata meshes using the Point Cloud Library (PCL) and VTK data conversion tools. The ROS point cloud was converted to a PCL point cloud which allowed the point cloud to be processed using PCL’s built-in manipulation tools. Under the recommendation of a developer of the ROS-OpenIGTLink-Bridge, the marching cubes reconstruction algorithm was first implemented, but this attempted to create a mesh which best connected the surface normals of all the points, whether they were points on the edge of the workspace or if they were points within the workspace, which resulted in some
broken borders and internal meshes in the workspace. Once this was attempted, a concave hull approximation was used instead, which essentially tries to shrinkwrap a mesh around a point cloud. This achieved the desired result, and after some parameter tuning to ensure a smooth shape which closely held to the underlying structure of the point cloud, point clouds in ROS could be sent over OpenIGTLink to Slicer as seen in Figure 4.2.

![Figure 4.2: A point cloud of a cube in ROS (left) with the resultant polydata mesh object in Slicer (right).](image)

In slicer, the 3D displays under the model module had to be changed to show all surfaces, not just what Slicer thought were the front planes. When this viewing option was not selected, the mesh would appear jagged and incomplete, as seen in Figure 4.3. This was simply an issue with the way Slicer attempted to show through parts of the mesh, but toggling the option for that model to show the full surface rendered it properly as seen in Figure 4.2.
Figure 4.3: Polymesh object in Slicer with only the front faces being rendered, resulting in a choppy-looking mesh.

4.2.4 Receiving Robot State

For the purposes of motion planning, the robot’s starting configuration needed to be known. With the robot modeled in ROS, the joint state publisher would hold the state of the robot, but as will be discussed in subsection 5.1.3, a difference between the ROS model and the actual robot necessitated a conversion between the two joint spaces, which was done using ROS message types. With this transformation in place, the joint state had yet to be communicated between OpenIGTLink and ROS, and this proved to be somewhat of a challenge. The ND Array type in OpenIGTLink was ideal for holding this information, but this was not implemented in the ROS-OpenIGTLink-Bridge package. The bridge package repository was forked and the ND Array type was attempted to be added as a message and to the bridge. Unfortunately, the implementation of this was difficult on both the ROS side and the robot controller due to lack of documentation and samples utilizing the ndarray data standard. Instead, the joint state was sent as a string of comma separated values. Then a node was written that took properly named incoming strings
representing the robot’s state sent from the robot controller and published those as a ROS joint state message, which could then be transformed into the modeled state and sent to the robot joint state publisher so that the motion planning package and other packages using the TF library could use the updated robot state.

4.2.5 Sending Joint Trajectories

Trajectories in ROS were well standardized as messages with a list of joint names and a list of joint trajectory points which included position, velocity, accelerations, and effort of each of the active DOF and a duration time from the start of the trajectory at which that trajectory point was to be achieved. Of course, this format did not exist in OpenIGTLink, so another conversion had to be made. Again, the NDArray type would have been most suitable, but instead a string was used, in which a time stamp, joint positions, and joint velocities were sent separated by semicolons and commas. For instance, the joint positions were listed with commas while a semicolon separated both the duration time and the joint velocities. A node listened to trajectories being published from the motion planner, transformed the joint space as discussed in subsection 5.1.3, converted units, and sent those messages over OpenIGTLink. The robot controller would then parse this string to perform the desired trajectory.

4.3 Software Packaging

In order for the system to run on Windows and Mac systems, a VM image of Ubuntu 16.04 was created and maintained with ROS, MoveIt!, and OpenIGTLink installed. Oracle’s VirtualBox was the virtual machine monitor used, but others were compatible with the VM image format as well. The image was essentially
identical to the setup on a developer computer, but included scripts on the desktop for downloading and rebuilding code changes for the software, automatic login, and shortcuts for configuring OpenIGTLink parameters to adjust changes in the robot’s or Slicer’s IP address. The VM was run on the laptop computer which ran the Slicer interface during procedures, where the entry and target points would be specified while viewing the MRI captured of the patient.

4.4 Software Documentation

As with all software projects, documentation of code is key for enabling the system to continue to be useful and ensure that changes can be made easily. Documentation from a non-technical user perspective is written so that new operators can learn how to use the system without necessarily being trained personally. The user documentation can be found on the project repository. Code is commented throughout with Doxygen, but having dedicated technical documents for specific procedures to complete routine changes to the software is necessary. Appendix B contains the technical writings developed during the course of this project.
Chapter 5

Neurobot ROS Modeling

In order to utilize the tools within ROS for completing collision detection and conducting motion planning with MoveIt!, the robot had to be modeled in a ROS-compatible format, as discussed in section 2.4. This chapter discusses modeling the neurobot in ROS, creating its MoveIt! package, modeling environment objects, and interacting with the system and visualizing its outcomes.

5.1 Representation of Neurobot Kinematics

Generating this particular robot model in ROS required a bit more processing than a normal serial manipulator. The URDF does not allow closed loop linkages since the kinematic structure of the robot is eventually described using TF, which expresses coordinate frames as a tree structure. This complicated the development of an accurate model since the neurobot includes parallel linkages in the design, resulting in more joints than degrees of freedom. Further, as described in subsection 2.2.3 axial and vertical motion is attained through a pair of lead screws, where the differential between the two achieves vertical motion and synchronized actuation produces axial motion.
A workaround the limitation of the URDF to describe closed loop chains was to use mimic tags, which describe how one joint mimics or follows the pattern of another and removes the mimicking joint from the list of the robot’s active DOF. This allowed the parallelogram linkage mechanism, which was actually two prismatic joints, to be modeled as two decoupled joints with a prismatic joint for axial translation and a revolute joint for changes in height. The state of the other leg of the trapezoid simply mimicked that of the driving one, and the vertical height was attained simply by rotating the revolving joint. A similar mimicking pattern was used for the pitch axis, and the result of the final modeling of the neurobot is shown in Figure 5.1. In that figure, the revolute vertical translation joint is the bottom-right transform at an angle. The other axis in those legs of the trapezoid simply mimicked its motion with some sign flips to move properly.

Figure 5.1: The finished ROS model of the neurobot showing joint frames on the parallelogram structure. Height was changed by rotating the bottom right joint on the linkage.

The frame conventions on the robot did not follow the Denavit-Hartenberg (DH)
convention. The base link and tool frames were the only ones with intentionally defined frames so that end effector state would match that of the robot controller. Since many robots utilize this convention, it would be good to define the frames in this way, but with the transforms handled by the robot state publisher, the ambivalence of the numerical IK solver in MoveIt!, and the lack of a need to utilize intermediary transforms, this was eliminated in favor of faster model development in SolidWorks.

5.1.1 Exporting Robot Model to ROS

The robot was drafted in SolidWorks as an assembly of several parts and subassemblies. Fortunately, a tool (creatively) named “Solidworks to URDF Exporter” was available as a SolidWorks add-in that simplified the creation of URDF’s by starting from a SolidWorks assembly. General operation included opening up the add-in window and selecting parts or assemblies that represented a linkage on the robot, defining the frame and axis for each joint, selecting the joint type, and building a tree structure of joints which could be converted into a URDF format by exporting the URDF and properly generating STL files of the robot geometry. A screenshot of the tool in action can be seen in Figure 5.2.

While the add-in featured an auto-detection tool for identifying joint types, this did not work for the neurobot, and presumably would only work on simple serial mechanisms that were properly constrained to move only at the joints. Rather, at every joint, a coordinate frame had to be created which defined the zero location of the joint, and an axis was also defined through that coordinate frame around which (or along which) movement on that joint would be rendered. Of somewhat inconvenience, these frames and axis had to be created in the top-level assembly; when these were defined in subassemblies, the export process would improperly
assign frames and the STL files were not properly transformed to the right frames.

Once the joints were assigned and defined in the tree structure with the associated parts assigned to each linkage, the export process could begin. Joint limits, movement axis directions, efforts, and velocities were specified for each joint, rotational inertia and mass properties were inputted, colors were specified for linkages, and the ROS package was generated. However, some of these parameters, such as the joint limits, were not saved in the configuration file and had to be regenerated every time the URDF was exported. Once generated, the package was placed directly in the repository for the software system, but additional changes had to be made as explained in the next section.

5.1.2 Final URDF Cleanup

Some further tweaks were necessary to the URDF once it was generated by the SolidWorks to URDF add-in, which were outlined in the user documentation stored in the project repository. This included fixing author tags in the package, editing file
paths for some of the mesh resources, adding mimic tags to the appropriate joints, and ensuring joint limits and directions were accurate. Since some of the parameters for generating the URDF had to be reentered every time the package was generated, sometimes it was easier to edit the joint limit information after the fact. There was some trial and error involved with correcting motion directions, usually amounting to a sign flip to have the joint values increment and decrement as desired. Velocities also had to be specified for the joints so that generated trajectories would not try to move the joints in each time step faster than they were capable of moving. Once the joint limits, velocities, and directions were corrected, mimic tags were added to joints appropriately to achieve the desired kinematic chain. In the end, the robot had seven active degrees of freedom with seven additional joints that were either fixed or mimicking the active joints to produce the correct motion of neurobot, and the GUI for controlling the simulated robot is shown in Figure 5.3 with a rendering of the neurobot with the base plate in RViz.

![Figure 5.3: End result of neurobot modeling in ROS.](image)
5.1.3 Transforming Joint Space Representations

A single node was used to convert the joint state sent from the robot controller into the serial-chain model used within ROS and to convert the trajectories generated by MoveIt! into the joint state used by the robot controller itself. This was a relatively straightforward process, with most joints simply needing a sign flip and unit conversion (ROS uses meters, the robot controller and Slicer use mm). The parallelogram formed an isosceles trapezoid as shown in Figure 2.10, so given the joint states of both of the axial legs, the difference between them could be used to establish the base of the trapezoid and compute the acute angle quite readily. When going in the opposite direction for trajectories, the same sign flips and unit conversions were used, and a simple trigonometric calculation yielded the correct axial foot joint state. The node then republished the correct joint states on separate topics that were either converted into strings to send over OpenIGTLink for the case of trajectories or published as the joint state for the robot state publisher.

5.2 MoveIt! Neurobot Configuration

The MoveIt! Setup Assistant tool was utilized to generate the MoveIt! package for the neurobot. The URDF for the neurobot was selected, and the setup tool loaded the robot and stepped through the process of generating the SRDF that would be used for motion planning. A self-collision matrix was first generated, which tests for collisions between different links to see which links, if any, can have collision checking disabled between them to speed up collision checks. The highest resolution of checks was selected to ensure that there would not be any skipped checks that may cause self collisions.

Planning groups were configured next, which described chains of joints considered
as complete manipulators that motion planners would have to find trajectories for. Two planning groups were created. The first described the full arm, which went from the base frame to the needle tip, which included all seven active degrees of freedom. A second planning group included only the first five degrees of freedom to the needle driver base. This second planning group would simplify motion planning to the entry point, where the needle driver and rotation were not needed to be controlled.

Planning groups also needed an IK solver to be specified that would be instantiated upon a planning instance. The default IK solver utilized KDL, which was a numerical IK solution capable of solving serial chains and dealing with underconstrained systems. It properly supported the mimic joints within the planning group chain but would not always update the joints of the robot outside of the planning group which were mimicking joints within the planning group. This opened up the necessity for some additional workarounds in the workspace analysis and motion planning aspects of the project.

While the KDL IK solver worked properly, a custom solver implemented with IKFast, which was an analytical solver, could significantly speed the execution of the program. OpenRAVE includes IKFast as a plugin, so OpenRAVE was installed as an attempt to quickly generate an IKFast solution for the neurobot. This included converting the neurobot URDF to a COLLADA file using a command line tool and running the IKFast generation using that COLLADA file. Unfortunately, IKFast could not recognize the mimic joints on the robot to generate an analytical solution. As a faster IK solver was seen as an optimization, further work on this front was halted, but eventually including an analytical IK solution for the neurobot could noticeably speed up the workspace analysis and motion planning.

Robot poses could also be specified, which were joint configurations that would
be meaningful to drive the robot to in common practice. For the neurobot, the
default zero position sufficed. End effectors were also chosen, which were tied to
planning groups which controlled them. The needle and cannula were both selected
as end effectors for the full arm planning group and driver base planning group
respectively. Finally, no virtual joints were created for the robot, but this option
existed in the setup assistant. With all of the above options set, the MoveIt! package
for the neurobot was generated in the workspace.

5.3 Environmental Modeling

The environment for the neurobot consisted of the MRI scanner bore and the patient
head. The MRI bore was exported from a SolidWorks model of the scanner with the
bed removed to reduce the polygon count of the STL to speed up collision checking.
For testing purposes, the patient head was imported as an STL of a mannequine
approximation of a person’s head, but as discussed in Chapter 8 Slicer was used to
take subject images and generate an accurate collision model for planning around
the subject’s head.

These models were loaded into the MoveIt! planning scene over a planning
scene monitor to utilize as collision objects for the workspace analysis and motion
planning. Using the monitor allowed the collision objects to be added and removed
from the world at will and to be manually modified within the MoveIt! RViz plugin.
An RViz screenshot the typical planning scene environment is shown in Figure 5.4.

5.4 System Interaction and Visualization

MoveIt! included a plugin for RViz that allowed motion plans to be generated
by moving an interactive marker representing the end effector state around the
environment while the IK was computed for the robot in real time, showing the actual state of the robot necessary to reach the desired point. A motion plan could then be generated from the current state of the robot to the target state, and if the planning was successful, a preview of the trajectory was played on a loop on the robot model. The interface also allowed collision objects to be loaded from STL’s or added and removed once they were published to the planning scene monitor, as discussed in the previous section. Figure 5.5 shows the MoveIt! plugin being used in RViz.

Figure 5.4: An image from RVIZ showing the MRI Bore model and human head model loaded with the robot utilizing the MoveIt! API from within a C++ node.
Figure 5.5: Screenshot of RViz with the MoveIt! plugin enabled for the neurobot. A trajectory has just been found to the state with the brightly-colored model of the neurobot.
Chapter 6

Workspace Examination

In this chapter, the work of conducting workspace analysis of the neurobot is discussed. The different components of this objective were visualizing the full entry workspace of the robot, visualizing the workspace of the ultrasonic element, adding collision checks with the environment, and creating the workspace given an entry point.

6.1 Entry Point Workspace Visualization

The general approach for the workspace visualization was to iterate through the joint space of the robot and use the FK solution to store the end effector position in a list which was then published as a point cloud. In this step of the process, the workspace of valid entry points was generated. The entry point was based on the inputted offsets, discussed in section 2.3, and the first five DOF of the robot were sufficient to resolve this. A discrete linear space was created between the joint limits on each of the five active joints, and this space was then iterated through to fully explore the robot’s available workspace.

Self-collisions between links had to be checked, and this proved to require some
careful use and updating of the planning scene that was storing the robot state. The general process was to load the kinematic model of the robot and the planning groups defined using MoveIt!, and use the planning scene to update a virtual robot with new joint values, test for self collisions, and based on the result, add the reachable end effector point to the list of reachable workspace points.

Issues arose with simply using the active planning group for the workspace analysis; setting the planning group joint limits did not update the full kinematic state of the robot. In most applications, this is desirable, but for the neurobot, the changes in the serial chain needed to be mimicked by the other joints in the robot to check that the state was valid without collisions. This was resolved by algorithmically detecting the active degrees of freedom of the robot from the entire kinematic model, storing their names in order, and then rather than updating the planning group joint values, the active joint values for the entire robot were updated and the Robot State API would update any mimicking joints as necessary in the entire robot’s kinematic model. Better support for mimic joints in MoveIt!, especially with the planning groups, would have simplified much of the development of this node, but even with these limitations, the entry workspace of the robot successfully generated as a point cloud and wrapped with a mesh that was viewable in Slicer, as shown in Figure 6.1.

6.2 Element Workspace Visualization

The ultrasonic element on the probe was essentially the true end effector of the robot. For determining the element workspace of the robot, the same process described for the entry workspace was used, but including the additional DOF provided by the needle driver. The needle rotation was ignored since it would not move the element
workspace, but it is possible the geometry of the cooling tubes could allow needle rotations to have a slight impact on the workspace. This check should be added in the future, since it would be desirable to have the ability to spin the needle for conducting treatments using a non-symmetric ultrasonic probe.

Since a fixed-length needle based on the SolidWorks model of the robot was used and the actual offsets could change to accommodate different probes, this was compensated for while iterating the needle DOF through its range of motion. In the future, this could instead be handled by having a set of different probe models which could be placed on the robot as end-effectors, making the process more generic and adaptable. The cannula length could also change, so a check was made to ensure that the ultrasonic element would be exposed beyond the cannula to be considered a valid position. If the ultrasonic element of the probe was still within the cannula, the ablation and possibly the probe could fail. However, it was also possible to
adjust the cannula manually to achieve the desired size, so this was not particularly
critical to operations. Figure 6.2 shows the generated workspace as shown in Slicer.

Figure 6.2: Screenshot of Slicer with the ultrasonic probe element workspace mesh
overlaid on a pig head.

6.3 Collision-Aware Workspace Examination

Collision checking was added somewhat easily to the process of generating the
workspace. Collisions with the environment could be checked by publishing col-
lision objects to the planning scene. For the full workspace of the robot, collision
checks with the MRI scanner bore were necessary. The STL model of the bore was
loaded by a published file path and published to the planning scene monitor, such
that the internal scene in the node included the scanner and the published planning
scene in ROS also had access. Figure 6.3 shows the view in RViz with the scan-
ner bore loaded and the entry workspace generated with collision checking with the
bore.
The above two features were then modified to accept a collision checking function to allow flexible collision checking with either just self-collision checks or environmental checks. To speed up full collision checks, self-collisions were first checked since these were sped up by the self-collision matrix created during MoveIt! setup. If a self collision happened, then environment collision checks were skipped and the next configuration was tested. However, if the self-collision check passed, then collision checks with the environment were performed. Figure 6.4 shows the entry and element workspaces in Slicer with the collisions with the scanner bore considered.

6.4 Workspace Evaluation from Entry Point

The final workspace to evaluate was that of the element given an entry point for the robot. This analysis was a bit different from the previous two. Essentially, the IK was solved for the first five DOF of the robot to the entry point (with the

Figure 6.3: Screenshot of RViz with the scanner bore loaded in the planning scene and the entry workspace computed as a point cloud.
given robot to entry offset from the entry point to the robot’s cannula base) by iterating through valid rotations at that point, and when a configuration was found that solved the IK, collisions were checked and the needle joint was swept through its joint space. In order to speed up collision checking, when an IK solution was found, the needle was kept at the maximum depth so that if collisions occurred at this configuration, they would definitely still occur when the needle was retracted and those tests could be skipped.

While it was desired that the approach would be fairly generic and explore all 3 DOF of rotation, the IK solver was too slow for exploring spaces that were known to be invalid given the kinematics of the neurobot. This was due in part to the numerical nature of the IK solver - an IK solution really was a searching process, so failure to find a solution occurred at a user-specified timeout. This was typically set to around 10ms from experimentation, since a solution could be found for valid
configurations within that time on a laptop with an i7-4702MQ processor. Using an analytical solution would allow the full rotational space to be explored to make the solution more generic, but this solution approach worked well and could generate an accurate workspace in about 40 seconds for an entry point, the result of which can be seen in Figure 6.5.

![Figure 6.5: Screenshot of Slicer with the ultrasonic probe element workspace given an entry point.](image)

### 6.5 Selecting the Target Point

With the workspace of the probe element shown in Slicer given the entry point in the patient head, the user could then select the target point. Guided by this workspace, it was clear whether the target point was reachable by the robot and how far the robot could treat within the patient’s head with the current burr hole. Figure 6.6 shows the target point selected within the allowable workspace of the ultrasonic element on the probe.
The target point was then sent back over OpenIGTLink to ROS. This would allow ROS to resolve the entry point into an entry pose upon which a motion planning solution could be found. Figure 6.7 shows an image from RViz of the target and entry point as standard ROS point stamped messages.
Figure 6.7: Graphic from RViz showing the entry point (blue) and the target point (pink) in the point cloud (orange) representing the reachable element workspace given the entry point.
Chapter 7

Motion Planning

For this part of the system development, the MoveIt! API was used within a ROS node to solve the motion planning problem of moving the robot to the entry point without colliding into the patient head to achieve a pose which allowed the target point to be reached by the ultrasonic element on the probe. Once the trajectory was generated, it could be transformed into the joint space of the actual robot and sent through OpenIGTLink the robot controller for trajectory execution. The following sections in this chapter discuss the development of these capabilities.

7.1 Motion Planning with MoveIt!

7.1.1 Resolving Entry Pose

In order for a motion plan to be generated to the entry point that reaches the target point once the ultrasonic probe is added and inserted, the entry point must be resolved into an entry pose in order for the IK to solve for a configuration and the motion planner to have a goal configuration to reach. This was achieved by using both the entry point and the target point to resolve an orientation that could be
combined with the entry point to get a full 6 DOF pose. Of course, two 3D points could only resolve two degrees of rotational information, but the last degree (the rotation about the needle) was not needed for the five joints that moved the robot to the entry point and could be determined by the kinematic constraints of the five DOF system in that case.

The pose was resolved by computing the angles between the nominal axii of the neurobot’s tool tip frame with joint values set to zero and applying a series of rotations, with the position of the frame being located coincident with the entry point. A rotation about the roll and yaw oriented the frame to align with the tool tip frame. The rotation about x was then calculated using atan with the y and z changes in position. In the next computation, since the frame was locally rotated, the calculation of the y orientation had to take into account the changed dz value due to the rotated X axis. Finally, all these separate rotations were post-multiplied together to obtain the appropriate pose orientation based on the entry and target points. Part of the code responsible for this calculation can be found below.

```cpp
// Quaternion rotation_optical_to_tool_tip;
rotation_optical_to_tool_tip.setRPY(-M_PI_2, 0, M_PI);

double dx = _target_point.point.x - _entry_point.point.x;
double dy = _target_point.point.y - _entry_point.point.y;
double dz = _target_point.point.z - _entry_point.point.z;

// when in the same plane, keep at 0
if (dy == 0 && dz == 0)
    rot_x.setRPY(0, 0, 0);
else
    rot_x.setRPY(std::atan2(-dy, -dz) - M_PI_2, 0, 0);

// Quaternion rotation_optical_to_tool_tip;
rotation_optical_to_tool_tip.setRPY(-M_PI_2, 0, M_PI);

double dx = _target_point.point.x - _entry_point.point.x;
double dy = _target_point.point.y - _entry_point.point.y;
double dz = _target_point.point.z - _entry_point.point.z;

// when in the same plane, keep at 0
if (dy == 0 && dz == 0)
    rot_x.setRPY(0, 0, 0);
else
    rot_x.setRPY(std::atan2(-dy, -dz) - M_PI_2, 0, 0);
```

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else
    rot_y.setRPY(0, -std::atan2(std::sqrt(std::pow(dy, 2) + std::pow(dz, 2)), -dx) + M_PI/2, 0);

msg_out.pose.orientation = tf2::toMsg(*rotation_optical_to_tool_tip*rot_x*rot_y);

7.1.2 Updating Robot State and Environment

Motion planning to the entry point also required the current state of the robot, which was attained as described in subsection 4.2.4. The node simply subscribed to the joint state publisher for the robot and updated the planning scene’s robot state with the latest information before starting the motion plan. The patient head was also added as a collision object as provided by the MRI segmentation into a volume as detailed in chapter 8, or alternatively motion planning could be conducted on a generic head shape if the patient model was not available. Figure 7.1 shows the environment setup with the entry point, target point, and patient head mesh for the start of the motion planning process.

7.1.3 Checking Collisions and Assumptions

As review, the robot remained outside of the MRI bore and did not have the ultrasonic probe and cannula installed when the robot was sent to the entry point during procedures. This allowed the robot to move without the additional constraints of the probe and cannula potentially hitting the patient head and the rest of the robot striking the MRI bore. However, once the robot moved to the entry point, the needle driver was added and inserted to the required depth, and the patient bed was slid into the MRI tube for monitoring of the ablation.

Before beginning the process of motion planning, it was checked to see if an IK
solution could be found to bring the robot to the required entry pose and if that configuration (with the needle inserted to achieve the target point) avoided collisions with the MRI bore. It could be assumed that the target point was chosen within the allowable workspace given the entry point, but a warning would be thrown if this was a case, requiring an override by the user to continue to generate a solution with the knowledge that the final configuration of the robot would be in a collision state.

With it now known that the entry pose and target point were reachable without collision, the next step of the motion planning process could be completed. Since the robot did not have the cannula and ultrasonic probe installed during movements to the entry point, these links could be omitted from collision checking. This was accomplished by modifying the Allowable Collision Matrix (ACM) within the MoveIt! planning scene to ignore collisions between the cannula link and the needle/probe link. Collisions between mimicking joints outside of the planning group were also
skipped since MoveIt! did not properly update those based on motion planning using the planning group joints.

### 7.1.4 Generating a Motion Plan

With the setup work completed, searching for a motion plan was a relatively straightforward task. This was as simple as calling a function with the MoveIt! API to generate a plan to the entry pose joint configuration. The motion planner then searched for a valid path to the goal, and if the planner timed out, this was considered a failure and the operators would have to manually position the robot. With collisions considered between the robot and patient, trajectories were typically found within 20 seconds, so a 30 second timeout was used for the motion planning problem. It could also indicate that there was no way to move the robot to the entry point without violating its kinematic constraints while avoiding collisions with the patient head, so perhaps a safer response would be to reposition the patient, perform new scans, and begin the process again as shown in the workflow designed in section 3.3. If the plan was successful, the trajectory was published, and the set of end effector poses was published as well.

As discussed in section 2.6, LBKPIECE was supported by MoveIt! and produced faster trajectories than other motion planning algorithms. Unfortunately, the MoveIt! wrapper did not expose the ability to manipulate or change cost functions. Thus the default cost function was utilized, which optimized the path distance. Below are the available configuration options and the values chosen for the experiments. Figure 7.2 shows a trajectory being executed within RVIZ given this configuration to move the robot from the current state to the entry pose without striking the patient head, and Table 7.1 documents the planner parameters used for the project.
Table 7.1: Motion planner settings for LBKPIECE.

<table>
<thead>
<tr>
<th>Parameter Name</th>
<th>Value</th>
</tr>
</thead>
<tbody>
<tr>
<td>Border Fraction</td>
<td>0.9</td>
</tr>
<tr>
<td>Failed Expansion Goal Factor</td>
<td>0.5</td>
</tr>
<tr>
<td>Goal Bias</td>
<td>0.9</td>
</tr>
<tr>
<td>Min Valid Path Fraction</td>
<td>0.5</td>
</tr>
<tr>
<td>Border Fraction</td>
<td>0.9</td>
</tr>
<tr>
<td>Min Valid Path Function</td>
<td>0.5</td>
</tr>
<tr>
<td>Range</td>
<td>0.0</td>
</tr>
<tr>
<td>Type</td>
<td>Geometric KPIECE</td>
</tr>
</tbody>
</table>

Figure 7.2: RViz screenshot of the trajectory solution showing the starting and ending state of the robot. An looped animation plays showing the robot following the trajectory.
Chapter 8

MRI Segmentation

In order to have an accurate collision model of the patient head for motion planning, a 3D mesh of the patient head was extracted from available MRI scans and sent over OpenIGTLink to the ROS network for further processing and loading into the planning scene monitor. The following two sections describe this process in more detail.

8.1 Slicer Processing

The process of generating a solid mesh from a series of scans within Slicer was generally straightforward with the Segment Editor, which was well documented on both the Slicer Wiki and through a tutorial written for completing segmentations for 3D printing body parts. Slicer and this tool in particular have been noted at providing excellent algorithms for processing medical images, for instance, segmenting brain tumors, in a very accessible way for researchers. A segment was added using the thresholding tool with the lower bound set to a number which just eliminated the background from the image and the upper bound set to the maximum. Figure 8.1 shows the scan planes with the threshold mask and the rendered
3D volume.

Figure 8.1: Screenshot of Slicer showing the rendered mesh of the patient head based on mask thresholding.

Once this mask was applied, the islands tool was used to remove all but the largest islands from the selection, which eliminated noise in the image and any items which were visible in the scan but not connected to the largest object, which was the patient’s head. The rendered 3D volume is shown in Figure 8.2.

To remove noise from the model and smooth the surface finish, two rounds of filters were used on the segment of the patient head. The first was a closing filter, which connected the mesh surface over holes under 3mm in diameter. This had the effect of also closing some internal geometry that added unnecessary polygon counts to the mesh, as seen in Figure 8.3.

Next, a normal filter with a setting of 4mm was used to generally smooth the model’s surface. This would remove some of the detailed texture of the patient’s head, but kept the geometric shape suitably. A rendering of the volume after the
Figure 8.2: Screenshot of Slicer showing the rendered mesh of the patient head once islands were removed from the mask.

Figure 8.3: Screenshot of Slicer showing the rendered mesh of the patient head after a closing filter was applied.
filtering is shown in Figure 8.4. The segmentation was then exported as a model within Slicer through the segmentations module, which was the same format in which the workspace polydata objects were stored in. This model was then named patient head and sent over OpenIGTLink, which could take a couple dozen seconds to serialize, transmit, and deserialize on the receiving end due to the size of the polydata.

Figure 8.4: Screenshot of Slicer showing the rendered mesh of the patient head after normal filtering.

8.2 ROS Processing

Since collision checking is a time-consuming process, any method of simplifying the geometry of the collision objects can help reduce the processing time for motion planning. The fidelity of the meshes produced by the segmentation in Slicer was much higher than would be required for motion planning, so the triangle count of the polydata objects could be reduced to speed up collision checks. Ideally, collision
meshes should simply be a solid shell as only the outer surface is necessary for checking against collisions. To quickly remove the internal geometry of the mesh, the same concave hull process for wrapping point clouds with meshes (as discussed in subsection 4.2.3) could be used. In any case, the rest of the processing was moved to ROS, so the volume from the segmentation completed in Slicer was then sent over OpenIGTLink.

A different ROS node was created with the same basic format as the point cloud converter, but in this case, it subscribed to the OpenIGTLink polydata topic and published to the planning scene monitor. Incoming polydata messages were converted to VTK types, then a decimation process was used to reduce the triangle count by 95%. Figure 8.5 shows what that decimated mesh looked like if it was viewed in Slicer.

![Figure 8.5: Screenshot of Slicer showing the STL of the decimated patient head after processing from the ROS node.](image)

To ensure that the decimated mesh (reducing the STL file size from 90.4 MB
to 16.5MB) would not significantly impact the accuracy of collision checks, the original 3D segmented model was compared with the decimated version of the model. Figure 8.6 shows the decimated mesh overlaid with the original mesh. Examining the scan planes, it can be seen that the two stay within about two pixel widths of each other, which shows how tightly the decimated mesh still matches the original shape. MoveIt! also adds an extra cushion around environmental objects during collision checks to help ensure that uncertainties in sensing do not cause unforeseen collisions with the environment.

Figure 8.6: Screenshot from Slicer showing the STL of the decimated patient head in blue with the original mesh in red shown in the 3D preview and overlaid on the original images.

The decimated polydata object’s points were converted to a PCL point cloud
so the same code could be used for performing the concave hull approximation
developed in subsection 4.2.3 and the resulting VTK object was saved to disk so
that it could be viewed in Slicer for reference purposes, as shown in Figure 8.7.
The VTK object was then saved as an STL, and the STL path was published to
the Workspace Visualizer so it could load the mesh resource as a collision object at
the appropriate time and publish that to the planning scene. Figure 8.8 shows the
final outcome of this process with the patient head model properly loaded into the
planning scene, attached to the base link of the robot to allow collision checks to be
skipped for that link and to prevent false positives.

Figure 8.7: The patient head shown in Slicer as a result of the concave hull operation.

8.2.1 Segmentation of Human Head

Slicer includes several freely available data sets of medical images for conducting
research, and one of those data sets is of MR images of a human head. To verify
that this procedure worked well for a human head, the same steps described above
were repeated for the human head model. The results showed the process (even using the same threshold values) produced a good model of the patient head and indicated the process could be replicated for human subjects. Figure 8.9 shows the human head segmentation in both Slicer and RViz.
(a) Slicer view of segmented model of human head.
(b) RViz view of human head as attached collision model in the planning scene.

Figure 8.9: Results of segmenting a human head from MR images.
Chapter 9

Experiments

In order to validate that the system meets the user requirements and to better understand its performance and capabilities, several tests were completed. A series of metrics were collected for the IK solver and motion planning performance of the system. Then, tests were conducted using the actual robot to evaluate the modeling of the neurobot and workspace analysis.

9.1 Benchmarking

Several experiments were conducted in order to gain a better understanding of the execution time of the IK solver and the motion planning implementation. Tests were configured by determining a viable end effector needle pose by uniform random sampling from the configuration space of the robot and using the forward kinematics to compute the end effector position. This position was then fed back into the IK solver and the time it took the IK solver to compute the solution was measured. Tests were conducted on a computer with a Ryzen 7 1700 processor operating at 3.00 GHz with 16GB of DDR4 RAM.
9.1.1 IK Solver

In the first experiment, the parameters of the IK solver, namely the number of allowed solver attempts and the allowed search time, were set at default values of 5 attempts and 0.01 seconds. The end goals were selected by randomly picking a valid joint space, computing the forward kinematics for that state, and then supplying the IK solver with that end effector information to solve for. The start state was kept constant at the robot’s zero position. Figure 9.1 shows the execution time plotted over 200 iterations of the solver. Interestingly, the solving time was somewhat dependent on the order of the queries: the initial query takes the longest and the number of long queries decreases over time. This seemed to indicate possibly some caching optimizations or time used to compile Python code at run time.

![Plot of IK Solver Time Over 200 Attempts](image)

Figure 9.1: Plot of inverse kinematics solver execution time to find viable joint configuration in collision free environment given a random reachable target over 200 consecutive trials. Average 0.004546s. 5 allowed attempts with 0.01s to find a solution.

In the next experiment, the number of attempts was fixed at 1 and the time allowed for finding solutions was varied from 0.00001s to 0.00991s. Figure 9.2 shows
the results of the experiment. While a slight minimum appeared around 0.003s, there was not a strong trend, other than in longer search times, the execution time generally increased. Figure 9.3 shows the same data but as a histogram plot. This seemed to indicate that providing longer search times did not have an impact on the execution time; solutions generally took around the same amount of time to converge.

![IK Execution Time over Varied Search Time](image)

Figure 9.2: Plot of inverse kinematics solver execution time over allotted search time to find viable joint configuration in collision free environment given a random reachable target. 200 trials.

In the final experiment with the IK solver, the number of attempts allotted to the solver was varied from 1 to 19, with 200 tests being conducted for each timing point. Figure 9.4 shows the results of the test. Perhaps somewhat unsurprisingly, with only 1 attempt at solving the IK, the average execution time was highest since failure to find a solution would maximize the timeout allowed on the search without being able to spawn another search. Beyond 2 allowed attempts, there was no strong trend, indicating that the choice did not matter significantly.
Figure 9.3: Histogram plot of inverse kinematics solver execution time to find viable joint configuration in collision free environment given a random reachable target. 200 trials. Average 0.00207957s

Figure 9.4: Plot of inverse kinematics solver execution time over number of allowed attempts to find viable joint configuration in collision free environment given a random reachable target. 200 trials averaged for each number of allowed attempts. Average 0.001707s
9.1.2 Motion Planner

The motion planner was examined in the next set of experiments. Similarly to the tests for the IK solver, the same random valid configuration generation was used to set a desired pose to the needle planning group, and collisions with the environment were not considered. Further tests should include sampling collision free space for starting and ending goals to compare the planning time with environmental collisions considered.

Figure 9.5 indicates that most planning sessions lasted from 0.1s to 0.225s, with a distribution tail towards longer execution times. However, no plans took longer than 0.5 second to generate, even given a limit of 5 seconds. This showed that motion plans without environment collision models to consider were fairly quick, but performance with environment collision checks could be significantly longer.

![Histogram of Motion Planning Execution Time](image)

Figure 9.5: Histogram plot of motion planner execution time to find path in collision free environment between 2 random configurations. 200 trials. Average 0.209805s

The path smoothing operation was also examined in some detail. Figure 9.6 shows the execution time of the path smoothing algorithm, which indicated a rela-
tively narrow distribution with a tail end towards longer execution times, similar to
the distribution of planner times. This was expected as more complex trajectories
take longer to generate and will also have a longer path, increasing the time it takes
to adequately smooth it. In general, the path smoothing was a very small part of
the total path planning operation.

Figure 9.6: Histogram plot of motion planner path smoothing execution time. 40
trials. Average 0.0000073s

9.2 Joint State Evaluation

To ensure the robot was correctly modeled in ROS, the neurobot was moved to sev-
eral different configurations and checked against the model in ROS visually. Calipers
were used to measure datums between the parts of the robot and verified to be within
0.1mm of the transforms maintained by the robot state in ROS. Figure 9.7 shows a
comparison noting the proper modeling and interpretation of the joint state of the
actual robot within the ROS environment. Appendix A also includes a video of the
robot joint state being updated in real-time.

(a) View in RViz of the robot state in ROS. (b) Photograph of the robot’s actual configuration in test.

Figure 9.7: Test confirming correct modeling of neurobot.

9.3 System Workspace Validation

To test the workspace validation, motion capture was used on the neurobot as it completed movements mimicking those generated during the workspace analysis on the modeled neurobot. The end goal was to directly compare the shape of the workspace obtained by the real hardware with that achieved by the software system. Fiducials were used on the robot’s base to establish the base frame of the robot and a valid transform to directly relate the modeled robot frames with the simulated robot frames. Fiducials were placed on the robot where the probe would normally be inserted, as shown in Figure 9.8. The offset from the needle tip fiducial was measured from the entry point location on the robot’s needle driver base so that the same offset software could be used to define the robot to entry offset and directly compare the tip location as computed by the software.

The robot was manually moved to joint positions roughly outlining the config-
Figure 9.8: Picture of the experimental setup of the neurobot for workspace validation.
uration space. Care was taken to ensure that the robot did not self-collide since
the robot controller could not check for self collisions before executing a motion.
The fiducial position was logged into a CSV and a node was written which parsed
the text file into a PCL point cloud and published the point cloud with ROS. The
mocap frame was also saved in the TF tree and the point cloud was transformed
into the optical frame. Figure 9.9 shows the computed point clouds in ROS with
the mocap frame defined where the fiducials were placed on the base plate.

![Figure 9.9: RViz view Entry workspace point clouds compared between the real
neurobot (green) and that computed by the workspace visualization system (purple).]

This point cloud was then processed through the same mesh generation that the
software system used for the simulated workspace. Figure 9.10 shows an overlay
of the real and simulated results of the entry workspace. The workspaces had very
similar boundaries, with the actual robot having a generally smaller workspace in
efforts to be more conservative and avoid breaking the robot by hitting joint limits.
Notably, the lateral axis had a larger range of motion than the modeled robot,
indicating some minor modeling inaccuracies.

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Figure 9.10: Slicer view of entry workspace meshes compared between the real neurobot (green) and that computed by the workspace visualization system (purple).
9.4 Trajectory Validation

The robot controller trajectory follower was not implemented in time to conduct tests using the software system and neurobot. However, the trajectories were successfully received and parsed by the system into feasible joint values, and the first waypoints in the trajectory were executed by the robot before it stopped. The following experiment was proposed for evaluating the accuracy of the trajectory following.

To validate the trajectory generation and execution, a similar setup was used as in the previous section. For this test, a simple cube was placed in the actual robot’s space and carefully measured from the base frame. A geometric cube of the same size was added to the planning scene in the same position within ROS. The test environment remained the same as the workspace analysis experiment. In the software system, a motion plan was requested to a goal in the robot’s workspace. Once the motion plan was computed and the trajectory of the entry point was saved, the real robot executed the same trajectory with mocap recording the marker position. The path of the actual robot was then saved as a point cloud and compared to the desired trajectory as determined by the motion planning software.
Chapter 10

Conclusions

This project demonstrated the successful implementation of a system which yielded practical improvements to the neurobot capabilities through an extensible framework which increased perception of the robot’s workspace in the confines of an MRI scanner and ability to generate collision free motion plans without striking the patient. The system achieved its objectives, namely:

- Modeled the neurobot using ROS tools
- Generated collision-aware workspace of the entry and target points
- Computed collision-aware treatment workspace through an entry point
- Computed collision-free trajectories to the entry point
- Interfaced with Slicer over OpenIGTLink
- Interfaced with the robot using OpenIGTLink
- Provided workflow for creating 3D model of patient from MRI for collision avoidance

The system design proved suitable for achieving the objectives at hand and maintaining a good level of modularity and maintainability. The software system
was implemented successfully with special attention paid to interfaces between environments. While software remained research-grade, it succeeded as a prototype for a proof-of-concept method for improving procedure workflow by visualizing the workspace of the robot and conducting motion planning using tools already available to a developer, and these worked sufficiently to meet the need. There would be considerable work necessary to bring this system into production, as will be discussed in chapter 11. The software distribution through a VM met requirements, but perhaps a better solution would be to use a system such as Docker to avoid the less involved process of installing and maintaining a VM.

Modeling the robot in ROS was possible by decoupling the motions on the differential screw drive. While the URDF format did not offer a mechanism for modeling parallel robots directly, the system functioned adequately but with some added complexity of having to manage the entire robot state when planning for a single chain of joints. KDL does allow for solving parallel kinematic chains, but the ROS TF package cannot truly describe the kinematic state of such systems. There could be benefit to directly interfacing with KDL, yet the benefits of utilizing higher level features found in MoveIt! tended to offset drawbacks for this use case.

Workspace analysis proved very valuable in granting the ability to visualize the entry workspace of the robot directly in Slicer overlaid on surgical images. The examination included consideration of the scanner bore as a collision object, which ensured that the outputted workspace would be accurate and prevent the possibility of finding a robot configuration that reached the target but collided with the bore. Providing the workspace given the entry point was very helpful in formulating a method for determining how flexible a chosen entry point was in allowing a greater number of treatment sites to be targeted and helping the surgeon realize the importance of locating the burr hole in a way to best utilize the ability of the
neurobot.

The motion planning components functioned desirably in generating a trajectory which moved the robot to the entry point without colliding with the patient head (though full implementation on the robot controller side was not completed in time for testing), removing the cumbersome step in the procedure of moving the robot joint by joint to the correct configuration in order to avoid collisions. MoveIt! proved a reasonable platform for developing a ‘turn-key’ motion planning system, yet it did mask several more advanced features of OMPL. For instance, cost functions could not be modified; they were locked to OMPL defaults by the MoveIt! interface. To attain cost functionality adjustment, OMPL must be utilized directly without the MoveIt! wrapper. This would significantly increase development time, but could prove advantageous in other ways, allowing better tuned motion planning algorithms to be used to speed up planning time and potentially develop better paths. Yet with some of the shortcomings with MoveIt!, its ease of use and installation met the requirements set forward by the project and performed as necessary.

Working from MRI scans of the patient’s head, the segmentation and extraction of the volume for use in collision detection proved adequate for accurately producing motion plans within a tightly constrained environment. The hand-off between Slicer and ROS yielded the best case of software reuse and utilized each systems’ capabilities to the fullest.

Results and testing show that the system performed the motion planning sufficiently quickly to be useful on the actual robot hardware. Trajectory generation to a valid pose consumed an average of 0.21 seconds with no collision objects, and as the motion of the robot to an entry point on the skull will generally be collision free, this approach promised to be significantly faster than manually planning and executing trajectories by driving the robot joint by joint in the operating room. Fur-
ther experimentation showed that the actual neurobot produced a similar practical workspace, validating the modeling of the robot and the accuracy of the workspace generation using the tools supplied by ROS and MoveIt!.

In the process of developing these capabilities for the system, there were some libraries and tools that could use some additional development that, if fixed, would make consequent developing on them easier and reduce the amount of work for bringing new systems online. To summarize, these were:

- **ROS-OpenIGTLink-Bridge** - including conversion nodes to only publish standard ROS messages outside the bridge package.

- **SolidWorks to URDF** - saving joint limit configurations, improvements to automatic detection of joint type.

- **MoveIt! Planning Groups** - better support for mimic joints.
Chapter 11

Future Work

While the system was successful in meeting its objectives and requirements, there were some improvements that could be made to further improve its capability and utility. Natural extensions to this project could benefit the greater community as a whole with sufficient resources and time. To eventually move this research-grade system into the production environment where the software could eventually be used clinically with humans, the software stack would be subjected to significant scrutiny to prove it was robust and safe. Several different improvements and extensions to this work are discussed below, in addition to finishing the proposed evaluation tests of the motion planner.

Improved Software Packaging

As mentioned previously, utilizing the software within a VM was not an ideal design choice, but given the time constraints, requirements, and delicate nature of software environments and library installations, this was the best balance of performance and cost. With additional time, it would be ideal to contain the software with Docker or in some way integrate the system better to reduce this extra complexity.
It could be possible to make a hardware change to solve this problem too. A high-performance traditional x86 processor in a small form factor (such as an Intel NUC) could be added to the neurobot control box or as the means of providing network communication over the fiber cable and hosting some other higher level functions. This would also yield the benefit of exposing the neurobot to the ROS environment, and perhaps tighter integration of the neurobot with ROS could spur some additional development or integration with other hardware platforms.

**Modeling Improvements**

In modeling the neurobot, there were some changes that could be made to the SolidWorks to URDF add-in that could enhance its functionality, such as saving the joint configuration data and improving the joint type auto detection. While the workflow of updating a robot model and migrating those changes into URDF was not too burdensome, addressing these problems would greatly speed the process of completing those updates, as had to be done in this project. Eventually, building a simulation of the neurobot in ROS’ simulation environment, Gazebo, could help fine-tune the neurobot controls and allow the motion planning solution to be more vested and allow proper unit testing, especially since it is possible in Gazebo to model parallelogram linkages. It would be interesting to develop a simulation of an MRI scanner as well and attempt to model the entire workflow with imaging.

The joint velocities of the neurobot were not computed for the control system, so the ROS model did not have these values available. There was not time to perform analysis and testing to determine the max velocities for each joint, which would influence the generated trajectory. Since the controller was designed to just move to each joint state in the trajectory and wait for the other joints to reach the same state, this was not a problem. In the future however, the velocity should be used in
the trajectory following.

The needle modeling could be improved by making several different end effector URDF’s that could be added or removed from the robot depending on the procedure configuration. Such a configuration change could be enabled through using the macro XML extensions in the URDF. This would eliminate some of the less conventional methods in the workspace analysis of artificially lengthening or shortening the probe tip by manipulating the treatment offset. Additionally, this approach would be more general and extensible if different tools were used on the robot.

**IK Solver Improvements**

Another improvement would be to develop an analytical IK solution for the ROS model of the robot. With the robot model properly represented in OpenRAVE, a functioning IKFast plugin could be created to enable analytical IK solving of the robot’s end effector, which would improve the execution time when solving the IK during the entry point workspace checking. Alternatively, a custom kinematic solver plugin can be written for the MoveIt! robot configuration, which would potentially allow the same IK solver used on the robot controller to be imported and used on the ROS model. However, the joint space converter would have to operate each time an IK query was called, but this would likely remain faster than the numerical IK solver currently implemented with KDL.

**Workspace Analysis Additions**

In the workspace analysis, the probe should be rotated during collision checks to make sure that a given configuration allows the probe to be rotated during treatment without hitting the MRI bore. This was not important during current testing of the neurobot since a symmetric probe was used during treatments, but more sophisti-
cated treatments using a narrow beam probe would need this enhancement, albeit at a performance cost. The resolution of the workspace checking could also be enhanced, and perhaps an analytical workspace analysis would be worth developing to reduce the processing time. Alternatively, saving workspaces for the different probe configurations or effectively caching them and loading the proper configuration given the offset specifications could massively reduce the computational time currently required for checking collisions with the bore and generating the workspace. Collision checking could also be parallelized or GPU-accelerated to reduce the time it takes to produce the workspaces.

Since a significant portion of this project involved developing a tool for completing the workspace analysis of the neurobot, it would be a valuable endeavor to make the solution more generic and release it as a freely available ROS package for conducting workspace analysis. This would take some additional development effort and a commitment from developers (or enough interest in the community) to continue supporting and updating the package as ROS continues to develop. Yet the effort would be worth it by helping WPI and AIM Lab become more public and well-recognized in the robotics community for their contributions.

**Slicer Additions**

Building a module within Slicer to handle some of the manual steps of the workflow would have increased the ease of use of the system and allowed the interface to be less complicated. A module should be built that controls and supervises the state and actions of the software system by providing an interface within Slicer itself. In evaluating the validity of the planned trajectory in Slicer, it would be helpful to have a visualization of the path, the robot, and the ultrasonic probe that could be traced along the planned trajectory. This would provide a more thorough
understanding of the positioning context of the system as it relates to the patient within the MRI machine as well as ensuring that parts of the robot would not strike the patient when executing the trajectory. Once this trajectory is checked by an operator, it can be confirmed and then executed by the robot once the safety foot pedal is pressed (necessary for all movement), improving the safety of executing trajectories generated by the software system. These additional features in a module could have improved the interaction with the motion planning system, but further improvements were not possible in the given time frame.

**Optimization of Burr Hole Location**

The patient head pose and burr hole location could be algorithmically optimized (rather than through an operator-informed decision) to ensure maximum manipulability around the tumor location. One way of achieving this could be by porting the robot model into OpenRAVE and utilizing an existing optimality planners for determining where to locate the base of industrial robot manipulators. It is likely that this technique could be replicated in ROS itself, as a manipulability index is computable for the kinematic chain of the robot, thus lending itself to possible adaptation for positioning the patient head in order to maximize the manipulability around the target site, which is of more importance in the future where multiple ablations could be performed to fully destroy the tumor conformally. Another approach at this problem is to sample entry points in the robot’s workspace and compute the volume of the reachable element workspace for each one. In this way, it can be demonstrated which entry points provide the largest access in the patient, and this could be used to position the patient head and/or burr hole.
Automation of Head Segmentation

There were a number of human-in-the-loop tasks to complete for completing the head segmentation process for obtaining the volume of the patient. This included setting the threshold for masking, removing islands, and running filters on the selected volume. Besides the segmentation masking step, which was done by a threshold specified by a user since background noise could change between scanners, the other steps can use preset parameters. Most likely, this automation could be achieved by creating a Slicer module to use the segmentation editor to automate the process. The serialization and deserialization of the polydata objects also can take a couple dozen seconds to complete, so reducing the polygon count within Slicer using a module before packaging the data would speed up the workflow.

Bridging and Progress Towards Production-Ready Software

More development support for the ROS-OpenIGTLink-Bridge package would open up the door for greater interoperability between the medical and robotics research worlds. However, there are some soul-seeking questions that need to be answered. How much reinventing happens on each side of the fence? There are excellent visualization tools for each platform, but it would be difficult to migrate features from one into another; perhaps there must always be two interfaces for interacting with medical data and robotics sensory information.

Additionally, a lot of these tools are for research purposes only and could be hard to attain Food and Drug Administration approval for human treatments due to validating operational safety on a large, open-source software stack and ensuring proper security protocols (to ensure patient privacy and safety) on both ROS and OpenIGTLink communications, not to mention potential licensing issues. To bring this software system into production, the open-source software platforms would
likely be forgone to enable greater control over revisions, licensing freedoms, and ability to validate and certify the custom-built software. Surpassing these more intricate issues can necessitate developing in-house solutions, which greatly increases cost and tends to result in reproducing solutions to basic problems that already exist. Perhaps there is a greater conversation that needs to happen in this front, but at the same time, it is sometimes better to prove out a system concept quickly with the best tools on hand rather than miss an opportunity for sake of technicality to offer a novel treatment system for treating clinical threats to human life and improving the quality of living for those suffering from neurological diseases.
Appendix A

Multimedia Resources

The presentation of this thesis can be found at [https://docs.google.com/presentation/d/1Phs1Y4dXaJT5bG8Cnvt8NBwlvA2UG12JH63xtuhlZ850/edit?usp=sharing](https://docs.google.com/presentation/d/1Phs1Y4dXaJT5bG8Cnvt8NBwlvA2UG12JH63xtuhlZ850/edit?usp=sharing).

Videos of the system in operation are listed below:

1. Robot state in ROS being updated by actual robot: [https://youtu.be/k7m0hYMJvkc](https://youtu.be/k7m0hYMJvkc)
2. Robot and joint movements visualized in RViz: [https://youtu.be/A8vxQcD-D1M](https://youtu.be/A8vxQcD-D1M)
4. Collision-aware IK in MoveIt! Scene: [https://youtu.be/aWqPCrOiD8k](https://youtu.be/aWqPCrOiD8k)
Appendix B

Technical Documentation

The following sections are latex-generated versions of the documentation stored on the repository as markdown-style files.

B.1 Neuro Robot RVIZ/ROS Interface

This repository contains the following packages:

- neuro_robot_model_2018_final - URDF file of the neuro robot and bringup launch files for visualization of the most recent version
- neuro_robot_model_2018_final_moveit_config - The moveit package for the neuro robot
- neuro_planning - Includes workspace analyzer, which computes the workspace and handles trajectory following
- motion_planning_utils - Conversion nodes for going between igtl bridge messages and normal ROS messages.

B.2 Virtual Box Image

This package is already installed in a VM. Recommend using 4GB of RAM with 2-3 processors, with 3 or more preferrable if running the RViz interface.

For running this software in a VM, install Virtual Box (recommended) or VMWare for your operating system.

You can then load an image of Ubuntu with this package already installed at

\research.wpi.edu\users\Christopher_Bove\Motion_Planning_VM_Images\Ubuntu 16.04 ROS-MoveIt

B.2.1 Starting the Software Using the VM

- Start Oracle VM Virtual Box.
• Click the **Ubuntu 16.04 ROS-MoveIt** image and click start. The VM will begin booting.
• While this is occurring, open up Slicer.
  – Open the IGTLink module (Modules->IGT->OpenIGTLinkIF).
  – Click the + Button to add a connector to the scene.
  – Set the type to be **Server**
  – Select the Active checkbox to start the server. The status should turn to **Wait** in the connectors list at the top.
  – Note the IP address of this computer (using ipconfig on windows or ifconfig on linux).

• Configure the VM for running the code:
  – On the VM desktop (it logs in automatically), double click the Link to igtl_bridge_slicer.launch and configure the IP address and port in the launch file to match the server created in Slicer. Save (ctrl-s) and close the editor.
  – On the VM desktop, double click the Link to igtl_bridge_robot.launch and configure the IP address and port in the launch file that is appropriate for the robot. Save (ctrl-s) and close the editor.
  – On the VM desktop, double click the Link to start_workspace_software.sh. This opens a terminal and launches the software. Once you see, **All done! You can start planning now** in green followed by some matrices, you can start using the code.

**Other Options**

• If you want to see an RViz visualization (at the expense of a lot more CPU and GPU overhead on the VM), instead of using the script, do:

  rosrun neuro_planning workspace_analysisbringup.launch

• To check for updates to this package and rebuild the software, click the “Link to pull_software_changes.sh” on the desktop.

**B.2.2 Utilizing the Software**

• On the robot controller:
  – Send the registration transform “scanner_to_robot_reg.”
  – Send the “robot_to_treatment” offset.
  – Send the “robot_to_entry” offset. This starts the generation of the entry workspace by the program. You should see a progress bar in the VM terminal window.
* Note - to send this over ROS, do rostopic pub /robot_to_entry std_msgs/Float64 "data: 0.01"

- In Slicer (for v 4.8.1):
  - The entry workspace will show up automatically. To adjust its appearance:
    * Click Modules->Models.
    * Click “reachable_entry_work”.
    * Under Display:
      - Change Color to whatever you prefer (default is OK for the entry workspace - red works well for element workspace given entry point).
      - Change Opacity to 0.1 or 0.2.
      - Change Visible Sides to All.
  - Now, define the entry point.
    * Scroll the Coronal plane (green) into the workspace desired.
    * Select the “Create and Place Fiducial” Button in the center of the top toolbar (A blue arrow with a red dot on top), and click where you want the entry point on a 2D slice.
    * Click Modules->Markups, and select the point in the drop down and rename it to entry_point.
  - Now, we send it over OpenIGTLink.
    * Go back to the OpenIGTLinkIF module.
    * Under I/O Configuration, expand the IGTLConnector, and then click “OUT”.
    * On the drop down menu, select the Fiducials “F” and click the “+” button.
    * Now, expand the “OUT” menu and click the MarkupsFiducials Point “F” and click the send button. The program will compute a quick workspace check before developing a more thorough analysis of the element workspace given that entry point.
    * The workspace will show up under the “Models” module, and the characteristics can be adjusted like the “reachable_entry_work” as before.
    * If you resend the Markup Fiducials over OpenIGTLink, the software will again compute the element workspace given the entry point.

**B.3 Running the Software**

There are a few different functions provided by this package. You can run the ROS robot model visualizations or the workspace analysis toolkit.
B.3.1 Running Robot Model Visualizations

This will launch RVIZ with a Python GUI to control the joint values:

```
roslaunch neuro_robot_model_2018_final rviz_display.launch
```

This launches MoveIt! in RVIZ with the ability to generate trajectories by dragging around the end effector or other planning groups:

```
roslaunch neuro_robot_2018_final_moveit_generated demo.launch
```

B.3.2 Running Workspace Analysis

Setting up OpenIGTLink Connections

The software connects to 2 OpenIGTLink servers - one to the robot and one to a Slicer interface. The ports and IP’s should be configured properly in these two files before launching the software:

- motion_planning_utils/launch/igtl_bridge_slicer.launch
- motion_planning_utils/launch/igtl_bridge_robot.launch

If the igtl bridge nodes cannot connect to the server, they will die and keep respawning (in case the crash was caused by a temporary network outage). If you see a bunch of error text in red, this is probably what happened.

Launching the Software

This will launch RVIZ with the Neuroablation moveit robot model and enable planning in the environment.

```
roslaunch neuro_planning workspace_analysis_bringup.launch
```

Interactive Marker Controls  Workspace visualizer will create an interactive marker in RVIZ that can be right-clicked to compute the workspace of the robot and spawn different collision objects. The following are some of the options:

* "Workspace (no environ)" - Generate element workspace without checking collisions with bore.
* "Workspace with Bore" - Generate element workspace considering collisions with bore.
* "Entry Workspace with Bore" - Compute entry point workspace given bore collisions.
* "Workspace given Entry Point" - Compute element workspace given the entry point as selected by the interactive marker pose.
* "Check Collision" - States in the console if the robot is currently in a collision state with the published collision environment.
* "Add MRI Bore" - Add the MRI Bore as a collision object.
* "Remove MRI Bore" - Removes the MRI Bore from the world.
* "Add Entry Mesh" - Adds a proxy head mesh with a hole around the canula interactive marker location.
* "Remove Entry Mesh" - Removes the head mesh from the world.
Open IGTLink Reactions  Primarily, actions are controlled through OpenIGTLink interfaces. The following is the expected flow of events: 1. Expects the Transform to be published between the optical frame and z_frame registration (named “scanner_to_robot_reg”). 2. When the robot to entry offset is published (“robot_to_entry”), the entry workspace is generated and sent over OpenIGTLink. 3. When an entry point is published (“entry_point”), the workspace given that entry point is generated and sent over OpenIGTLink. It first sends a rough outline of the entry workspace then a more complete one after about a minute.

B.4 OS and ROS Installation

For full installation on a new computer, do the following.

B.4.1 Installing Ubuntu

Download an Ubuntu 16.04 image, make a bootable USB stick, and boot it on your computer/VM.

When on the screen “Preparing to Install Ubuntu” select options to download updates and install third-party software.

Install as normal.

Reboot and run updates from the Software Updater. Also install additional drivers as needed (System Settings->Software and Updates -> Additional Drivers

B.4.2 Installing ROS Kinetic

You can use a one-line install found here which is reproduced below: http://wiki.ros.org/ROS/Installation/TwoLineInstall

wget https://raw.githubusercontent.com/oroca/oroca-ros-pkg/master/ros_install.sh && chmod 755 ./ros_install.sh && ./ros_install.sh
catkin_ws kinetic

Or you can follow the ROS Wiki page for Kinetic.
Then install moveit and other dependencies:

sudo apt-get update
sudo apt-get install ros-kinetic-moveit python-catkin-tools ros-kinetic-tf2-sensor-msgs

B.4.3 Installing 3D Slicer

Download the latest version of slicer: http://download.slicer.org Unpack the archive file:

cd ~/Downloads
tar -xvzf Slicer-4.8.1-linux-amd64.tar.gz
mv Slicer-4.8.1-linux-amd64 ~/rm Slicer-4.8.1-linux-amd64.tar.gz
Now, open the file explorer and navigate to home -> Slicer-xxx and double click on the Slicer executable to start Slicer.

Once it’s open, right click on the icon on the dock in the left and click “Add to Dash” so you can easily open up slicer from the GUI. You’ll have to edit the destkop entry, so do

```
gedit .local/share/applications/slicerapp-real.desktop
```

And put the following contents in:

```
[Desktop Entry]
Encoding=UTF-8
Version=1.0
Type=Application
Name=3D Slicer 4.8.1
Icon=slicerapp-real.png
Path=/home/chris/Slicer-4.8.1-linux-amd64
Exec=/home/chris/Slicer-4.8.1-linux-amd64/Slicer
StartupNotify=false
StartupWMClass=SlicerApp-real
OnlyShowIn=Unity;
X-UnityGenerated=true
```

### B.5 Workspace Setup

#### B.5.1 OpenIGTLink Setup

The instructions below are taken from: https://github.com/openigtlink/ROS-IGTL-Bridge

Complete these commands below:

```bash
cd ~/
mkdir -p igtl/OpenIGTLink-build
cd igtl/
git clone https://github.com/openigtlink/OpenIGTLink.git
cd OpenIGTLink-build
cmake -DBUILD_SHARED_LIBS:BOOL=ON .. /OpenIGTLink
make
```

#### B.5.2 ROS Workspace

It’s now recommended to use catkin build tools for compiling a catkin workspace. We’re going to unfortunately remove the workspace generated in the one-step install and set it up with catkin build tools:

```
Setup Catkin build tools workspace:
```
B.5.3 Setup ROS OpenIGTLink Bridge

Recommend adding the extra cmake arg for building at release optimizations.

cd ~/catkin_ws/src
git clone https://github.com/ChrisBove/ROS-IGTL-Bridge.git
cd ~/catkin
catkin config --cmake-args -DOpenIGTLink_DIR:PATH=$HOME/igtl/OpenIGTLink-build -DCMAKE_BUILD_TYPE=Release
catkin build

B.5.4 Setup Neuro Motion Planning Packages

Clone this repo into your catkin workspace:

cd ~/catkin_ws/src
git clone http://fischerlab2.wpi.edu:7990/scm/robctrl/motion_planning_software.git
# OR you can include your username to avoid typing it each time you pull or push:
# git clone http://USERNAME@fischerlab2.wpi.edu:7990/scm/robctrl/motion_planning_software.git

cd ..
catkin build
source devel/setup.bash

Re-source your bashrc with source ~/.bashrc (the last step) or close and re-open your terminal to have the package changes loaded in your terminal so you can run them.

B.5.5 Other Nice-to-haves

System Time on Dual Boot

Windows and Ubuntu use different times for interpreting what the system clock represents. Windows uses local, Ubuntu uses UTC. This can cause issues with clocks jumping when dual booting, so changing Ubuntu to use local is easiest.

timedatectl set-local-rtc 1 --adjust-system-clock
System Load Indicator

Puts little graphs of CPU, Disk, RAM, and network usage in your task bar. Pretty handy. Install with the Ubuntu Software Center or `sudo apt-get install indicator-multiload`.

Terminator

Arranging multiple terminator windows on the same screen. Install in software center or `sudo apt-get install terminator`.

Eclipse

Install Java first: http://ubuntuhandbook.org/index.php/2014/02/install-oracle-java-6-7-or-8-ubuntu-14-04/

```bash
sudo add-apt-repository ppa:webupd8team/java
sudo apt-get update
sudo apt-get install oracle-java8-installer
```

Download and extract Eclipse Neon C++ located here to your home drive: https://www.eclipse.org/downloads/packages/release/neon/3

Note: We use an older version since Oxygen has a bug which removed the preprocessor includes from project properties (https://bugs.eclipse.org/bugs/show_bug.cgi?id=529393).

You can use the install guide here or just follow the directions below: http://ubuntuhandbook.org/index.php/to-install-the-latest-eclipse-in-ubuntu-16-04-15-10/ Modify the desktop entry like so with `gedit $HOME/.local/share/applications/eclipse.desktop` and put the contents in below (modifying paths if/when necessary):

```ini
[Desktop Entry]
Name=Eclipse
Type=Application
Exec=bash -i -c "$HOME/eclipse/eclipse"
Terminal=false
Icon=/home/USERNAME_HERE/eclipse/icon.xpm
Comment=Integrated Development Environment
NoDisplay=false
Categories=Development;IDE;
Name[en]=Eclipse
```

You can reference the guides here and the ROS guide at 2.3 or follow the directions below:

Eclipse configuration: * Go to Windows->Preferences->C/C++->Build->Settings->Discovery * On “CDT GCC Build Output Parser [Shared]” under “Container to keep discovered entries” select “Project” and click Apply. * On “CDT GCC Built-in
Compiler Settings [Shared]” configure the line to read as: `{COMMAND} {FLAGS} -E -P -v -D -std=c++11 "{INPUTS}" * Click Apply then OK to close the window.

You’ll need to use a modified catkin build and run a script to generate an Eclipse project file. I like to combine these steps into one script. Do `gedit $HOME/catkin_ws/eclipse_build.sh` and paste the following:

```
catkin build --force-cmake -G"Eclipse CDT4 - Unix Makefiles" -DCMAKE_BUILD_TYPE=Release -DCMAKE_CXX_COMPILER_ARG1=-std=c++11

ROOT=$PWD
cd build
for PROJECT in `find $PWD -name .project`; do
  DIR=`dirname $PROJECT`
echo $DIR
cd $DIR
  awk -f $(rospack find mk)/eclipse.awk .project > .project_with_env && mv .project_with_env .project
done
cd $ROOT
```

Save the file, then do `chmod +x $HOME/catkin_ws/eclipse_build.sh` and run it by cd’ing to your catkin_ws root where it’s located and doing `./eclipse_build.sh`

Open Eclipse from the Unity launcher and do: * For workspace window, just use the default. * File->Import->General->Existing Projects into Workspace -> Next * Select Root Directory: catkin_ws * Click Finish * Right click on each Project->Properties->C/C++ General->Preprocessor Includes->Providers * Check “CDT GCC Build Output Parser [Shared]” and “Use global provider shared between projects” on its Language Settings Provider Options. * Check “CDT GCC Built-in Compiler Settings [Shared] and”Use global provider shared between projects” on its Language Settings Provider Options. * Click Apply and OK. * Right click each project->Index->Rebuild.

Let the indexer chew on that for a while... And then you should be set!

### B.6 OMPL Windows Installation

This section includes some of the information found and steps for installing OMPL on Windows, which was eventually abandoned in favor of utilizing the OMPL installation included in ROS MoveIt!. It is included here as a reference for other individuals who may wish to continue these efforts.

Install OMPL on Windows:

- [http://ompl.kavrakilab.org/installation.html](http://ompl.kavrakilab.org/installation.html)
- [https://bitbucket.org/ompl/ompl](https://bitbucket.org/ompl/ompl)
Install python 2.7: Make sure to select to add python to system path
Install these things: [http://pygccxml.readthedocs.io/en/develop/install.html](http://pygccxml.readthedocs.io/en/develop/install.html)

```bash
pip install pygccxml
```


CMAKE: [https://cmake.org/download/](https://cmake.org/download/)
Add to system path in install options

```bash
$env: Path += ";D:\MinGW\bin"
cmake -G "MinGW Makefiles" ../..
    [-DCMAKE_INSTALL_PREFIX=../..]
cmake -G "MinGW Makefiles" ../..
```

Note: cannot have AVR installed on the same machine...

[https://andres.jaimes.net/718/how-to-install-the-c-boost-libraries-on-windows/](https://andres.jaimes.net/718/how-to-install-the-c-boost-libraries-on-windows/)
Download and install boost for Windows: [https://dl.bintray.com/boostorg/release/1.65.0/binaries/](https://dl.bintray.com/boostorg/release/1.65.0/binaries/)

Need 1.64 [https://sourceforge.net/projects/boost/files/boost-binaries/1.64.0/](https://sourceforge.net/projects/boost/files/boost-binaries/1.64.0/)

Do:

```bash
$env: Path += ";D:\MinGW\bin"
\cd into the directory
\bootstrap.bat gcc
\b2.exe install —prefix=d:/boost_install toolset=gcc

b2 install —prefix=c:/installation/path toolset=gcc
```

```bash
$env:BOOST_ROOT = "D:\boost_1_64_install"$
$env:BOOST_LIBRARYDIR = "D:\boost_1_64_install\lib"
$env:BOOSTINCLUDEDIR = "D:\boost_1_64_install\include"
```

[https://andres.jaimes.net/718/how-to-install-the-c-boost-libraries-on-windows/](https://andres.jaimes.net/718/how-to-install-the-c-boost-libraries-on-windows/)

Apparently, boost is always built on a CMake version that is in the development pipeline... [https://github.com/Kitware/CMake/commit/fa114e7d708b76f33878f8f82a6c2ae50c1c10f](https://github.com/Kitware/CMake/commit/fa114e7d708b76f33878f8f82a6c2ae50c1c10f)
So have to reinstall an older version of boost or build CMake from source.

Install [https://tortoisehg.bitbucket.io/](https://tortoisehg.bitbucket.io/) to pull and build Cmake from source...
Appendix C

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