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**The Design and Evaluation of a Four
Degree-Of-Freedom Cooperative Handheld
Robotic Device in a Simplified 3D
Construction Environment**

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Executive Summary

Acknowledgements

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1 Introduction

1.1 Current Robotics

Advancements in the field of robotics have been plentiful over recent years, stemming from increased knowledge of the subject, improved processing power, as well as mechanical and material developments. As such, the capabilities afforded by robots have also increased, enabling a range of applications that, until recently, would not have been possible. As the area has progressed since its inception, it has naturally split into two distinct categories; external robotics and wearable technology. They are each characterised by the respective workspace in which they operate. This branching has appeared to come about due to typical robotic task requirements catering to either one or the other.

External robotics are distinctly separate from humans in terms of physical proximity. They are becoming increasingly common in large-scale industrial spaces due to their ability to provide improved efficiency and task precision through autonomy. A typical application is the automotive industry, in which production has been accelerated through the introduction of external robotics, performing specific assembly and painting operations, mostly in detached workspaces from the human workers. The other discrete area of robotics is referred to as ‘wearable technology’, which, as implied by its name, is concerned with robotics that operate in the immediate personal space of the human user. This domain is typically associated with the augmentation of human actions, such as increasing physical strength or permitting certain motions that those with disabilities have been rendered incapable of. A well known use of this technology is the ‘exoskeleton’, which can be tailored to accomplish either of these two functions through structural support and/or actuation.

However, both of these forms of robotics comprise inherent limitations, reducing their suitability to certain situations. As such, certain physical tasks exist that cater to neither domain. One case where this is evident is the *tool space*. This area encompasses the typical range of tasks that humans have long since utilised passive tools to accomplish, along with exceeding performance possible by hand. This area is closely associated with flexibility, where unique tools can be used to deal with the same general task in various environments, or with completely different tasks altogether. Conversely, external robotics are typically limited to a rigidly defined workspace, performing very specific and repetitive tasks that they were designed for. Whilst wearable technology is more suited to this flexibility, it suffers from its limited adaptability to different users. Devices such as exoskeletons are very costly, and are uniquely tailored to individual body dimensions. This lack of generality equates to reduced applicability to the tool space, in which it is essential that other users are able to operate the tool when a certain individual is absent, or where a tool is marketed to a wide variety of customers.

1.2 Motivation

A proposed solution to bridge the gap between the two established forms of robotics, providing performance improvements in areas that neither are suited to, is *handheld robotics*. This area aims to combine the key benefits of robotic efficiency and augmentation of human performance afforded by both fields, whilst maintaining the flexibility from human intuition in traditional tool use. Early work in this area has hinted at a wide range of potential applications where handheld robotics may excel, particularly in the fields of assembly, construction and maintenance/quality control. The fundamentals of their operation is that they cooperate with a human user, who in turn, positions the robot in a general location in which it can perform its desired operations, leading to improved performance. This sequence of events however, relies heavily on a key aspect of the device; its *task knowledge*. This refers to XXXXXX

1.3 Aims and Objectives

1.4 Document Layout

1.5 Summary

2 Background and Context

2.1 Robotics with Humans In-The-Loop

Although robots have become vastly more capable over recent years with technological improvements, many tasks still require some form of human assistance in order for the objectives to be adequately met. Robots can be perfectly equipped to handle operations which involve precise motion or repetitive tasks, which may be present physical challenges to humans [1]. Conversely, they still lack ability in areas which require the high-level cognitive reasoning that comes naturally to humans. As such, there exists many applications in which humans are in-the-loop for robotic operations. This applies to both forms of robotics that the handheld area sits between; external robots and, especially so, wearable technology.

2.1.1 Safety

Perhaps the most important factor in these situations is how the robot receives and processes user intent. An input from the user should be analysed by the robot in terms of importance and the choice of subsequent actions should be made appropriately. A prime example of where this is of paramount concern is in vehicular operations, where potential accidents could be catastrophic. Anderson et al [2] investigated a semi-autonomous hazard avoidance framework and tested it in a real-life driving environment, where the human driver received full control during non-threat situations. However, in times of threat detection, the control system became fully autonomous and guided the car to the safety. In this situation, the robot assumed the expertise of hazard assessment and reaction time, whereas the human provided general navigation that was accepted by the system under normal circumstances. This is one extreme, where the human input is essentially ignored under certain conditions for the safety of the user and those nearby.

The other end of the spectrum, human input must be immediately responded to in order to act safely, where it is assumed the human possesses greater threat detection. Thus safety to both the human and robot must be positively ensured [3]. In industrial applications, the robot often operates in an isolated workspace from the human, where an extreme human input would constitute an emergency button or entering the workspace via a safety door/light gate, which the robot must respond to by halting all physical actions for human safety [4]. However, it is becoming increasingly common for the robot and human to share the same workspace, occasionally cooperating in a physical manner, such as the humanoid robot presented by Ott et al [5] and illustrated in Figure 2.1, where parts can be passed between the robot and human.

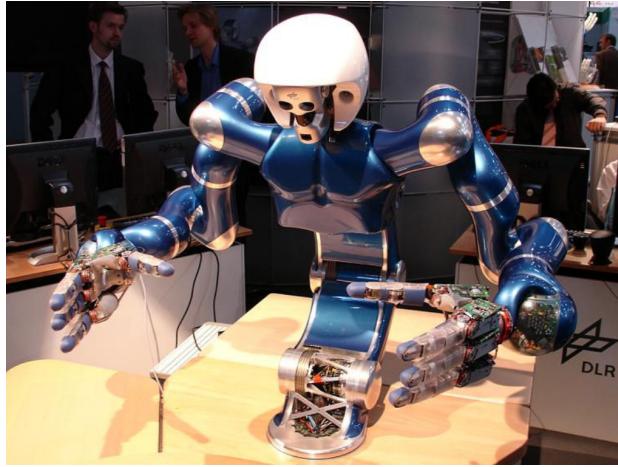


Figure 2.1: Two-Arm Humanoid for Dexterous Manipulation, image from [5].

Existing operational requirements for collaborative robots have been specified in ISO-10218, which focus on defining the upper limit of either the tool centre point velocity, the dynamic power, or the static force. However, as discussed by Haddadin et al [3], these values are merely based on heuristics. As such, more specific severity indices were introduced, taking from biomechanically motivated quantities. In the case of this project, the robot was designed with the consideration of limiting the required dynamic power to suitable levels to be deemed safe as a collaborative robot. This entailed lightweight and flexible material design choices for moving parts, greatly reducing any risk to a human user.

2.1.2 Autonomy Study

The reliability offered by human controllers can rarely be matched by autonomous agents. Similarly, the efficiency afforded by robots is highly attractive in physical tasks. Finding a strong balance between the two is essential in successfully accomplishing a task. Heger et al [6] investigate a mode of shared control, referred to as “Sliding Autonomy”. Fundamentally, this is an analysis to find the optimal flow of control between autonomous agents and a human operator for a given task, where an operator observes the operation and intervenes when necessary. A scenario was performed based on the model where three robots work cooperatively with a human operator to assemble a physical structure, with numerous failure modes deliberately implemented. The results indicated that the cooperation compromise resulted in task efficiency similar to a fully autonomous mode, while maintaining a level of reliability associated with teleoperation systems (remote human control). The constant attention of the human, without being cognitively demanding, improved the success rate while acting as a safety net for failures that the robot alone cannot recover from.

There are plenty of other investigations into the study of variable autonomy, such as in [7], which looks into variable interactions in space missions, and in [8], which concentrates on the costs (time delays and other effects) of such transfer-of-control systems. Although this area of interchangeable control is mostly focused on with regard to external robotics, it is still relevant to the handheld field. As covered in [9], examining the varying degrees of autonomy that a handheld robot may exhibit for a given task is necessary to locate the control set-up to give optimal performance. A different approach was taken to shared control in [10], where the efficiency offered by both members in the cooperation was measured at each time instant from the point of view of reactive strategy. The experiment in this paper focused on the shared navigation of a Pioneer AT robot, where the commands of both the human and the robot were weighted and linearly combined, resulting in a single motor operation, as illustrated in Figure 2.2.

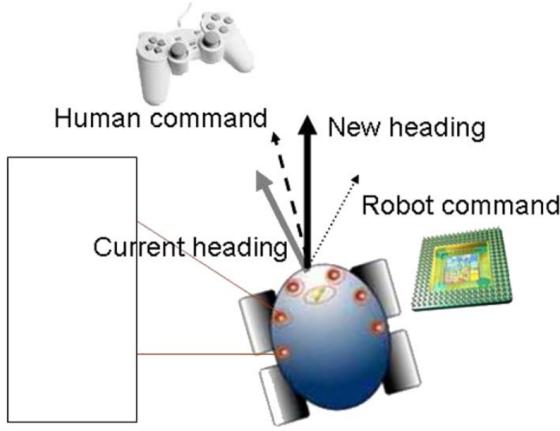


Figure 2.2: Resulting command from human and robot inputs, image from [10].

The results of this experiment showed that the weighting between the two inputs led to performance which, although not matching that offered by each member alone, did seem to combine the benefits of both when taking a range of scenarios into consideration. This is a rare situation in which the task at hand could be performed with equal skill by both human and robot, and caters to the speciality of neither. However, the approach of taking a weighting between the two inputs is interesting.

It may be worth investigating such a method with handheld robots in the future. This would entail combining the two signals of human intention and task planning, and deciding the action based on the weighting the given scenario. This could add a degree of flexibility to the robot, adapting to new changes in the task, as identified by the human, or compromising between the two signals. With such a case, different settings could be explored by tuning the weighting given to each signal on a scale with a range of values, as opposed to just three discrete levels of autonomy as was the case in [9]. However, such an investigation would require an extensive user study, with participants performing tasks repeatedly for a large variety of weightings. This is outside the scope of what is hoped the current project will achieve.

2.1.3 Collaboration with Assistive Robotics

Rather than replacing human operations, robotics can be employed to assist in them, maintaining human flexibility while increasing productivity. One such case where this is applicable is in the manufacturing industry, particularly assembly. Robots with highly specific purposes have long been used on assembly lines to increase output, performing simple and repetitive tasks at a faster rate than is possible with human workers. However, humans are frequently used for more complicated procedures that require motion in multiple degrees of freedom. The adaptability and range of motion of human workers is often unrivalled compared with traditional robotics, as well as being much more available.

A sensible compromise is investigated in [11], where robotics are used to provide assistance to a human-oriented assembly task. A Hidden Markov Model (HMM) based analysis was used so that the robot may predict the next stages of the task, lending assistance by passing the required components to the user at the appropriate time so that they may continue with the operation. As a pre-requisite for any useful interaction between the human and the robot, the robot must first be trained. In order to accomplish this, the assembly tasks were performed solely by the human whilst sensors were used at various hand and body positions, so that the task motions could be adequately tracked. A similar study, focused primarily on the timing of assistance, contains an in-depth description of this preliminary set-up [12]. The task comprised the assembly of a six-block tower, bolting the blocks together in varying manners. Once the data was collected for the human performed task with sensors, the results were used to form the model.

The model generated in [11] could then be transferred to the robotic platform JAHIR (Joint-Action for

Humans and Industrial Robots). This hybrid assembly platform for the context outlined in the paper is illustrated in Figure 2.3. Both the robot and the human share a designated workspace, whereas the storage areas for the parts are only accessible by the robot. An interaction area exists which overlaps with the human workspace, in which the robot may assist by transporting required parts to within human reach. There are numerous options for estimating the progress of human-performed tasks, one such technique is the one used for the preliminary experiment; strategically placed accelerometers and microphones. This technique is outlined in detail in [13], where most workshop activities could be identified at 84.4% accuracy. In the case of the preliminary experiment in [11], pertaining solely to the assembly task, greater accuracies were achieved.

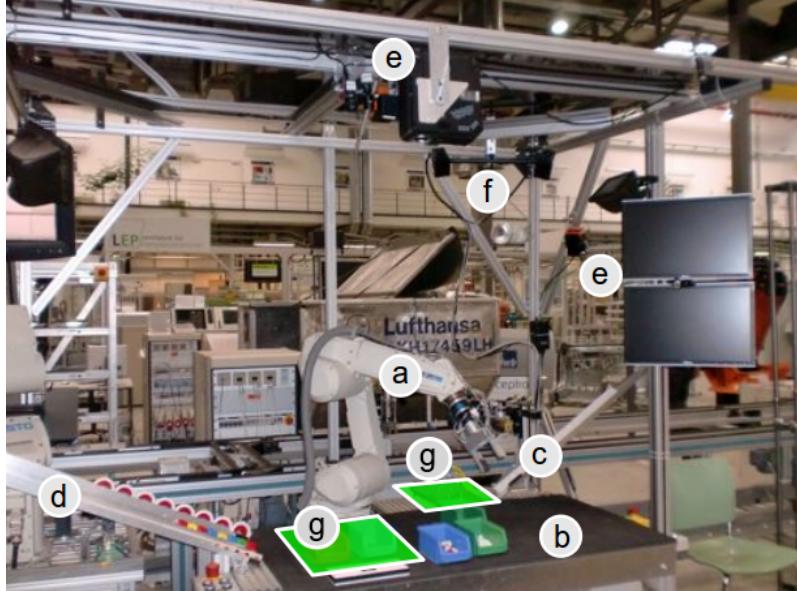


Figure 2.3: JAHIR assembly demonstration platform - a) Industrial Robot, b) Shared Workspace, c) Slide for Assembly blocks, d) Slides for Other Parts, e) Sensor Devices for Input/Output, f) ARTrack System g) Storage Space for Robot. Image from [11].

However, in reality, it is not feasible for assembly workers to wear a plethora of wired sensors while performing tasks. The physical and mental comfort is greatly reduced under such conditions, with the sensors creating a feeling of intrusiveness. A viable solution to this issue is the use of occupancy maps, which are commonly employed in robotics to determine position and motion, providing a robust approach to spatial perception and navigation problems. This technique, covered in depth in [14], was implemented for the robot assisted assembly task in [11], using the three-dimensional velocity, acceleration and jerk as input to the individual HMMs. This idea of measuring movement in physical space without excessive wiring is an essential component to the viability of handheld robotics, which have previously employed optical tracking from an external device as a solution. Other sensing methods will be explored in the future, with the aim of improved practicality in real-world scenarios.

This occupancy map method allowed for a recognition accuracy of 92.26% for the right hand, comparable to the recognition rate of the reference experiment that relied on sensors. Therefore for the vast majority of the time, the system could accurately predict the task progress, and respond with the correct actions accordingly. The key here is that the robot must be context-aware. This enables much greater cognitive capabilities and a higher standard of interaction with the user. Cognitive Technical Systems (CoTeSys) is a group with a research focus on sensing and actuation in the physical world. As explained in [15], this allows the robot to exhibit behaviours that suggest it can ‘learn’. The foundations of cognition in humans and animals are used to develop cognitive models, which can then be applied when designing control systems for more specific technical applications.

The group uses a cognition-based perception-action closed loop as a basis for investigating cognition in technical systems, as illustrated in Figure 2.4. This architecture focuses on five main areas; perception, action, knowledge, learning and reasoning. The model may be directly applicable to the field of

handheld robotics. Task knowledge and the way in which the device receives and processes data from both the environment and the user, responding accordingly, bears many parallels to the goal of handheld robotics in general. Thus, such an architecture may be considered as a reference when designing applications for handheld robotics.

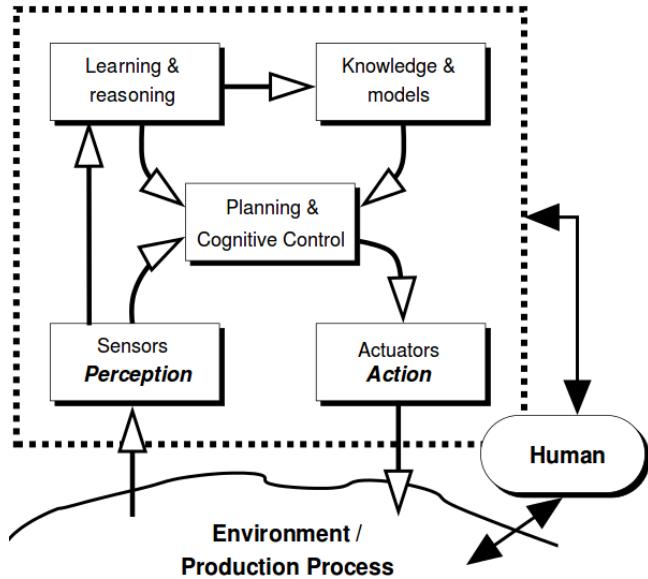


Figure 2.4: Cognitive System Architecture: Perception-Action Closed Loop, image from [15].

2.2 Robotics in Building Environments

From a fully autonomous perspective, using robots in construction environments is becoming an increasingly attractive proposition. Although machines have been utilised to assist in construction since the industrial revolution, there has typically been little in the way of automation. With advancements in robotics, they may be incorporated into building applications to reach unprecedented levels of speed and efficiency. Numerous investigations and demonstrations have been carried out of robots performing assembly operations from a set of known requirements, corresponding to ‘task knowledge’. However, most of these examples are based around two-dimensional systems, such as manipulating blocks to construct flat structures, as in [16], not considering height.

The building and construction industry has been identified as an area of strong potential for handheld robotics. It is thus used as background to highlight examples where the introduction of a new field between the two established forms of robotics could prove worthwhile and beneficial. External robots, although able to possess task knowledge, are limited in this area in their ability to deal with changing requirements and their on-site mobility. Of course, the limitations of wearable technology remain their lack of adaptability to different users. These issues in particular may have a solution in handheld robotics. On top of this, the notion of unskilled workers being able to cooperate with such a device to accomplish a task may also have applicability to the construction area.

2.2.1 Mobile Robots

Petersen et al [17] present a hardware system in which a mobile robot manipulates blocks in three-dimensional space to match desired structures. As well as this, they also produced a high-level control scheme that could enable multiple robots to autonomously share the workload of a user-specified building task. The results were successful and the prototype robot constructed structures that were greater in height than itself, doing so via climbing, navigation and manipulation. This was accomplished without the need for complex sensing and control mechanisms by the design of the simplified interconnecting structural blocks and the way in which the robot interfaces with them, as illustrated in Figure 2.5. It has been suggested that there exists a future opportunity for a series of handheld robots and respective human users to work together to accomplish tasks of much higher complexity than possible from just a single device. This would bear parallels to the work in [17] in which the task knowledge is shared and updated between multiple devices accordingly.

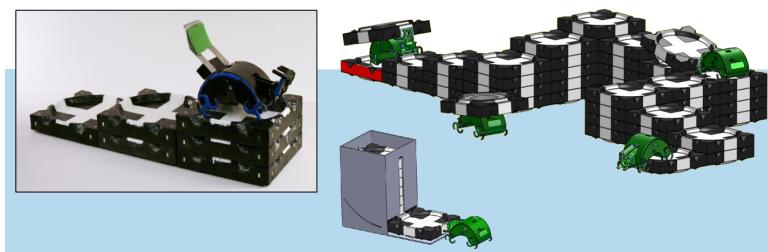


Figure 2.5: Hardware system in which robot builds specified structures. Inset: Physical image of prototype atop six-block structure. Outer image: Concept for system goal, in which multiple robots act together to distribute workload of a much larger structure. Image taken from [17].

This idea of designing the structure blocks in a passive manner so that they naturally fit together, reducing sophistication in orientation, provides a neat solution to overcoming positional errors. In this project, the physical experiments and tasks to be performed is not the sole focus, as a large portion of the scope is dedicated to the design and control methodology of a new device. However, future work will address the physical performance to a greater degree, and in such case, will likely take strong inspiration from such self-aligning building blocks as observed here. One common feature that is simpler to incorporate is the use of magnets for physical interaction with robot to block. Blocks in this project will likely have to be of a much larger scale due to the realistic accuracy expectations of

the design in the short-term, thus cost reduction in terms of material and manufacture will be a large consideration.

2.2.2 Drone-Based Assembly

With recent developments in Unmanned Aerial Vehicle (UAV) technology as a whole, it is becoming increasingly viable to deploy rotary wing platforms to interact with the physical environment. The main advantage they offer over typical ground based alternatives is improved speed and accessibility with regard to hard-to-reach regions. Moreover, using rotary wing based UAVs, such as quadcopters, over their fixed wing counterparts, allows for a more controlled operation in a tighter workspace. The design and control associated with transporting a payload with such a rotary based UAV is presented in [18]. The paper looks at different gripper designs and how successful the robot was in estimating and manipulating the mechanics of the varying payloads. Several issues arise when using UAVs to transport payloads, namely the limited weight bearing capacity and the fact that the dynamics are altered when carrying payloads.

The grippers investigated in this paper either operate by traditional clamping (impactive) or surface penetration (ingressive). However, the reduced motion of the fixed claws inevitably leads to more positional uncertainty during hovering. A solution is proposed in [19], which covers the design of a 6-DoF aerial manipulator (see Figure 2.6) for UAVs to assemble bar structures. This is made possible by developments in high power aerial platforms, reducing weight restrictions. The 6-DoF manipulator in the paper focuses on three main functions; the *capture* (approaching and grasping the bar), the *transport* (transferring load from storage to site) and the *assembling* (installing to the structure).

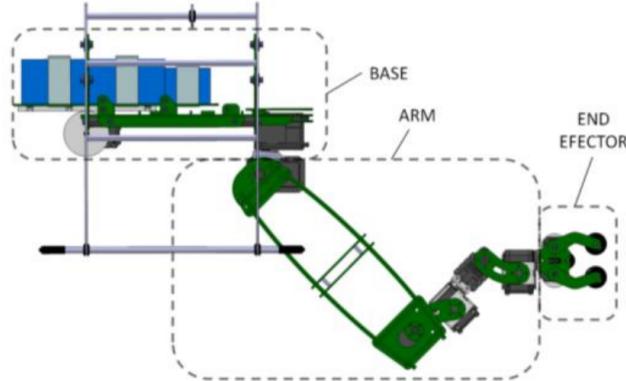


Figure 2.6: Components of Aerial 6-DoF Manipulator: Base, Arm and End Effector. Image taken from [19].

Although this aerial design remains relatively untested in terms of control, it possesses similarities to the Mark II handheld robotic design in [28] (similar DoF but contains roll), and inspiration may be taken in the future when considering alternative end effectors for assembly-based tasks. This project aims to provide further groundwork and breadth in the design of new handheld robotics, with less focus on advancing the physical capabilities, and as such magnets are to be used for interacting with external bodies due to design simplicity. However, in the future when application and capabilities become more of a focus, new end effectors may be designed in tandem with specific assembly blocks, and each of impactive, ingressive, rollers or passive-jointed magnets would all be viable options.

The feasibility of using rotary UAVs to construct building-scale structures is explored in [20], in which a project is outlined that envisions an alternative construction approach similar to that observed in nature. This entails the gradual addition of more material to an existing structure in a safe and precise manner. Despite the energy considerations, it was concluded that large-scale construction may be possible through drones, with the potential improving in parallel with advancements in drone technology. An issue that was identified, as was also the case in Section 2.2.1, is the level of precision

required for typical construction that is finer than afforded by fully mobile robots. Certainly in the early stages of handheld robotics, this will be a key issue as well. In this project it is avoided by considering part dimensions based on the accuracy of the robot, not the other way round.

This achievable accuracy problem is even more evident in aerial vehicles, and in [20], a similar solution was employed to [17], mitigating positional imprecisions through the use of modular self-aligning components. Three different approaches were investigated for such a component; *droxels*, conical interlocking *dricks*, and rounded conical interlocking *dricks*. These designs provide further inspiration for designing the assembly components for the handheld robotics construction task. In future development, when the capabilities associated with construction are more rigorously investigated, these self-aligning designs will be used as strong inspiration. However, due to the limited time-scale of this project and the comparatively smaller focus on assembly capabilities, such designs were disregarded in favour of plain blocks sized to meet the precision afforded by the new device.

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- Droxels (Drone Voxels) - Voxels are three dimensional pixels, different designs can lead to strong space-filling capabilities and allow for construction of complex shapes. They are somewhat more difficult to grasp and position but possess strong potential. One of the key issues is the manufacture complexity, with 3D printing being the most viable. See Figure 2.7.
- Conical Interlocking Dricks (Drone Bricks) - This option allows greater flexibility regarding positional tolerance. A lot of their potential lies in the fact that they can easily be constructed from more rigid structures such as concrete. However, unique blocks have to be designed to allow cornering. See Figure 2.8.
- Rounded Conical Interlocking Dricks - Similar to the standard counterpart, except with a rounded shape and two full cones on top to allow for curved walls. See Figure 2.9.

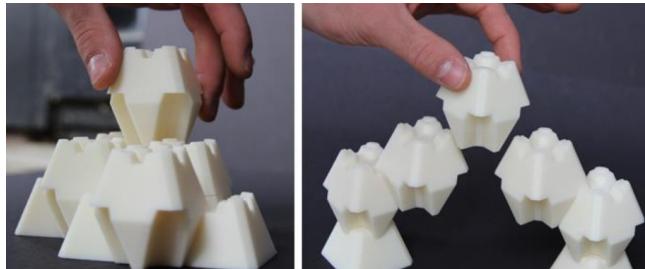


Figure 2.7: Droxel Design, image taken from [20].



Figure 2.8: Conical Interlocking Drick (CID), image taken from [20].



Figure 2.9: Round Conical Interlocking Drick (RCID), image taken from [20].

2.2.3 Scene Knowledge

One of the key considerations in robots performing construction tasks is how they navigate around the environment. The careful pre-planning of the construction environment is often necessary, so that autonomous agents can be informed of valid paths to avoid undesired collisions. Whole studies have been dedicated to this area, such as in [21]. In this paper, the autonomous agents are given prior knowledge of the environment as a whole and not just specific paths. This behaviour-oriented approach encourages vision/ultrasonic sensing to detect obstacles, reducing reliance on dead-set routes. Being more than fifteen years old, such techniques have become more established now, as in [17]. In many situations however, there is still a relatively strong dependency on a well defined environment, beyond just the relative locations of source/targets.

This reliance on environmental knowledge, as well as pre-planning of paths, is one of the areas in which handheld robotics can excel, providing an elegant solution to the guidance issues experienced by separate and autonomous robotics. As a user can transport and orientate a handheld device in any way they intuitively see fit, much greater flexibility is provided. The need for a rigidly defined environment and careful pre-planning, which is imperative to external robots, is vastly reduced with handheld robotic tools. Of course, fixed obstacles can still be made known in order to provide the most efficient movement. However, as the user is in full control of the general location of the robot, the user actively recognises and avoids obstacles with minimal cognitive effort. Hence this is an element of construction based robots that is of far less concern when designing new applications, further outlining the potential for handheld robotics as a whole.

2.3 Handheld Robotics

2.3.1 Medical Devices

The field of handheld robotics has been largely unexplored in comparison to the established areas of robotics in which it sits between; fully external and wearable technology. The vast majority of the existing work in the handheld area is focused on the medical industry. Numerous surgical devices exist within this area, supplementing the skill and performance of the surgeon during precise operations. Payne and Yang [22] review the existing and emerging trends in handheld medical robots, stating that the technical challenges differ from grounded robots in that miniaturising the devices is essential, whilst having multiple degrees of freedom is often less important.

An example of such a device is the handheld master-slave combined manipulator (MCM), developed by Matsuhira et al [23], which offers added functionality to laparoscopic surgery that is not possible with conventional instrumentation. The design entails the addition of two degrees of freedom to the typical forceps, forming a cable-driven slave hand controlled by a master grip mechanism that houses two servo motors, as illustrated in Figure 2.10. The added yaw axes and the gripper end effector then enable internal suturing tasks, where potentiometers on the master grip interface with a notebook computer for controlling the motors. Other handheld robotics in the medical field include cutting implements incorporating haptic feedback (indicating force exertion) [24]. There is also a significant amount of work regarding self-stabilising devices to compensate for physiological tremors, such as the "iTrem" [25], which utilises sensing, filtering and manipulation to enhance surgical precision, or the mechatronic cup holder that alleviates the symptoms of Parkinson's patients [26].

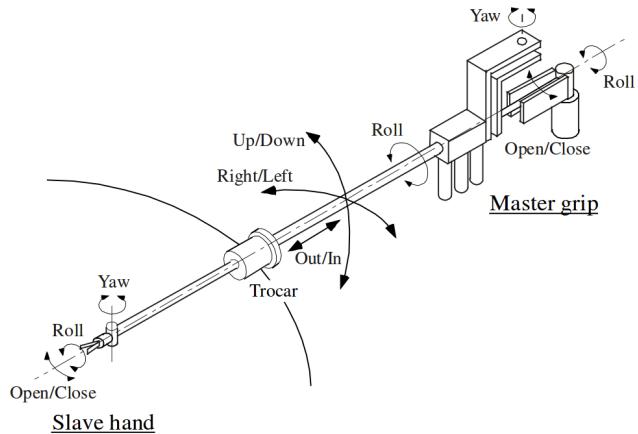


Figure 2.10: Master-Slave Manipulator for laparoscopic surgery, image from [23]

However, up until 2015, there had been little research in the way of handheld robotics in the *toolspace*. Tools have become a fundamental part of the way humans interact with the physical world, enabling the completion of tasks which would otherwise propose physical or mental challenges. They often allow operations to be performed that would otherwise be impossible, or simply reduce the task duration and workload, thus leading to an improved quality of life. One such example of work in this area is a handheld automated welding device, which enables welding to occur only when positioned correctly [27]. Although this constitutes an application of handheld robotics in the tool space, it is constrained by similar limitations to those present in the examples cited above when looking to classify this as a generalised handheld robotic tool that cooperates with the user. The aforementioned novel devices were designed with relatively few degrees of freedom and with very specific tasks in mind. This means that they are limited to single tasks, and cannot easily be adapted to alternative applications. The other key factor associated with the existing devices is their limited 'task knowledge', reducing the potential for cooperation between the device and the user.

2.4 University of Bristol Handheld Robotics Project

2.4.1 Initial Work

Recent work undertaken at the University of Bristol has formulated the “Bristol Handheld Robotics Project”. This has focused on the potential of multi-degree-of-freedom handheld robotic devices operating in the tool space. A critical component of the research has been the capability of the robot to assume task knowledge for the desired procedure. The foundation for this branch of work was laid by A.Gregg-Smith and W.W.Mayol-Cuevas in [9], where a handheld robot was constructed and evaluated in terms of user cooperation. The underlying motivation of the paper was the assertion that there exists a much greater potential for handheld robotic applications than just the medical industry.

The work in [9] covers the development of a four degree-of-freedom (4-DoF) prototype handheld robot, in which the aim was not to design a robot which can surpass human capability, but rather to evaluate the cooperation between the user and robot for various degrees of autonomy. This design and evaluation here has certainly laid the groundwork for future development of handheld robotics. The arm comprises a carbon-fibre backbone, and is cable-driven by two pairs of motors and pulleys located in the base; one opposing pair providing a degree of freedom each to the midpoint and the tip, with the other pair providing a second degree to each point, as illustrated in Figure 2.11.

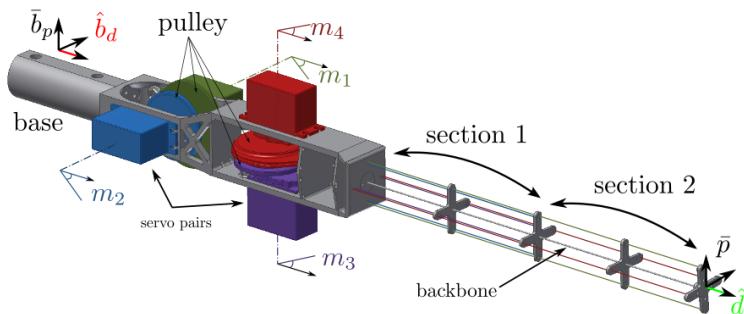


Figure 2.11: CAD model of the Four Degree-of-Freedom Mark I robot, image from [9].

The design takes inspiration from [31] which, unlike typical cable-driven designs, employs one motor to provide tension to a pair of opposing wires, which reduces the the motor count to DoF ratio from 3:2 to 1:1. An external computer is used to interface with the servomotors and the trigger, responding to external cues or program instructions by providing the necessary motion. In order to locate the robot spatially, an OptiTrack system made up of six strategically mounted cameras is used to provide motion tracking, identifying the retro-reflective markers mounted on the robot. This provides an accuracy of 0.2mm, although it was stated that the future aim would be to employ on-board sensors for portability and practicality.

In order to calibrate the robot, it was necessary to perform kinematic modelling. Conventional forward kinematics methods were not applicable to this design due to the non-linear interaction between sections 1 and 2 of the carbon-fibre backbone, shown in Figure 2.11. As such, a regression based method was employed. This entailed sending a wide range of commands to the arm and recording the resulting pose with the motion tracking equipment. In total, 14641 vectors of motor angles were sent, with the corresponding poses comprising a relative tip-to-base position and direction. A regression technique was then used on the results to interpolate between the points to acquire estimated poses for unrecorded motor vectors. This is covered in more detail in Section XXXX.

MENTION K-D TREES HERE FOR ACCELERATED SEARCH AND IMPROVED COMPUTATIONAL TIME, OR MENTION IT LATER ON IN THE CALIBRATION SECTION AS AN OPTION AND FUTURE IMPROVEMENT????XXXXXX

It is important to note that due to the limits of the carbon-fibre backbone, the arm cannot self-intersect, therefore it is safe to interpolate between the motor angles.

Inspiration was taken from this Mark I design in the development of the newer, flexible tube-based model, and naturally, similarities will be observed between the set up and control processes and the external infrastructure to interface with the robot and provide track motion. The key differences originate from the different arm material/structure, having a chain reaction of required design changes. Also, with the benefit of hindsight, design and material choices that hindered this Mark I design could be avoided, while strong points could be adapted. An example of this is the issue of the friction from the cables wearing through the base material. This led to the choice of an ultra low-friction cable material, and a stronger base material, using PLA over ABS.

ALSO MENTION SIMILARITIES WITH THE TLX STAGE IF ANY OF THIS GETS COMPLETED IN PRACTICE XXXX

, as well as the techniques used to quantify the task performance and perceived effort.

Subsequently, a more advanced six degrees-of-freedom device was constructed over a longer period of time, as covered extensively in [28]. This design comprised distinct solid sections connected by links, as opposed to a continuous backbone, to allow for a greater range of motion. Originally, this project's objectives comprised the design and evaluation of new applications for this device. However, the robot experienced an improbable critical failure during a capability demonstration, rendering it unrepairable within the project time-line. As such, the new direction was taken regarding the design and construction of a simpler four degrees-of-freedom device and the corresponding evaluation. Hence the details of the Mark II device have become largely irrelevant in terms of this project, and will therefore not be covered. Both the Mark I and II devices are made openly available at [30].

2.4.2 New Design Inspiration

Section on cable driven robots and the flexible arm paper, called "Existing Designs" or "Design inspiration" or whatever, mentioning choices similar and different to mark 1??

2.5 Designing Experiments for Handheld Robotics

At such an early stage in the development of handheld robotics, it is essential to carry out experiments that reveal as many details as possible, contributing to existing knowledge that may be used as reference as the field in general progresses. Focus must be placed on important factors such as how well the user and device cooperate, how the desired actions of the robot are communicated to the user and the benefits of using the device in terms of perceived effort and performance. The earliest experiments undertaken focused on evaluating the cooperation between the human and robot. Other experiments include investigating the effectiveness of different feedback methods for the robot in communicating desired positions, as well as the kinematics and motion performance in general. In the first two instances following, the cooperation was highlighted by the performance associated with three different levels of robot autonomy.

2.5.1 Painting Task

The painting task from [9] entailed users being shown a template pattern on a screen and assigned with tracing the pattern with "virtual paint". The motion capture system would register the position and orientation of the tip of the arm to establish the absolute direction of the painting. Then, a 'key action' (user or program oriented action associated with progressing the task) would result in virtual painting, changing the colour of the pixels that the tip is pointing to.

In both manual and semi-autonomous mode, the key action was a user oriented cue, in this case the pressing of the trigger to initiate painting. The manual mode involves a straight arm with the paint turned on with the trigger, while the semi-autonomous mode draws upon the pre-programmed task knowledge to enable movement along paths of contiguous pixels in the same target region while the

trigger is held. However, it cannot move across gaps to unpainted regions by itself, requiring the user to reposition the tip. Alternatively, the fully autonomous mode involves the robot initiating and halting the painting action where it deems appropriate, providing the key action itself. This task setup is illustrated in Figure 2.12.

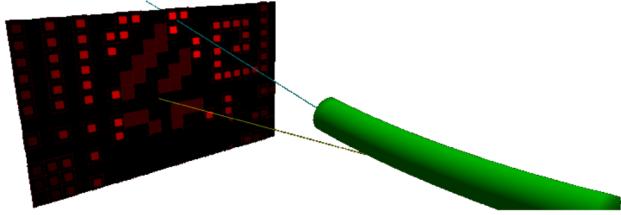


Figure 2.12: Model of painting task setup in autonomous mode (lights indicate preference of subsequent motion), image from [9]

2.5.2 Tiling Task

The tiling task, also from [9], involved fitting the end-effector with an electromagnet and picking up red or black tiles from designated hoppers before placing them in a predefined chequerboard pattern, as illustrated in Figure 2.13. The magnet is activated/deactivated by pressing the trigger, except in the case of full autonomy in which the robot fully provides the key action, controlling the magnet as necessary. In manual mode, the absence of task knowledge means the arm remains straight, responding only to the trigger press with the magnet activation. In this case, it is possible for the user to provide an incorrect pattern, therefore a template was made visible so that all users performed the task correctly.

In the two modes that comprise task knowledge, the arm moves towards the nearest hopper that it knows will provide a suitable colour, before moving towards the nearest suitable empty space on the board. In both cases, the robot will refuse to perform an action that conflicts with the task specification. In the case of full autonomy, the electromagnet is also controlled by the robot. Thus the tool will keep seeking to perform necessary actions, gesturing to the user where it should be positioned to do so, until the task is complete.

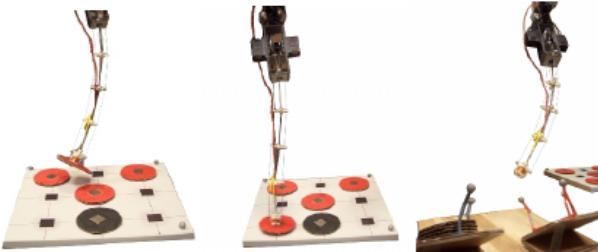


Figure 2.13: Pick and Place Tiling Task with tip of first prototype, image from [9]

The results from the previous two tasks indicated that the completion time was significantly improved with increasing degree of autonomy. The accuracy of the tasks however, showed little difference among the different modes. This was to be expected as the robot end effector was designed with an accuracy comparable to that of a human. The combined Task Load Index (TLX) scores also showed improvement with increasing autonomy, indicating that introducing cooperation with the robot reduces the perceived task effort. The individual TLX scores for various factors are shown for both the painting and tiling tasks in Figure 2.14.

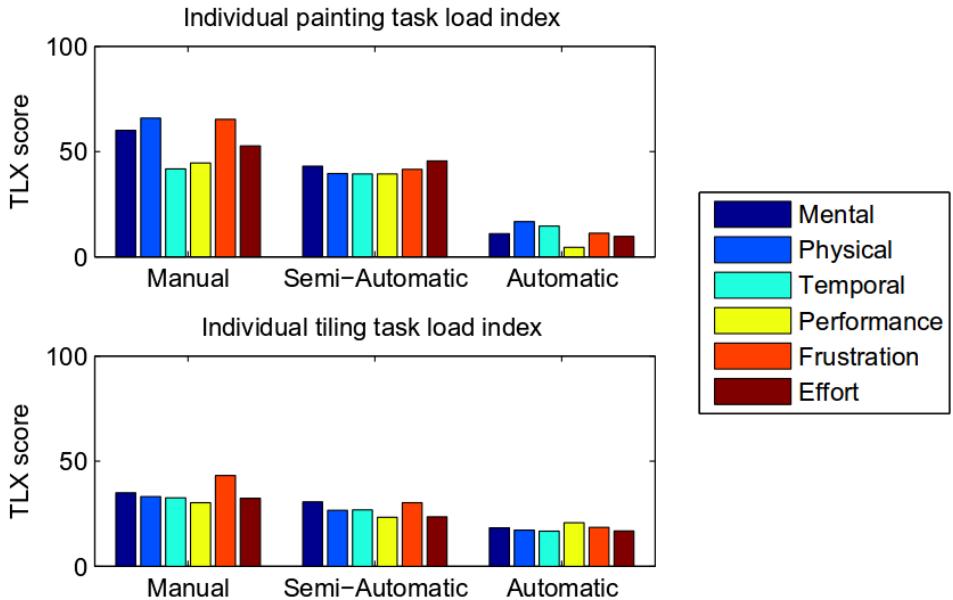


Figure 2.14: Individual TLX scores for both the tasks, image from [9]

As the task accuracy exhibited little difference among the modes, and the completion time and perceived effort showed improvement with task knowledge, it was concluded that the users cooperated well with a task-aware handheld robot, where introducing levels of automation had a positive impact on essential elements of task performance. Although sharing elements of the task among the user and robot did not negatively affect performance, maximum overall benefit (accounting for perceived effort) was attained when the robot was given full control. Of course, there is still a strong element of cooperation in this scenario as the user must move the robot to its desired location, responding to its gestures accordingly.

2.5.3 Spatial Guidance Task

Manoeuvring the robot to a specific location so that it may perform its duties is an essential element of strong cooperation. Without effectively conveying desired locations in order to proceed with the task, the usefulness of the robot is severely hindered. Thus, some form of guidance is necessitated. Fitt's law is an established method of estimating the time to reach a known destination [33], and although was originally created with a 2D perspective, it has been extended to a 3D environment, such as in 3D pointing tasks [34]. In either case, the way in which information is presented to the user from a human interaction point-of-view has often been considered to affect performance [35] [36].

As covered by A.Gregg-Smith and W.W.Mayol-Cuevas in [37], various methods were investigated in which the robot provided spatial feedback to the user during a 3D guidance task, using the more advanced 6-DoF device as mentioned in SECTION XXXXXXXX. The task comprised positioning the robot so that the end effector could be aligned to within 5mm and 5 degrees of a target pose for at least 200ms. The target pose took place at a range of locations on a small side table, and was also performed with a stand-alone handheld wand in order to evaluate the benefit of using the robot in a performance and feedback sense.

The seven conditions investigated are listed below, with the experimental set-up for the robot with virtual reality illustrated in Figure 2.15.

1. Robot provides gestures by pointing to location
2. 7 inch 2D screen mounted to robot, showing wireframe model of table and target point
3. 7 inch screen held in one hand, performing task with wand in other hand
4. Augmented reality headset (with transparent lense) while using robot
5. Augmented reality headset while performing task with wand

6. Virtual reality headset while using robot
7. Virtual reality headset while performing task with wand

In each case, an infra-red optical tracking system was used with retro-reflective markers attached to the key objects; the arm, wand, table, augmented/virtual reality headsets and the 7 inch screen. This allowed the pose of the robot to be calculated.



Figure 2.15: Spatial guidance task being performed with the handheld robot where feedback is provided via a virtual reality headset, image taken from [37].

The robot itself possessed a 6-DoF pose in world space, while the joint angles constitute a further six DoF, resulting in an overall 12-DoF pose. Solving from one of these fully defined poses to another is computationally expensive, requiring minutes to be calculated where sub-second times are needed for a responsive tool. In order to overcome this issue, the path planning ability of the human was exploited, reducing the robot path-planning problem to 3-DoF. A variant of informed Rapidly Exploring Random Trees (RRT*) [38] was then produced in [37] to calculate the shortest route from robot to target, using the frame of reference of the target as the root of the tree. Once such a trajectory is known, the robot can use it to gesture along the absolute line as one of the feedback methods, as well as searching for a valid solution, as explained in [28]. This algorithm was always run during robot use, regardless of the feedback method.

The results of the completion time and the TLX combined scores in [37] indicate that all forms of feedback (augmented reality, virtual reality, mounted screen and gesturing) are shown to improve performance compared to the handheld wand. Although the gesturing mode itself does not show as great a performance difference as the visual methods, it does show an improvement nonetheless, justifying it as a viable option. As such, feedback through gesturing was the chosen option for this project, given its feasibility in the time-frame compared to the alternatives. The visual feedback methods themselves exhibited little difference between them. These strong results demonstrate the effectiveness of the robot in assisted tasks, outlining its potential for use in the tool space.

2.5.4 Added Value

TALK ABOUT ADDED VALUE OF DESIGN AND EXPERIMENT IN GENERAL. IE DEMONSTRATES THE DESIGN PROCESS THOROUGHLY FOR THIS TYPE OF FLEXIBLE TUBE CABLE DRIVEN ARM, ALLOWING OTHERS TO FOLLOW SUIT. ALSO SHOWS HOW THE FLEXIBLE TUBE IDEA IN GENERAL WORKS, ADVANTAGES AND DISADVANTAGES AS THIS HAS NEVER BEEN TESTED BEFORE. THEN OF COURSE ANY EXPERIMENTAL DETAILS. FURTHERS THE KNOWLEDGE OF HANDHELD ROBOTICS IN GENERAL AND ISSUES, AREAS OF OPPORTUNITY IT MAY FACE, PARTICULARLY FOR A CHEAPER, SHORTER TERM DESIGN THAN THE FULL 6 DOF MODEL

The objectives of the current project entail interacting with physical objects in a 3D representative environment, which differs from the physical interaction task performed in [9]. The tiling task, although in reality involved relocating 3D objects, can be represented by a 2D environment. The task

knowledge did not involve the depth of the tiles, and therefore they could be assumed to be a perfect disk with zero depth. The user would lower the robot until the discs were picked up or placed, whereas the two dimensions the robot focused on were in the x and y axes; lining up the end effector with the centre of the chequerboard positions.

Alternatively, this project aims to evaluate tasks performed in a 3D rich environment, interacting with objects of known width, height and depth, which must be placed accordingly to simulate a simplified construction environment. A different end effector will be designed, and although it may possess similarities to that used in the tiling task (electromagnet based) it will require an alternative mechanism to consider true 3D space. That is, the alignment of the objects will be important so the end effector must be such that different angles of approach will not cause alignment issues. Moreover, the work in [9] focused more on generic tasks that would require no specific skill and be of little difficulty to the average person. As this project is investigating a more definitive scenario (building environment), the robot used in the new tasks will likely possess task knowledge and capabilities exceeding those of an unskilled worker. This coincides with the idea of enabling complex task completion by those without conventional training in that area, such as crowd-sourced building projects.

The spatial guidance experiment in [37] has yielded useful information regarding the development of handheld robotics in general, considering the importance of effectively conveying preferences to the user. For the current project, the guidance provided to the robot to the user will certainly play a big part in performing the experiment. This is especially so if mental attributes are added to the task to make it harder for the user to complete without assistance. Thus, if the tasks contain elements where the target objects are not immediately visible to the user, one of the three visual feedback methods demonstrated in this paper can be used with confidence. The use of the path planning algorithm in [37] to come up with a suitable 5-DoF pose will also be a key component of the new experiment.

2.5.5 Task Load Index (TLX) and Perceived Effort

When investigating task effectiveness, it is essential to quantify the perceived workload of the performer. This is especially true when researching methods that may improve the task performance. The established means of this quantification is the use of the NASA TLX (Task Load Index). A multi-dimensional scaling was introduced in [39] under the belief that subjective workload is in fact measurable, and has since become widely accepted. Subjective responses can thus be reliably evaluated to meaningful ratings. The six key rating scales associated with the index are mental demand, physical demand, temporal demand, performance, effort and frustration level.

This TLX methodology was employed on each of the experiments that have been carried out for handheld robotics thus far [9] [37], yielding valuable information regarding the subjective workload of cooperative tasks with the robot. As such, it is logical that a similar technique will be used on the new experiment. An ethics approval agreement is already in place that will span the duration of the project. This will allow the undertaking of studies with human participants, concluding with them filling out a TLX form.

3 Summary

A review of existing literature was carried out in order to better understand the area in which this project aims to contribute to. Most notably, the current work regarding handheld robotics at the University of Bristol was looked into. This is especially relevant as it formed the groundwork on which this project will be based, providing crucial details of the apparatus that will be used. Particular focus was given to the design process which has led to the robot's current state and the evaluations of its performance and cooperation.

Coinciding with this is the kinematics information on which the current robot is based. It was necessary to understand this as the approach used for the robot control will need to be exploited to perform any task scenarios which make up the evaluation. The recent study on the methods of feedback for the robot in [37] has enabled confidence when choosing how best to convey spatial information to the user. This is of obvious importance, as the robot will have to interact with the user to indicate where it should next be located during the task, regardless of the specifics of the experiment. It may also become more pertinent throughout the development, if objects are out of line of sight.

A brief review was performed of the field of human in-the-loop operation. Despite being mostly based around external robots that require human input, it still provides useful insight into the way that humans interact with shared robotic tasks. An area of interest that was found dealt with 'sliding autonomy' in which the degree of autonomy the robot exhibits can be fine tuned in a system of transfer of control. This led to improvements in task performance by combining the efficiency of the autonomous agent and the reliability of the human perspective, aiming to determine an optimal balance. This may be worth investigating in the future as a means of potentially locating a more efficient degree of autonomy for certain tasks.

Another approach taken dealt with the weighting of inputs, in which both the human and the robot provided an input, and the singular operational result was formed of a combination of the two. Although this may not be applicable to precise operations, the idea of weighing two inputs based on a current situation may prove beneficial in selecting the operation. For instance a human user could indicate somehow that the situation has changed and the robot must carry out a certain operation, regardless of the task specification.

Furthermore, a brief overview was provided for robots in construction related tasks. This proved to be rewarding in that papers were found that provide a simplified construction environment for testing, albeit on a much smaller scale. The idea of passive structure blocks which easily fit together will most likely be employed on the design of the experiment, and the methods in which the blocks were produced (foam cutting around a template) will also be looked at in detail as a cheap and effective solution. A number of alternative designs were found for such self aligning assembly blocks, providing a basis of inspiration for the new experiment.

4 Handheld Robot Design

4.1 Flexible Robotic Arms

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4.2 Physical Design of Four Degrees-Of-Freedom Device

- The initial design took direction based on the use of four Hi-Tec MS7980 servomotors to provide a degree of freedom each, driving the arm via some form of lightweight cable.
- A flexible foam tube was provided to act as the arm of the robot with outer diameter 42mm and inner diameter 19mm
- Early design was based around a single solid part, where two servos were mounted at the back and another two slightly offset. A tubular section was also present to attach the foam arm. However, after attempting to 3D print this, the quality was severely hindered by the need for excessive scaffolding to support the overhanging elements of the design
- As a result, the base was designed in two halves, with a separate piece for the tubular mount for the arm. These three parts could then be printed separately in an orientation that didn't result in any overhanging sections. As such, a much higher 3D print quality could be achieved.
- At this point, alternate shapes were explored based around this fundamental design, with compactness being a priority. One such design would entail a cubic section, where all servomotors would be at the same level, reducing the required moment arm to hold the device
- The cable management was then accounted for in order to drive the arm. Tubular linkages are attached to the arm with holes to connect the cables to the servomotors. When connecting the cables from the servomotors to the tubular linkages, the orientation of the cables in 3D space must be considered in order to determine how the arm will move in response to servomotor angle changes. Initially, the cables were planned to be connected in a straight line direct from the servomotor to the arm linkages, eliminating the need to account for friction
- INSERT IMAGE OF THE CAD OF THE CUBIC DESIGN WITH CONNECTING LINES
- However, by connecting the cables directly, in a manner with a relative angle between the axis of the arm, the motion of the arm in response to a servo angle change is unpredictable, and more importantly, a single servo motion may not necessarily constrain the arm to motion in a single degree of freedom. That is, the arm will move in one direction initially, resulting in a different angle between the cable and the arm, and further servo motion will cause movement along the axis of the cable (in a different direction to the initial motion).
- As such, it was decided that a configuration was necessary that routed the cables so that each servo acts solely on a single degree of freedom, with the cables running parallel to the arm. This of course required changing the direction of the cables by passing them through guiding holes, accepting friction where they make contact with the edges of the holes.
- Further iterations of the dimensions in the design were explored in order to find a configuration in which the cables of the rear pair of servos did not interfere at all with those of the front pair. Holes were implemented on the central tube holder pillar so that the cables could pass through in an orientation normal to the servo arms of the rear (so the driving force on the servo arm would be tangent to the radius of the servo arm motion), meaning that motion from the servo would result in the maximum movement of the arm. If it was directed non-tangentially, motion from the servo would have less of an effect on the motion of the arm.
- PUT IN IMAGE OF SERVO ARM WITH CABLE ATTACHED TANGENTIALLY AND MOTION TO SHOW LARGE CORRESPONDING MOTION, AND THEN SIMILAR FOR ATTACHED AT AN ANGLE SHOWING LESS MOTION.
- A similar approach was taken to guiding the cables of the front two servos by passing them through holes in the base mounting for the tube holder
- Once the cables have been guided in such a way that they are perpendicular to the servo arms, they then required another set of holes in order to guide them along the axis of the arm. Therefore

a faceplate with four guiding holes was fitted to the front of the housing.

- A low friction fishing wire was thus selected as the cable material [INSERT REFERENCE TO PRODUCT??], and was found to move freely against the 3D printed PLA plastic with little signs of wear or grazing
- An initial prototype was printed with relatively low percentage infill in order to test the concept. The medium sized servo arms that were supplied with the servo were used, feeding the cables through the set of holes along each side of the arm to lock it in place. Exiting the cable at the outermost hole gave an effective diameter (which acts on the cable) of 45mm, whereas the overall diameter of the servo arm was 52mm. Initially, this prototype was ran with all four servos connected to the respective arm linkages, achieving correct motion. This proved that the friction of the cables was not a cause for concern, and the routing of the cables was successful. However, the range of motion of the front two servos were limited by potential collision with the base mounting for the central tube holder, only affording 35 degrees either side of the neutral position. (the max range of the servos is 70 degrees either side but this should have been mentioned already.)
- Another issue found was the inconsistent rate of motion in a given degree of freedom with increasing servo angle, this is for similar reasons as explained earlier in FIGURE X [THE ONE SHOWING THE CABLE ATTACHED TANGENTIALLY TO THE SERVO ARM]. At larger angles, a similar movement of the servo arm results in a smaller response motion from the arm. Hence the effect is an overall reduced range of motion of the arm as high servo angles have minimal effect on the motion as they don't pull the cable further. The other key problem observed was that on the opposite side to the pulling action of the arm, the cable became slack. This is because as the arm flexes, the extra cable available afforded by the forward motion of this end of the servo arm [USE A DIFFERENT WORD TO ARM] is not made up by the increase in arc length on that side of the robot arm as the cable itself does not follow the curve, but takes the direct path between the two linkages, a distance which has actually become shorter as it will always be a maximum when the arm is straight [THE CABLE IS FED THROUGH THE MIDDLE LINKAGE ON WAY TO THE TOP OF COURSE, MENTION THIS EARLIER], [ALSO PUT IMAGE OF THIS EFFECT AND MAYBE REFERENCE AUSTIN PAPER AS SIMILAR EFFECT WAS OBSERVED]
- The latter problem was addressed by adding another pair of linkages that merely act to guide the cable, and are not driven unlike the other two. This reduces the relative difference of the distance the cable has to cover in normal and flexed states, thus reducing the amount of slack occurring.
- Pulleys used to address other problem....
- Dimensions changed to afford more room to front servo to allow it to move full range of servo motion.....

4.3 Controlling The Device

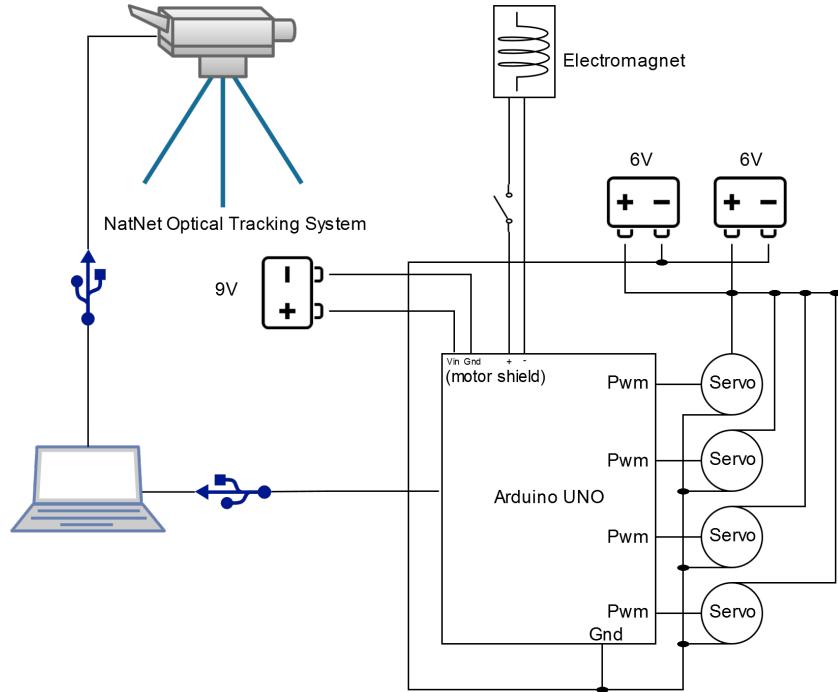


Figure 4.1: Schematic showing how the controlling equipment interfaces with the pc and the robot components.

5 Developed Program

5.1 Interfacing With NatNet Optical Tracking System

In order to determine where in three dimensional space the robot is, as well as the relative distances between itself and other points, it is necessary to track physical objects. This was achieved using a single optical tracking device, the OptiTrack V120 Duo [40]. The device features two cameras in order to provide depth perception, and works as a stand-alone device unlike the syncing of several distinct cameras as used in [9] and [37]. The accompanying software library however, the ‘NatNet SDK’, as well as the actual tracking software, ‘OptiTrack Motive’, are the same. This library was provided to run in a .NET framework environment, hence the decision to use C# as the primary development language.

The device recognises the supplied retro-reflective markers, and is able to provide six degrees of freedom tracking. It is important to note that the coordinate system used by the software (the global axes) originates from the positioning of the Duo tracking device itself, as illustrated in Figure 5.1. Single markers can be identified in terms of their X, Y and Z positions, and can be grouped together to form ‘rigid bodies’. These bodies, requiring a minimum of three markers, have a designated ‘centre’ that can be used to refer to the position of the body as a whole, and provide a pivot about which rotation can be expressed. The rotation is inherently given as a quaternion, which can be converted using built-in library functions to give euler angles in the forms of roll, pitch and yaw (rotations about the global X, Y and Z axes respectively).

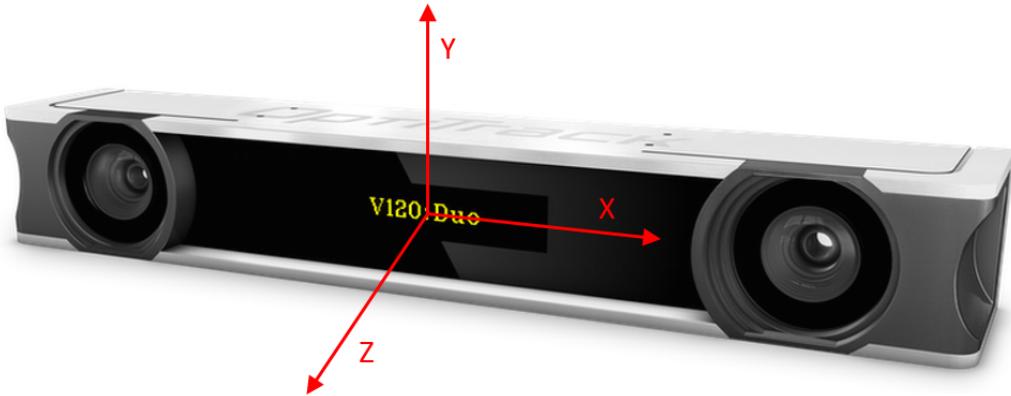


Figure 5.1: Axes as used by the OptiTrack Motive software, these global axes originate from the centre of the device, image adapted from [40]

The rigid bodies must be configured within the Motive program, selecting which markers comprise them and giving them a name. Moreover, individual markers can also be identified as requiring to be tracked. Hence a rigid body will be configured for each of the base and the tip as a bare minimum. With the Motive program running in the background, the NatNet library can be used to extract tracking information from the program. This is done by identifying the program as a server application and connecting it with the local ip address. Once established, and the fundamental NatNet client object is instantiated, the event based data handling is started. This refers to the server application (Motive) triggering an event with a corresponding callback every time a new frame of motion capture data is ready, which is passed into the callback to be processed as desired. This frame of data contains all of the tracking information that Motive is set up to capture. For example, a list of all the rigid bodies, which takes the form of objects storing the name, positional values, rotation (quaternion) and a contained list of individual markers with their own information.

Therefore, rigid body objects can be set up to represent the robot base and tip which, every time a new frame is sent from the program via the callback function, can be updated so they represent the current tracking information. Similarly, other variables pertaining to these bodies, such as the relative distance between tip and base, can be set up and updated as part of the callback function once the

rigid bodies themselves have been updated. As a more practical alternative, a separate thread can be set up so that, although a global variable representing the current frame of motion capture data is updated as soon as a new frame is ready, the rigid body objects or other relevant variables can be updated when desired (after a set time interval) by fetching information from the current frame of data object at any point. Of course, when doing so, it is important to put a read-lock on a shared object so that the frame can be read and data extracted safely without the event based thread updating any of the data mid-read.

Once the required tracking data has been extracted, it can then be manipulated to create relative target points to the base depending on the situation. These can then be used as inputs to the regression function in order to solve for the motor angles. The tracking data at certain points is also written to an XML file during the calibration procedure, as mentioned in SECTION XXXXX.

5.2 Kernel Regression Method

Due to the nature of the design of the robot, particularly the flexible arm in which motion from one of the core arm linkages would affect the other, solving for the kinematics was non-trivial. The arm itself exhibits non-linear characteristics when flexing, rendering established forms of kinematic modelling unsuitable. Thus, an approach was taken similar to that in [9], which is to be expected due to the parallels between the two designs; the arms are of different size and material but are both non-linear and face this problem. The solution, as covered in [9] and again in more detail in [41] is to use a kernel regression method.

The method specifically is known as the Nadayara-Watson kernel regression [42]. It operates on the principle of using known data points to interpolate between, and yield a solution from an input that is not necessarily present in the known data. The function itself is given below in (1).

$$\bar{y}(\bar{x}) = \frac{\sum_{i=1}^n y_i K\left(\frac{x-x_i}{\alpha}\right)}{\sum_{i=1}^n K\left(\frac{x-x_i}{\alpha}\right)} \quad (1)$$

Where K represents a kernel function, α the bandwidth, x the input and y_i and x_i the values of the stored data for the output and input values respectively. The bandwidth, as expected, will have a large effect on the output of the function. This could best be represented by a plot of the output of the function for a range of inputs at a given bandwidth. Of course, due to the number of dimensions in this case (three for the position and four for the motor angles), it is not easy to give a visual representation. For a smooth plot, covering the areas not covered by the stored points but missing the actual stored point values, a higher bandwidth would be expected. However, a lower bandwidth would result in a more jagged plot, virtually intersecting all of the stored points. For very high degrees of non-linearity, a low bandwidth would typically be the most suitable, coming close to the known points.

FIGURE ILLUSTRATING THIS EFFECT, USING 2D GRAPH AS IN AUSTIN THESIS???XXXXX

Given the way this regression works, all of the recorded points will have an effect on the outcome. However, it is desirable for the recorded points close to the input point to have a greater effect on the result than those further away, meaning smaller values of $x - x_i$ will make less of a contribution to the \bar{y} value. Therefore, to cater to this requirement, the chosen kernel function, K , was a Gaussian kernel function, as given in (2), due to its bell-curve shape and how input values close to zero have a larger output effect.

$$K(u) = \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}u^2} \quad (2)$$

With the given kernel regression function, the corresponding output values could be found for various different types of input values. The function itself does not depend on the number of dimensions, and

hence the y and x values can be of any vector size, as long as they are consistent and each input type value is paired with a corresponding output type value within the recorded data. This is because the value used as an argument for the kernel function depends on an absolute difference between the two vectors, which will be a single number regardless of their size, giving an overall output value of the correct number of dimensions. In the given scenario of the robot and the data that is recorded, the matching output types are shown for input types in Table 1.

Table 1: List of corresponding output vector types for a given input vector

Input vector	Output vector
motor angles \bar{m}	relative tip position \bar{p}
relative tip position \bar{p}	motor angles \bar{m}
relative tip orientation \hat{d}	motor angles \bar{m}
\bar{p} and \hat{d}	motor angles \bar{m}

This list of input and output types is similar to that observed for the table input/outputs given in [9], as expected due to the similar methods. However, in this instance, only the tip position is returned when given the motor angles, as this project is of much smaller time scale and the orientation is less critical to the required steps. Once the regression method itself was planned and tested with some sample data, it was necessary to acquire real calibration data in order for it to work and to fine tune the bandwidth appropriately.

5.2.1 Calibration Procedure

DON'T FORGET IMAGES OF ROBOT IN CLAMP FOR VISUALIZATION XXXXX

The procedure for obtaining the calibration data involves sending a range of different motor angle configurations to the servomotors, and recording the resulting tracking information (relative tip-to-base position and orientation). Each data point then comprised all of these values, and was written to disk for re-use in the future when using the regression function. When recording the data, it is imperative that the robot base remains stationary, so each data point is consistent with the same base position and orientation. As mentioned in Section INTERFACING WITH TRACKING XXXXX, the base is said to have zero orientation about any of the global axes if it is in the orientation that it was when it was first initialised in the Motive software (as a rigid body). Hence the base was secured in a vice and then initialised in Motive (along with the tip) before beginning the procedure. This set up is illustrated in Figure XXXXX, with the identification of the base and tip in the Motive software shown in Figure XXXXX. The physical orientation of the base during calibration is of little importance (as long as all points are always visible to the tracking device), as the targets will be relative to the base and converted to its orientation anyway, as covered in more detail in Section XXXX.

INSERT IMAGES XXXXXX

The calibration sampling is then performed at a chosen resolution (number of degrees interval) by stepping through the full range of motion for one servo for each step of a ‘parent’ servo. Thus if each servo (apart from the last in line) has a ‘child’ servo, the motions are performed recursively, with the servo stepping once, then its child covering the full range, with its grandchild then covering the full range for each step of its child and so on. Hence a recursive function was written as shown in Figure XXXXX INSERT CODE OF CALIBRATE FUNCTION AND PUT IN BODY OR APPENDIX?? XXXXX. The function is written such that after each step of a motor, its child motor then covers the range of motion in the opposite direction to its previous one. This is to reduce the number of times a motor has to move all the way back to the start of its range before stepping again, making the motion smoother in general and reducing calibration time.

INSERT CODE IMAGE OR PUT IN APPENDIX XXXXX

5.2.2 Calibration Results

In order to find an appropriate bandwidth for the regression function, it was first necessary to obtain calibration data through sampling. Once enough data is available to perform the regression, the function can be tested against other sampled data, not covered in the initial calibration data, referred to as the ‘test data’. By comparing the output of the function for different bandwidth values against the values of the test data, an absolute positional error can be determined.

The state-space of the arm was initially sampled at a chosen resolution of five points over a total range of 60° per servo, giving a total sample number of $5^4 = 625$ calibration points. This gave a suitable range of motion, with an improvement of 20° over the device in [9]. Higher ranges were initially used without any long term issues, although temporary deformation of the foam arm meant that the cables need to be re-tensioned, which was especially time consuming. A one second pause was issued after each calibration step to ensure the tip was steady before the position was logged. Hence this calibration took just over 10 minutes. Of course, higher accuracy could be obtained through more samples, with an increase in time, but this was kept short due to the fact that it needed to be repeated a lot with small design changes. Also, a meaningful increase in accuracy would require numerous extra sample points, quickly increasing the required time to well over an hour, which introduces other problems such as battery issues and gradual deformation.

The number of test data points was considerably smaller, with the primary aim of just collecting data for motor angle vectors that were not covered in the initial sampling. Hence $3^4 = 81$ samples were obtained, with starting position and step selected to deliberately avoid any calibration samples. For the comparison between the function output and the test data, the motor angles were to be the input, thus comparing the output positional values with the recorded points for the same motor angles. This ‘forward’ approach was used for two reasons. The first is because the errors could be compared in terms of millimetres in space, which is more meaningful than servo degrees. The other reason is due to the fact that a given 3D position could potentially be achieved (or very close to) by more than one configuration of servo angles, whereas there is only one matching position for a given angle configuration.

Due to the nature of the kernel function, large differential values (the $x - x_i$ term) can cause issues. The input to the kernel function becomes larger when divided by the bandwidth (typically less than one), meaning when this value is squared, negated and made an exponent, a very small number is the result. This becomes particularly problematic when the value is so small that the number of bits required to represent it exceed those available in the ‘double’ data type, giving a result of zero out of the Gaussian kernel function. This then can then cause the regression itself to fail as the denominator becomes zero, giving NaN values as output as opposed to any meaningful values. Due to the typically observed differences in degrees between the servo angles of the input and the sample points, this issue was present. As such, the value of servo angles need to be scaled so that the difference between any values is guaranteed to be in the range of $-1 < x - x_i < 1$. This is known as feature scaling, or min-max normalization [43], and is shown in (3), where x' is the new value after the original value, x , was scaled. This was performed for all servo angles during the regression function, and converted back after output to get results in degrees.

$$x' = \frac{x - x_{min}}{x_{max} - x_{min}} \quad (3)$$

The function was thus used for each of the 81 angle configurations covered by the test data, giving a mean positional error (mm) between the output and test data values. This was then repeated for a range of bandwidth values and plotted to determine the most suitable value to minimise error, as illustrated in Figure 5.2.

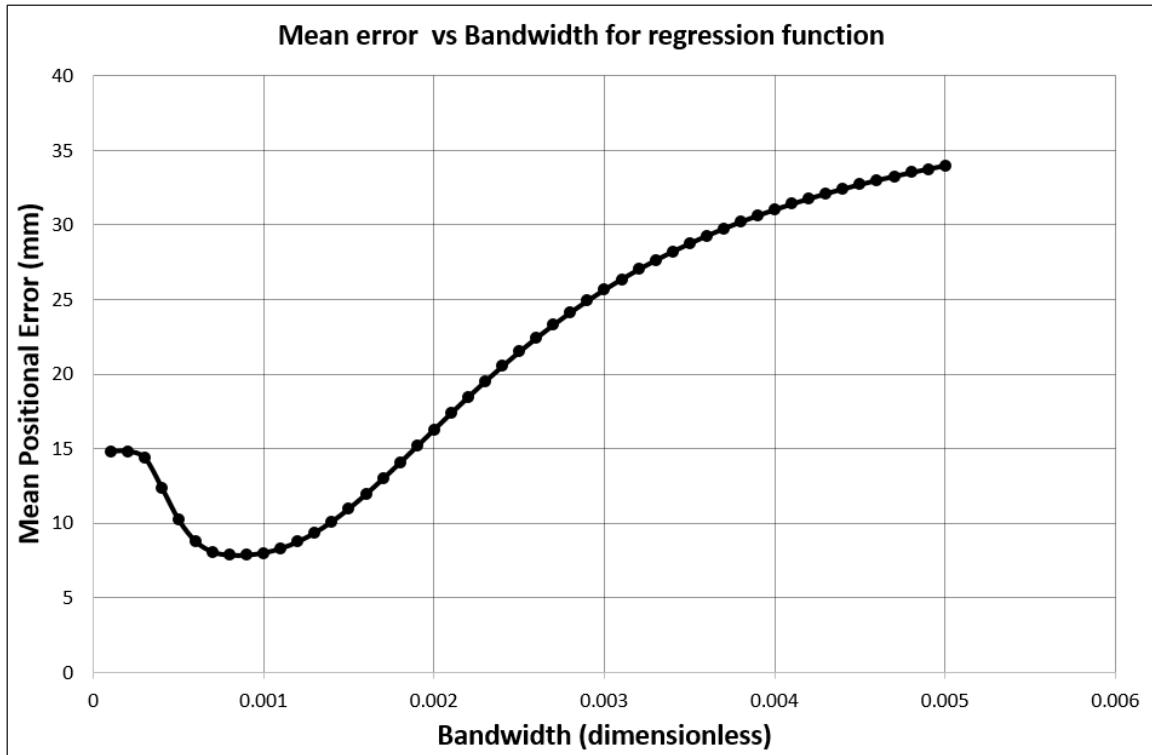


Figure 5.2: Mean positional error at various bandwidths for the output of the regression function compared with the corresponding real positional data

Low bandwidth shows high non-linearity as expected due to arm material etc

5.3 Algorithm for Responding to Three-Dimensional State Targets

Once the robot has been calibrated in such a way that there exists a suitable range of data points in which to perform the regression function, along with the selection of an appropriate bandwidth, the device may then be responsive to 3D state commands. The process of responding to commands is fundamentally built on inputting a positional vector, an orientation vector or a combined vector into the regression function in order to ascertain the required motor angles. This input vector is of course in regard to the tip relative to the base, as is the corresponding calibration data. The resulting motor angle vector is then issued to the servomotors via the Arduino in a controlled manner.

However, in order for this approach of obtaining servo angles from a given input to be useful in practice, it is first essential to determine the correct input to the regression function. This can depend on a variety of factors relating to the absolute 3D target point and its relation to the base of the robot, such as the current orientation of the base and the relative proximity of the target.

5.3.1 Acquiring Target Position Relative to Orientation of Base

The robot is fixed in place while the optical tracking software is configured to recognise the rigid bodies formed by the retro-reflective markers associated with the base and the tip, as covered in SECTION XXXXXXXX (just say 'is configured'?). At this point, the base is said to have zero orientation about any axis (as this was the point it was first initialised in). The calibration procedure is then of course undertaken while the robot is fixed in this position. Thus, the stored relative tip-to-base calibration data for different servo angle vectors is only valid for the robot in a zero orientation position compared to the starting position. Hence, in practice, when the base has changed orientation about any of the three axes inherent to the optical tracking device, a direct input for a relative tip-to-base position will yield incorrect motor angles to meet this position. Therefore, the target position must be adjusted in order to account for the varying orientation of the base.

The orientation of the base itself is initially stored as a quaternion within the rigid body object from the NatNet library [44]. This is then converted to euler angles using a built-in method to represent the orientation as three angles about the tracking device's X, Y and Z axes respectively, of which the order is important. It is thus necessary convert a real target point to an equivalent point relative to the original orientation of the base (zero orientation), so that the regression function and stored calibration data can be used to yield angles that result in the correct relative position in real space. This is accomplished using a rotation matrix that accounts for the rotation ordering of the base. Given that the base orientation is represented by rotation about the X, then Y, then Z axes, the converted point will be attained by taking the actual target point relative to the base and reversing these rotation operations; that is, the negatives of the base rotation values about the Z, then Y, then X axes. The rotation matrix in (4) is thus used to account for this axes of rotation order.

$$R_x R_y R_z = \begin{bmatrix} \cos \beta \cos \gamma & -\cos \beta \sin \gamma & \sin \beta \\ \cos \alpha \sin \gamma + \sin \alpha \sin \beta \cos \gamma & \cos \alpha \cos \gamma - \sin \alpha \sin \beta \sin \gamma & -\sin \alpha \cos \beta \\ \sin \alpha \sin \gamma - \cos \alpha \sin \beta \cos \gamma & \sin \alpha \cos \gamma + \cos \alpha \sin \beta \sin \gamma & \cos \alpha \cos \beta \end{bmatrix} \quad (4)$$

This equation gives the rotation matrix for rotation about Z, then Y, then X axes. The symbols α , β and γ represent rotations about the X, Y and Z axes respectively, adapted from [45]. A simplified example of solving from a target point to get the appropriate relative positional vector as an input to the regression function is illustrated in Figure 5.3. Note that the target position in the image is not to scale and the coordinate system used in reality is determined by the optical tracking device position, not aligned with the longitudinal axis of the robot.

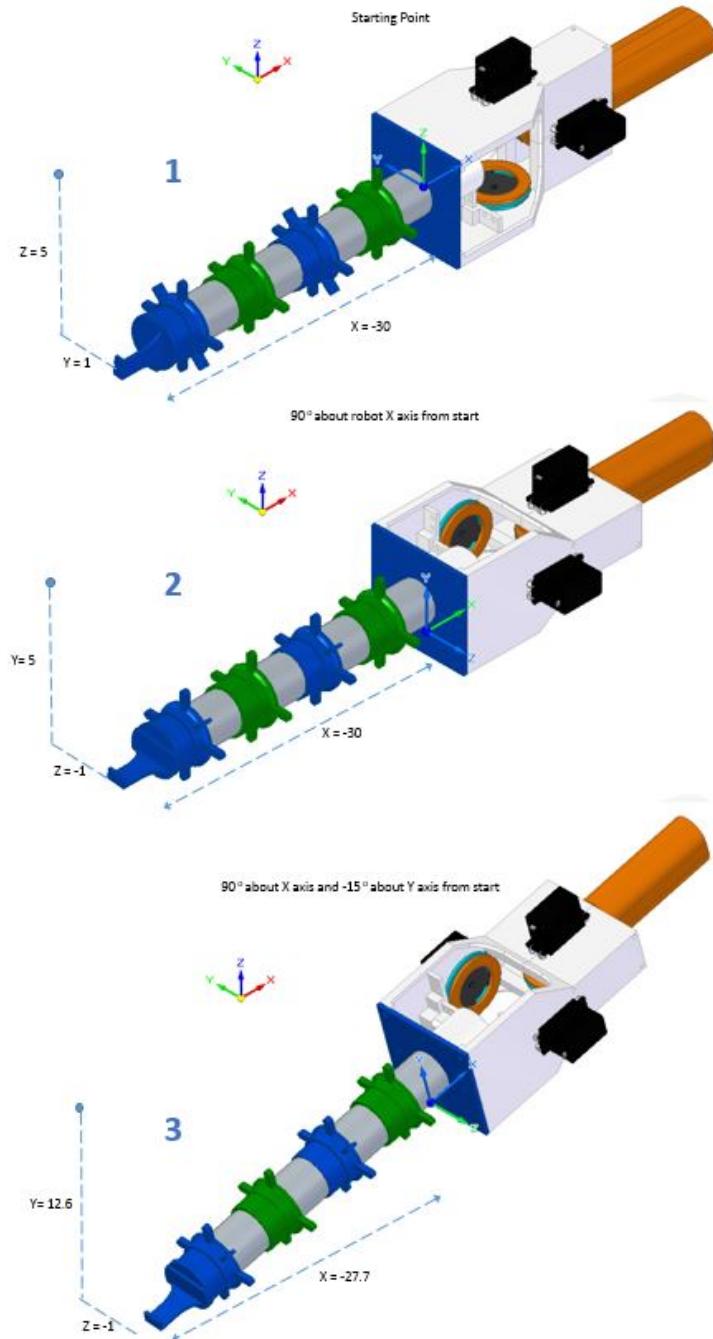


Figure 5.3: Example showing how the position and orientation of a point changes with base rotation

1. The robot is in its original starting orientation, and can be said to have zero rotation about any of the axes. There is no relative difference between the axes of the robot and the global axes (shown in the top left of each image). The target point has an actual relative distance to the robot base of -30, 1 and 5 units respectively for the X, Y and Z dimensions. As there is no relative axes difference, the relative distance in relation to the robot base axes is identical in all three dimensions. As such, in this situation, the actual relative distance vector could be inserted into the regression function as is.
2. At the second stage, the robot has been rotated 90° about the global x axis. The local axis of the robot base is therefore at a relative orientation difference of the same amount. The physical relative distance of the target point to the base is unchanged, but the target point relative to the local axes of the base is different, with the local Y and Z distances being swapped around, and the latter being negated.
3. The final image shows the outcome if the relative rotation of the base was a 90° rotation about the global X axis (as in the previous image) followed by a -15° rotation about the global Y axis.

The physical relative distance of the target point to the base is again unchanged in the global axis as the base itself has not translated in any direction. Again, the relative dimensional distances in terms of the local axes of the base are altered further, with relative values now of -27.7, 12.6 and -1 respectively for X, Y and Z. This is the vector that must be inserted into the regression function, to attain motor angles that result in a 3D point which, relative to the local axes of the base, is equal to the physical target point relative to the global and starting axes. This conversion of the target point to the local axes of the base is performed via the aforementioned steps with the negation of base angles and the rotation matrix of 4.

5.3.2 Gesturing Algorithm for Distant Targets

In order to be cooperative and interactive in practice, a handheld robot must be able to indicate positional preferences to the user. This may entail where it must be located in order to carry out a given robotic operation, or even just the next point the user must deal with for task progression. This process is referred to as gesturing from the robot to the user towards 3D locations. In this scenario, the selected method for gesturing is comparable to typical ‘pointing’ operations, similar to finger pointing, which is beneficial in terms of simplicity of use. Pointing is intuitive to the user as humans do this naturally to direct the attention of others to areas of interest, and is considered inherently linked to speech and communication [46]. Therefore confusion should be avoided in the user knowing where the robot desires to be relocated.

In order to gesture effectively to the user, the 3D target points must be categorised as achievable or not from the outset before deciding the next step. Initially, a point can be said to be immediately unachievable if the relative distance between itself and the base of the robot is greater than (or a certain amount less than) the typical tip-to-base distance in a straight-arm configuration. This tip-to-base distance can be used as the radius of a sphere which governs whether the points can be immediately disregarded as attainable, and requires gesturing, or whether further checking is required to assess whether it is appropriate to attempt to match this point directly. This concept is illustrated in Figure 5.4.

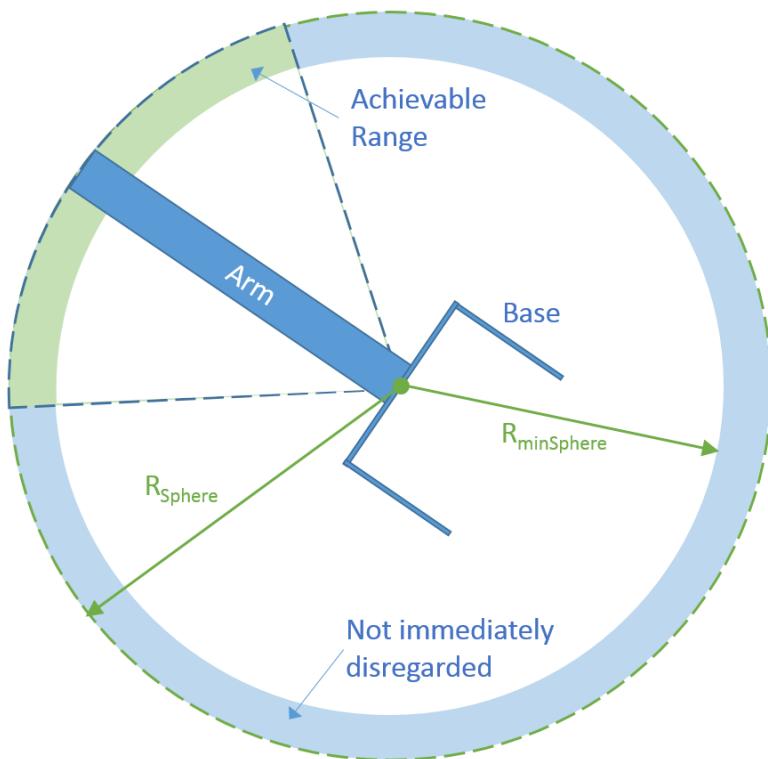


Figure 5.4: Concept of tip-to-base sphere marking immediately disregarded zone, and initial achievable range zone that requires further checking

If the point lies at an absolute distance that falls within the range between the radius of the sphere and another minimum radius, $R_{minSphere}$, the point is not disregarded as being unachievable straight away. This minimum radius is determined by the smallest absolute tip-to-base difference from the stored calibration data (3D Pythagoras). However, if it lies outside this range, a new target point is required that is equivalent to shifting the original along the vector between itself and the base until it falls in this range.

The current target then lies within this intermediate range, and there is a further check to determine whether the coordinates of the target point relative to the base fall outside any of the recorded minimum and maximum values for the X, Y and Z dimensions. If not, then the target may be directly achievable, and the positional vector is inserted straight into the regression function to check if motor angles can be yielded, giving a solution. However in the event of this still not generating a valid solution, or if the point does fall outside any of the min/max ranges, a compromise is required. This entails selecting a new target point from the stored calibration data that is geometrically closest to the target point in question. Thus, if not able to match the point, a new point should be determined to exhibit a suitable gesture motion. The typical process for responding to a far-off point is illustrated in Figure 5.5.

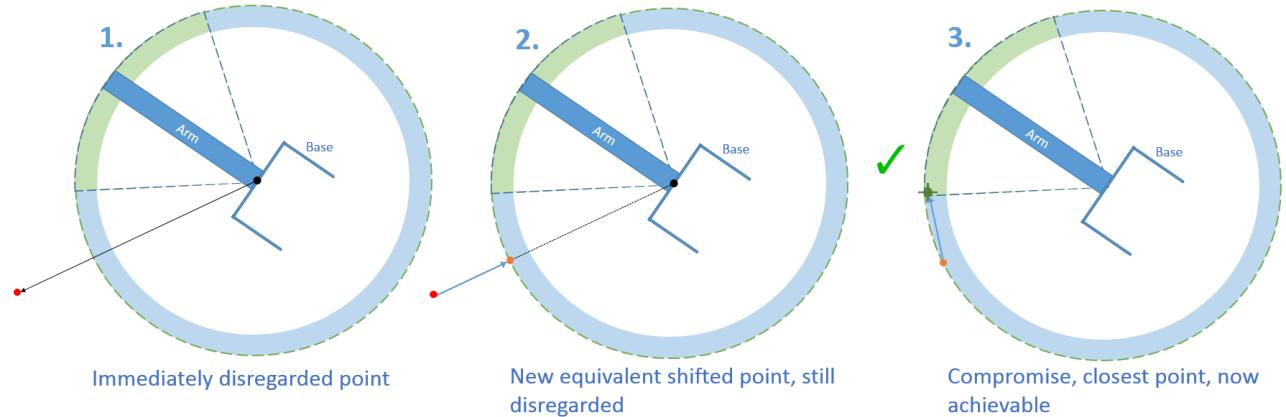


Figure 5.5: Typical steps in obtaining a compromised gesture solution from an initially unachievable point

1. The given target point is immediately disregarded as it lies at an absolute distance above the radius of the sphere.
2. A new target point is formed that is the equivalent of shifting the original point along the line from itself to the base. This is done by determining the required factor to shift the target to lie on the sphere. This factor is $ShiftFactor = R_{Sphere}/absolutedistance$, and each coordinate of the target point is multiplied by this to give a new vector on the sphere.
3. The point is still unachievable as it lies beyond the max/min range of some of the dimensions from the calibration data. The compromise is therefore to find the point in the calibration data that has the closest absolute proximity to this point and use that either as an input to the function of using the stored motor angles for this point (quicker).

With these steps of applying a rotation transformation and determining the best point to aim for relative to the base, there now exists a general algorithm from going from any given target point, expressed as relative to the base, and generating a solution in the form of angles to be applied to the servomotors, moving the arm to the required position. Regardless of the task or application at hand, this algorithm will be central, receiving inputs based on the situation and yielding solutions to be applied. This fundamental step sequence is illustrated in Figure 5.6.

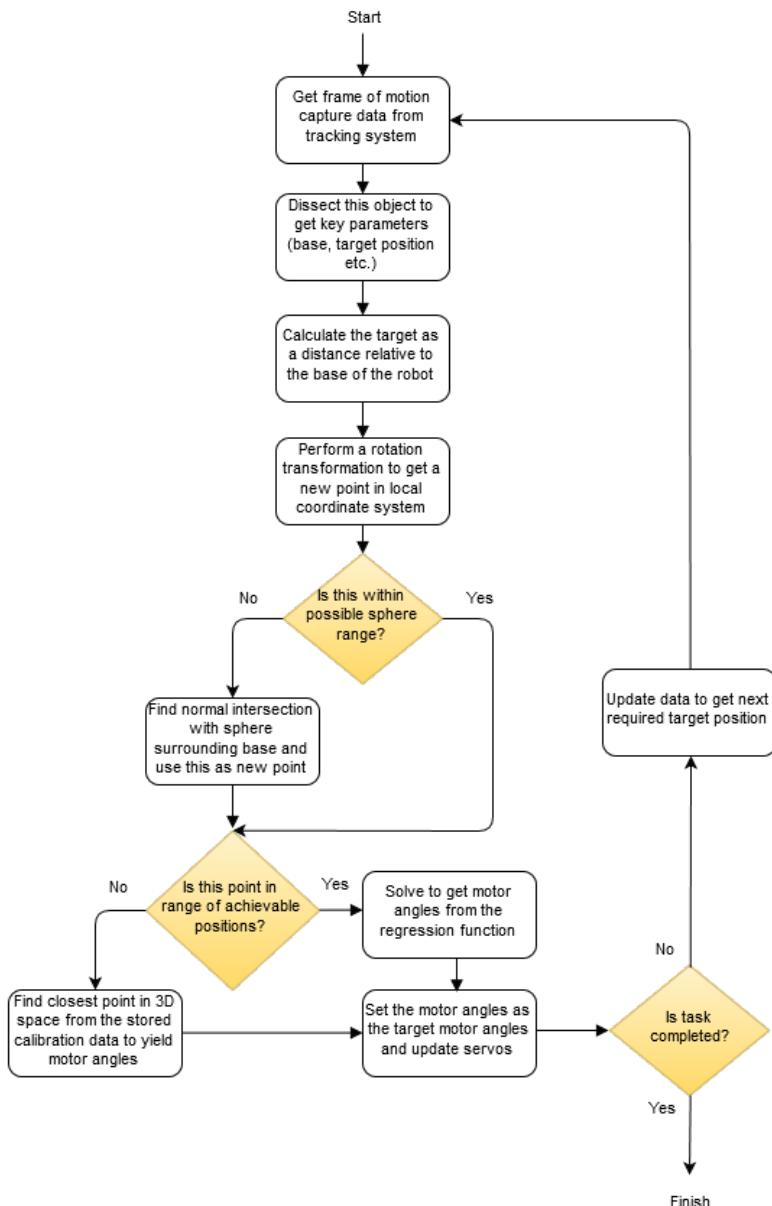


Figure 5.6: General algorithm procedure for taking a 3D target point relative to the base and producing the most suitable solution in the form of angles to be applied to the servomotors

5.3.3 Running with a Live Target in Real Time

RUNNING THE SET MOTOR INSTRUCTION IN A DIFFERENT THREAD IN A LOOP SO TARGET CAN BE UPDATED LIVE AND THE MOTION OF THE SERVOS IS ADJUSTED LIVE SO AS TO NOT FINISH THE PREVIOUS COMMAND AND JUST CONCENTRATE ON CURRENT TARGET.

SMALL PAIR OF FLOW DIAGRAMS TO SHOW ONE THREAD CONSTANTLY UPDATING THE POSITION AND THE OTHER WORKING OUT WHAT THE POSITION SHOULD BE AND CHANGING IT BASED ON TASK PROGRESSION AND TARGET POINT ETC. IN DIFFERENT DASHED BOXES TO SHOW THE DIFFERENT THREADS. THE LATTER WILL JUST SAY "POINT DETERMINATION ALGORITHM" AND WILL BE A COMPRESSED VERSION OF THE LARGER FLOW DIAGRAM IN THE PREVIOUS SECTION.

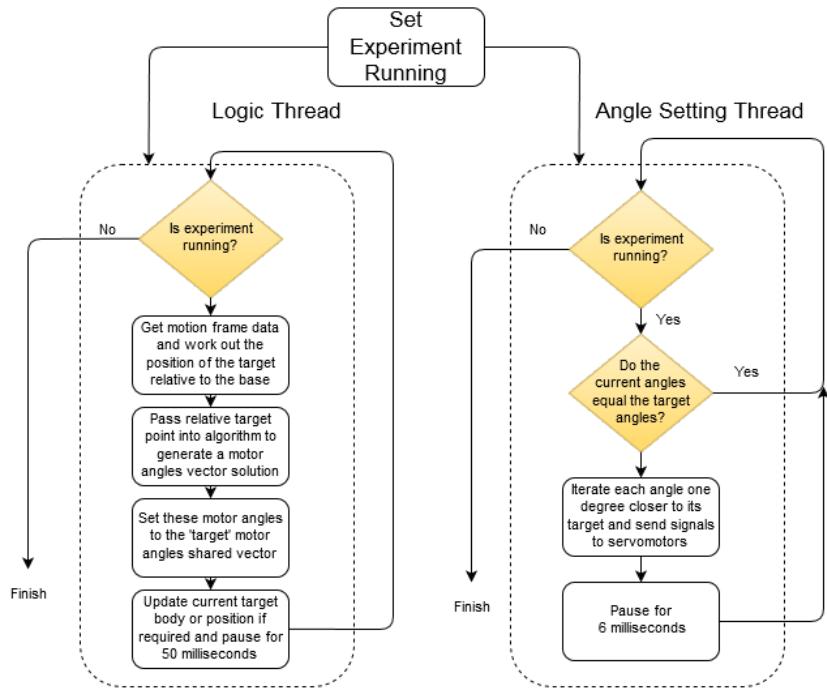


Figure 5.7: Simplified processes taking place in the two different threads, showing the servos are updated after a set interval regardless of the logic

THE IMAGE SHOWS THAT THE TARGET ANGLES CAN CHANGE WHENEVER AND THE ROBOT WILL RESPOND BY JUST MOVING TOWARDS THE NEW TARGET, MAKING IT MORE RESPONSIVE. THIS IS OPPOSED TO HAVING TO COMPLETE ITS ORIGINAL MOTION BEFORE MOVING TO THE NEW TARGET MAKING IT HAVE A LAG. ALSO SAY THE 6MS DELAY IS TO SMOOTHEN THE MOTION AS QUICK JERKS CAN PULL CABLES LOOSE ETC

5.4 Program Design

The developed program for using the robot was built around the fundamentals of the Motive software and the NatNet library. An example was provided that simply sets up a connection between the tracking program and displays the tracked data in a spreadsheet. This example took the form of Windows Form Application, due to the usefulness of a Graphical User Interface (GUI) with buttons and options to start/stop tracking etcetera. Hence the developed application was of the same type, adapting the example tracking code and removing and adding features as necessary for the specific purpose of tracking and controlling the robot. This entailed creating new classes that deal with interfacing with the robot as well as the logic involved with the calibration, target selection and solving processes.

Small tests were also added to the program so that the motion of the robot or acquiring of a solution from the regression for instance could be tested without having to fully connect to the optical tracking etcetera. This would use dummy variables where necessary just to quickly perform tests. The general process of adding a new feature to the program would entail adding a button to the GUI and attaching methods to the callback associated with it. It is often necessary to have this functionality run in a separate thread so as not to hinder the updating of data or the user interface itself (rendering it unresponsive while the operation is carried out). Hence it was essential to ensure thread safety, using a shared syncLock object when accessing shared objects, as well as delegating any user interface changes to the thread responsible for updating it. This simple user interface is illustrated in Figure 5.8.

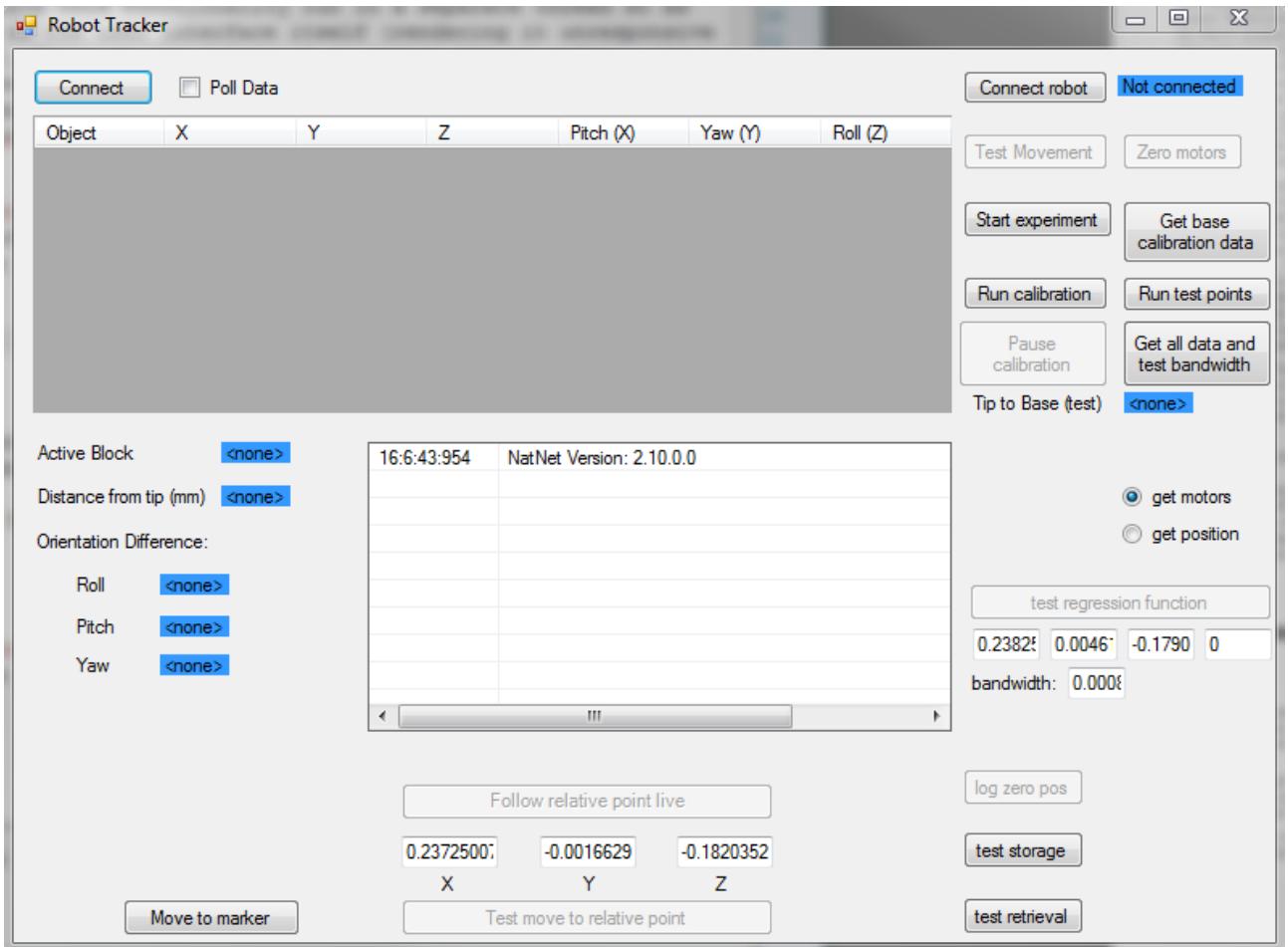


Figure 5.8: User interface of the program for controlling the current status and operations of the robot, as well as for connecting to the tracking system and reading key tracking data.

USE UPDATED SCREENSHOT WHEN ACTUALLY CONNECTED

The upper left and right corners provide buttons for connecting to the Motive optical tracking software and the Arduino for robot control respectively. As the typical procedures regarding controlling the robot depend on analysing tracking data and sending instructions accordingly, both connections must be established before any practical operations can be performed. Hence full connection is required to start an ‘experiment’, which entails creating an Experiment object based on the information provided from both connections. However, as alluded to earlier, a ‘dummy experiment’ will be created without full connection in order to provide some basic functionality so that simple motion and logic tests can be carried out without having to set up and connect to the external infrastructure associated with motion tracking.

Other buttons are provided for both testing and essential processes. The calibration procedure is initiated, stopped and paused (in case of battery failure)for both the main calibration data as well as the test sampling, using the corresponding buttons. There’s also a button for reading stored calibration data into memory, without which the regression function would be impossible. The process mentioned in SECTION XXXXXX for discovering the most appropriate bandwidth value can also be carried out here. There is a live spreadsheet section in order to display the current tracking data being recorded, such as the position and orientation of the various rigid bodies and associated markers. Similarly, labels have been added to display values that have been calculated from this tracking data, such as the absolute distance from tip to base and relative orientation. Moreover, buttons are present for performing the following test scenarios, which were especially useful during development:

- Performing a basic set of movement instructions for testing the servomotors, as well as zeroing them again
- Performing the regression function for an input of either the desired position or the actual motor

angles for user entered values

- Testing the storage and retrieval of calibration data
- Moving to a relative target position, as entered by the user
- Attempting to stay at a relative target position, as entered by the user, regardless of how the base is oriented/positioned
- Moving to a tracked target object which is also in view of the tracking system

The core classes of the program are illustrated in Figure 5.9. Note that only the essential fields and methods that dictate how the program operates are shown, with the boilerplate NatNet methods, the smaller sub-methods and the less critical methods for testing and set up all excluded for simplicity.

1. RobotTracker - The starting class that was built upon from the original NatNet code for interfacing with a tracking system via a simple Windows Form Application. This class is responsible for initiating the connection with the tracking system, as well as dealing with the corresponding data it receives. On top of this, it contains objects for the other main classes and methods that instantiate and utilise them. This is mostly achieved through callback methods associated with the user interface buttons, giving control to the user.
2. Experiment - The class that extracts the key information from the tracking data into specific variables, and performs the logic required for ascertaining the values that need to be passed into the regression function, and performing the regression itself. This entails all of the steps outlined in Section XXXX regarding converting the target orientation and finding a sphere intercetera. The calibration procedure, and the loading of the calibration data, is also performed in this class, along with other test processes.
3. DataPoint - An object to store the motor angle vector and its corresponding relative tip-to-base position and orientation values as recorded by the tracking system. Many of these objects will make up the calibration data, and the values contained within them will be used when the regression function is performed, depending on the type of input/output. Serializable as with CalibrationData.
4. CalibrationData - A collection of DataPoint objects, that can be accessed as if it were an array with an index, adding and removing points as necessary. Of course, this will be serializable as it will need to be written to and read from disk.
5. RobotControl - This class is responsible for physically controlling the robot via its connection to the Arduino which in turn controls the servomotors. A connection is established with the Arduino and values are sent via serial connection, which are subsequently interpreted by the Arduino script and servo signals are applied accordingly. As mentioned in Section XXXXX, the values of the current motor angles are compared with their targets, and iterated one degree closer if required before being written to the servomotors together.

ADD ANY MORE FOR THE NEW CLASSES FOR BODE PLOT AND USER STUDY????XXXXX

RobotTracker : Form	Experiment												
<ul style="list-style-type: none"> - currentFrame: FrameOfMocapData - syncLock: object - UIUpdateThread: Thread - experiment: Experiment - controller: RobotControl 	<ul style="list-style-type: none"> - controller: RobotControl - robotBase: RigidBodyData - robotTip: RigidBodyData - storedCalibrationData: CalibrationData - motorAngles: int[4] - relativeTipPosition: float[3] - bandwidth: float - eulerAnglesRobotBase: float[] - experimentRunning: bool 												
<ul style="list-style-type: none"> + setupConnection(): + updateUI(): + updateData(): + m_NatNet_OnFrameReady(FrameOfMocapData data, NatNetClient client): + attemptRobotConnect(): + requiredObjectsTracked(): bool 	<ul style="list-style-type: none"> + convertTargetOrientation(float[] targetPos): float[] + featureScaling(): + buildRotationMatrix(float[] baseRotation): float[] + update(FrameOfMocapData newCurrentFrame): + NWRegression(float[] input, RegressionInput type, float bandwidth): double[] + sphereIntersectionConvert(float[] relativeTarget): + calibrate(bool testSampling): + liveExperimentThreadLoop(): + getData(string filename): + bandwidthErrorPlot(): + saveDataXML(string filename): 												
<table border="1" style="width: 100%; border-collapse: collapse;"> <tr> <td style="padding: 5px;">[Serializable] CalibrationData : ICollection</td> <td style="padding: 5px;">RobotControl</td> </tr> <tr> <td> <ul style="list-style-type: none"> - tableData: List<DataPoint> - Count: int </td><td> <ul style="list-style-type: none"> - currentPort: SerialPort - connectedToPort: bool - targetMotorAngles: int[4] - currentMotorAngles: int[4] - syncLock: object </td></tr> <tr> <td> <ul style="list-style-type: none"> + Add(DataPoint newPoint): + RemoveAt(int index): + updateData(): </td><td> <ul style="list-style-type: none"> + detectArduinoPort(): bool + zeroMotors(): bool + shareMotorAngles(int[] targetMotorAngles): + setMotorAngles(): + anglesNotEqual(): bool + testMotion(): + updateCurrentMotorAngle(int index, int angle): + isConnected(): bool </td></tr> </table>	[Serializable] CalibrationData : ICollection	RobotControl	<ul style="list-style-type: none"> - tableData: List<DataPoint> - Count: int 	<ul style="list-style-type: none"> - currentPort: SerialPort - connectedToPort: bool - targetMotorAngles: int[4] - currentMotorAngles: int[4] - syncLock: object 	<ul style="list-style-type: none"> + Add(DataPoint newPoint): + RemoveAt(int index): + updateData(): 	<ul style="list-style-type: none"> + detectArduinoPort(): bool + zeroMotors(): bool + shareMotorAngles(int[] targetMotorAngles): + setMotorAngles(): + anglesNotEqual(): bool + testMotion(): + updateCurrentMotorAngle(int index, int angle): + isConnected(): bool 	<table border="1" style="width: 100%; border-collapse: collapse;"> <tr> <td style="padding: 5px;">[Serializable] DataPoint</td> <td style="padding: 5px;"></td> </tr> <tr> <td> <ul style="list-style-type: none"> - motorAngles: float[4] - relativeTipPosition: float[3] - relativeTipOrientation: float[3] </td><td></td></tr> <tr> <td> <ul style="list-style-type: none"> + setMotorAngles(int[] newAngles): + setTipPos(float[] newPos): + setTipOrientation(float[] newOrientation): </td><td></td></tr> </table>	[Serializable] DataPoint		<ul style="list-style-type: none"> - motorAngles: float[4] - relativeTipPosition: float[3] - relativeTipOrientation: float[3] 		<ul style="list-style-type: none"> + setMotorAngles(int[] newAngles): + setTipPos(float[] newPos): + setTipOrientation(float[] newOrientation): 	
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Figure 5.9: Core classes and their associated fields and methods that are central to how the program runs.

6 Physical Experiment

6.1 Setup

6.2 Construction Environment Parallels

6.3 Mental Aspect

6.4 Physical Aspect

7 Results

8 Discussion

9 Conclusion

WERE AIMS AND OBJECTIVES MET, WHAT ELSE WAS FOUND OUT????

10 Further Work and Areas of Improvement

10.1 Physical Design

10.2 Experiment

11 References

- [1] R. Chipalkatty. *Human-In-The-Loop Control For Cooperative Human-Robot Tasks*. PhD thesis, Georgia Institute of Technology, May 2012.
- [2] S.J. Anderson, S.C. Peters, T.E. Pilutti, E.H. Tseng, and K. Iagnemma. Semi-autonomous avoidance of moving hazards for passenger vehicles. In *ASME 2010 Dynamic Systems and Control Conference*, pages 141–148. American Society of Mechanical Engineers, 2010.
- [3] S. Haddadin, A. Albu-Schäffer, and G. Hirzinger. Safety evaluation of physical human-robot interaction via crash-testing.
- [4] EN953. Safety of machinery - guards - general requirements for the design and construction of fixed and movable guards, 1997.
- [5] C. Ott, O. Eiberger, W. Friedl, B. Bauml, U. Hillenbrand, C. Borst, A. Albu-Schaffer, B. Brunner, H. Hirschmuller, S. Kielhofer, R. Konietzschke, M. Suppa, T. Wimbock, F. Zacharias, and G. Hirzinger. A humanoid two-arm system for dexterous manipulation. In *2006 6th IEEE-RAS International Conference on Humanoid Robots*, pages 276–283, Dec 2006.
- [6] F.W. Heger and S. Singh. Sliding autonomy for complex coordinated multi-robot tasks: Analysis & experiments. 2006.
- [7] G. Dorais, R.P. Bonasso, D. Kortenkamp, B. Pell, and D. Schreckenghost. Adjustable autonomy for human-centered autonomous systems.
- [8] M. Tambe, P. Scerri, and D.V. Pynadath. Adjustable autonomy for the real world. *Journal of Artificial Intelligence Research*, 17(1):171–228, 2002.
- [9] A. Gregg-Smith and W.W. Mayol-Cuevas. The design and evaluation of a cooperative handheld robot. In *2015 IEEE International Conference on Robotics and Automation (ICRA)*, pages 1968–1975, May 2015.
- [10] A. Poncela, C. Urdiales, E. J. Perez, and F. Sandoval. A new efficiency-weighted strategy for continuous human/robot cooperation in navigation. *IEEE Transactions on Systems, Man, and Cybernetics - Part A: Systems and Humans*, 39(3):486–500, May 2009.
- [11] C. Lenz, A. Sotzek, T. Röder, H. Radrich, A. Knoll, M. Huber, and S. Glasauer. Human workflow analysis using 3d occupancy grid hand tracking in a human-robot collaboration scenario. In *Intelligent Robots and Systems (IROS), 2011 IEEE/RSJ International Conference on*, pages 3375–3380. IEEE, 2011.
- [12] M. Huber, A. Knoll, T. Brandt, and S. Glasauer. When to assist?-modelling human behaviour for hybrid assembly systems. In *Robotics (ISR), 2010 41st International Symposium on and 2010 6th German Conference on Robotics (ROBOTIK)*, pages 1–6. VDE, 2010.
- [13] P. Lukowicz, J.A. Ward, H. Junker, M. Stäger, G. Tröster, A. Atrash, and T. Starner. Recognizing workshop activity using body worn microphones and accelerometers. In *Pervasive Computing*, pages 18–32. Springer, 2004.
- [14] A. Elfes. Using occupancy grids for mobile robot perception and navigation. *Computer*, 22(6):46–57, 1989.
- [15] M. Buss and M. Beetz. Cotesys—cognition for technical systems. *KI-Künstliche Intelligenz*, 24(4):323–327, 2010.
- [16] J. Werfel, Y. Bar-Yam, D. Rus, and R. Nagpal. Distributed construction by mobile robots with enhanced building blocks. In *Robotics and Automation, 2006. ICRA 2006. Proceedings 2006 IEEE International Conference on*, pages 2787–2794. IEEE, 2006.

- [17] K. Petersen, R. Nagpal, and J. Werfel. Termes: An autonomous robotic system for three-dimensional collective construction. *Proc. Robotics: Science & Systems VII*, 2011.
- [18] D. Mellinger, Q. Lindsey, M. Shomin, and V. Kumar. Design, modeling, estimation and control for aerial grasping and manipulation. In *2011 IEEE/RSJ International Conference on Intelligent Robots and Systems*, pages 2668–2673, Sept 2011.
- [19] R. Cano, C. Pérez, F. Pruano, A. Ollero, and G. Heredia. Mechanical design of a 6-dof aerial manipulator for assembling bar structures using uavs. In *2nd RED-UAS 2013 Workshop on Research, Education and Development of Unmanned Aerial Systems*, 2013.
- [20] P. Latteur, S. Goessens, J. Breton, J. Leplat, Z. Ma, and C. Mueller. Drone-based additive manufacturing of architectural structures.
- [21] R.C. Arkin and R.R. Murphy. Autonomous navigation in a manufacturing environment. *IEEE Transactions on Robotics and Automation*, 6(4):445–454, Aug 1990.
- [22] C.J. Payne and Guang-Zhong Yang. Hand-held medical robots. *Annals of Biomedical Engineering*, 42(8):1594–1605, 2014.
- [23] N. Matsuhira, M. Jinno, T. Miyagawa, T. Sunaoshi, T. Hato, Y. Morikawa, T. Furukawa, S. Ozawa, M. Kitajima, and K. Nakazawa. Development of a functional model for a master–slave combined manipulator for laparoscopic surgery. *Advanced Robotics*, 17(6):523–539, 2003.
- [24] C.J. Payne, H.J. Marcus, and G.Z. Yang. A smart haptic hand-held device for neurosurgical microdissection. *Annals of Biomedical Engineering*, 43(9):2185–2195, 2015.
- [25] W.T. Latt, U.X. Tan, C.Y. Shee, and W.T. Ang. A compact hand-held active physiological tremor compensation instrument. In *2009 IEEE/ASME International Conference on Advanced Intelligent Mechatronics*, pages 711–716, July 2009.
- [26] M. Fischer, M. Schraufstetter, C. Richter, F. Irlinger, and T. C. Lueth. Tremor compensation by use of a mechatronic cup holder. In *2010 4th International Conference on Pervasive Computing Technologies for Healthcare*, pages 1–8, March 2010.
- [27] F. Echtler, F. Sturm, and K. Kindermann. The intelligent welding gun: Augmented reality for experimental vehicle construction. In *Virtual and Augmented Reality Applications in Manufacturing*, pages 333–360. Springer London, London, 2004.
- [28] A. Gregg-Smith and W.W. Mayol-Cuevas. Inverse kinematics and design of a novel 6-dof hand-held robot arm. In *IEEE International Conference on Robotics and Automation (ICRA)*, ICRA Proceedings. IEEE, February 2016.
- [29] A. Horigome, H. Yamada, G. Endo, S. Sen, S. Hirose, and E.F. Fukushima. Development of a coupled tendon-driven 3d multi-joint manipulator. In *2014 IEEE International Conference on Robotics and Automation (ICRA)*, pages 5915–5920, May 2014.
- [30] <http://handheldrobotics.org>.
- [31] A. Aristidou and J. Lasenby. Fabrik: A fast, iterative solver for the inverse kinematics problem. *Graphical Models*, 73(5):243 – 260, 2011.
- [32] Dynamixel mx-64t e-manual. <http://support.robotis.com/en/product/dynamixel/mx-series/mx-64.htm>. Accessed: 2016-04-20.
- [33] P.M. Fitts. The information capacity of the human motor system in controlling the amplitude of movement. *Journal of experimental psychology*, 47(6):381, 1954.
- [34] Y. Cha and R. Myung. Extended fitts' law for 3d pointing tasks using 3d target arrangements. *International Journal of Industrial Ergonomics*, 43(4):350 – 355, 2013.

- [35] R.W. Soukoreff and I.S. MacKenzie. Towards a standard for pointing device evaluation, perspectives on 27 years of fitts' law research in hci. *International journal of human-computer studies*, 61(6):751–789, 2004.
- [36] M. Akamatsu, I.S. Mackenzie, and T. Hasbroucq. A comparison of tactile, auditory, and visual feedback in a pointing task using a mouse-type device. *Ergonomics*, 38(4):816–827, 1995.
- [37] A. Gregg-Smith and W.W. Mayol-Cuevas. Investigating spatial guidance for a cooperative handheld robot. In *IEEE International Conference on Robotics and Automation (ICRA)*. IEEE, February 2016.
- [38] J.D. Gammell, S.S Srinivasa, and T.D. Barfoot. Informed rrt*: Optimal sampling-based path planning focused via direct sampling of an admissible ellipsoidal heuristic. *arXiv preprint arXiv:1404.2334*, 2014.
- [39] S.G. Hart and L.E. Staveland. Development of nasa-tlx (task load index): Results of empirical and theoretical research. *Advances in psychology*, 52:139–183, 1988.
- [40] Natural Point. Optitrack v120:duo product page. Accessed: 2016-09-08.
- [41] Gregg-Smith. A. *Cooperative Handheld Robots*. PhD thesis, 2016. unpublished thesis.
- [42] E. Nadaraya. On estimating regression. *Theory of Probability & Its Applications*, 9(1):141–142, 1964.
- [43] I. Mohamad and D. Usman. Standardization and its effects on k-means clustering algorithm. *Res. J. Appl. Sci. Eng. Technol*, 6(17):3299–3303, 2013.
- [44] Natural Point Optitrack. Natnet api user's guide version 2.10.0, May 2016.
- [45] G. Murray. Rotation about an arbitrary axis in 3 dimensions. <http://inside.mines.edu/fs_home/gmurray/ArbitraryAxisRotation/>, June 2013. Accessed: 2016-09-08.
- [46] G. Butterworth. Pointing is the royal road to language for babies. *Pointing: Where language, culture, and cognition meet*, pages 9–33, 2003.

12 Appendices

12.1 CAD Drawings

12.2 Servomotors

12.3 Tracking Device

12.4 Arduino Board

12.5 Code Snippets

Put code for writing the motor angles to the servos, perhaps starting the new thread that just continuously writes the angles. Perhaps code for starting an experiment and tracking the bodies, getting the information out, and using them as a target. Also perhaps the regression function code? XXXXXXXX