Contest 3 Report

MIE443 Project

Savo Bajic - 1003051485 Maximilian Glidden - 1002277396 Catherine Kucaba - 1003278026

1.0 Problem Definition and Design Requirements	2
1.1 Requirements and Constraints	3
2.0 Methods and Strategy	4
3.0 Detailed Robot Design and Implementation	8
3.1 Sensory Design	8
3.1.1 Bumpers	8
3.1.2 Depth Camera/Laser Scanner	9
3.1.3 Odometry	9
3.1.4 Gyroscope	9
3.1.5 RGB Camera	9
3.2 Controller Design	10
3.2.1 Exploration and Mapping	10
3.2.2 Movement	12
3.2.3 Emotion Detection	13
3.2.4 Victim Emotion Responses	14
4.0 Future Recommendations	16
5.0 Contribution Table	18
6.0 Appendices	19
Appendix A: contest3.cpp	19
Appendix B: costmap_client.h	23
Appendix C: costmap_client.cpp	26
Appendix D: costmap_tools.h	33
Appendix E: emotionClassifier.py	37
Appendix F: emotionHandling.h	39
Appendix G: emotionHandling.cpp	40
Appendix H: emotionTrainingSample.py	48
Appendix I: explore.h	56
Appendix J: explore.cpp	59
Appendix K: frontier_search.h	67
Appendix L: frontier_search.cpp	69
Appendix M: movement.h	75
Appendix N: movement.cpp	76
Appendix O: victimLocator.py	83
7.0 References	88

1.0 Problem Definition and Design Requirements

The objectives and requirements for this portion of the project are outlined in the Contest 3 Manual [1]. The primary goal of this contest is to develop an algorithm that allows a robot to autonomously explore and map an unknown disaster environment and locate 7 victims (users). An example environment is shown in Figure 1 below. In addition, the robot must interact with the victims and provide them with important emergency announcements regarding evacuation due to the team's chosen disaster, which is described later in this section. During the interaction, the robot must correctly identify the current emotional state of the victim and appropriately interact with them based on this information. The robot must locate and interact with all 7 victims within a certain time limit. Due to the Covid-19 pandemic, this project is completed only using software, as it would typically involve a physical robot navigating a real environment.



Figure 1: Example environment layout with 7 victims [1]

The robot is simulated using TurtleBot and it navigates an enclosed environment created in gazebo. While traversing the areas, the robot will use the ROS gmapping package to dynamically create a map from information provided by the sensors on the TurtleBot, including a Kinect sensor and front bumper sensors. This map is then used in the frontier based exploration algorithm to help navigate the robot to regions it has not yet explored. The robot must appropriately identify and react to the 7 different user emotions: anger, neutral, fear, sadness, happiness, surprise, and disgust. Figure 2 below displays sample images for each of the victims' emotions. The interactions must be distinct and identifiable, using primary/reactive and secondary/deliberative emotions. The team chose the disaster to be a fight between Godzilla and King Kong, due to the recent movie that was released titled *Godzilla vs. Kong* (Adam Wingard, 2021). The victims are required to evacuate the building before Godzilla and King Kong begin to fight in the city, in the event that the building is demolished in the ensuing battle. The victims are required to get to the roof of the building as helicopters are helping to evacuate citizens.

The robot will be tested in two trials, with the best counting towards the final mark. Scoring is based on two aspects: locating and classifying all of the victims' emotions and the robot's emotional responses while interacting with the victims. Additional marks will be given based on the creativity and complexity of the emotional behaviours, as well as the team's classification rate compared to the other teams.



Figure 2: Sample images of the 7 victims' emotions [1]

The concepts required for this project can be applied to many other industries that are starting to use and rely on robots to execute tasks typically performed by humans. The navigation and exploration portions of this project are similar to what search and rescue robots accomplish, as they must navigate unknown terrain to locate and possibly assist survivors after a disaster. Another similar scenario is presented for service robots in the food or medical industries. Robots that interact with human patients or customers must be able to sense and classify their emotions in order to provide appropriate emotional behaviours in response, so the person can understand and cooperate. For example, if a customer is angry, the robot should respond in a manner to calm the customer and determine what needs to be done to help them. As people begin to use robots for customer service, these robots must be able to perform the emotional labour that comes with these roles.

1.1 Requirements and Constraints

For this contest, there are some design requirements that the team must take into account when developing the algorithms for the robot. These requirements can be found in the Contest 3 Manual [1] and are also presented below.

- 1. The contest environment is a building with multiple rooms (see Figure 1 as an example).
 - a. There are seven (7) individuals distributed throughout the environment.
 - b. The victims are stationary in the environment (i.e., they do not move).
 - c. There are no objects/obstacles in the environment other than walls.
 - d. Victim locations are marked by a person on a blue square purely for visualization (this does not impact exploration).
 - e. The robot will start at a location chosen by the teaching team.
- 2. The layout of the contest environment is not known to the team in advance.
 - a. The teaching team will generate a new environment for the trials differing from the one provided for practice.
- 3. The robot must locate and interact with all 7 victims within a maximum time limit of 20 minutes.
- 4. There are two trials per team, with the best trial counting towards the final score.
- 5. For testing, the team was provided with the following:
 - a. A dataset of faces expressing one of the seven emotions: anger, neutral, fear, sadness, happiness, surprise, and disgust.
 - b. Sample code used to train an emotion classification convolutional neural network.
 - c. A victim detector that automatically detects simulated victims.

- i. When the robot approaches a set distance from the victim, multiple images of different faces with the same emotion are provided for the CNN to classify.
- ii. For the trial runs, a different dataset of face images will be used.
- iii. The overall accuracy of the team's emotion classification method will be tested using the dataset mentioned in ii.
- d. A frontier exploration algorithm used to find the victims.
 - i. The robot generates a map of the environment by navigating to the boundaries between known and unknown regions.
- 6. During interactions, the robot is required to have unique emotional responses to the 7 possible emotions (shown in Figure 2). It must also convey information and evacuation instructions regarding the disaster.
 - a. The disaster is chosen by the team.
- 7. For the robot's responses, the team must choose 3 unique primary emotions, 3 unique secondary emotions and 1 additional unique emotion of either type from the following list: fear, positively excited, pride, anger, sad, discontent, hate, resentment, surprise, embarrassment, disgust, and rage.
 - a. The team must implement original motion and/or sound for the interactions.
- 8. The team can utilize different emotion classification methods than provided, but cannot install additional software packages.
- 9. The team cannot use the 'detectedVictim.txt' file for emotion recognition during the trials. However, it can be used for testing and debugging in practice runs.
- 10. Automatically generated log files cannot be modified.

2.0 Methods and Strategy

The team implemented supervised learning for the machine learning portion of this project. During this learning method, the robot is provided many examples of input/output pairs. The robot then develops its own function(s) that can make decisions when provided a given input. For this contest, the robot was provided a dataset that included many different face images expressing the seven victims' emotions. This form of learning is usually accomplished using neural networks; for this project, the team utilized a convolutional neural network (CNN) to classify images. A k-fold approach was used for training the AI to maximize the utility of the limited training set provided to the team. The dataset was split into five equal portions (k = 5). Five models were then trained separately, with each one using a different set of data for validation (e.g., model 1 used fold 1 for validation) and the other folds for training (i.e., 2 to 5). Of these five models, the one that performed best at correctly identifying the emotions of its validation set was kept and used in the project. At times, the classification accuracy varied significantly among the five models, from ~93% to ~62% (shown in Figure 3). Other learning methods, such as unsupervised or reinforced, were not suitable nor required for this project.

Done training the model! Best validation score: 92.82 | Average across folds: 81.03 [0.6161819306930693, 0.7738242574257426, 0.82178217821782178217 0.9282178217821783, 0.9116646039603963

Figure 3: Output file from a practice run displaying each model's classification accuracy

For exploration and navigation, the team implemented a greedy frontier-based algorithm. This method utilizes an occupancy map/grid that is generated by sensor readings obtained from the robot while it moves throughout the environment. The boundaries between open space and unexplored space, as determined by the sensory information, are the frontier regions. The robot is inclined to move towards these areas in order to explore and map more of the environment. The team's robot begins by spinning in place and taking a rotating scan of the environment. It then uses built-in ROS explore functions to move the robot throughout the regions. During its motion, the robot takes in the environment (via sensors) and generates a map using the ROS gmapping package. This frontier-based algorithm was selected over other methods, such as random walk, as it is more efficient at exploring the environment and locating the victims that the robot is required to interact with. As the team saw in Contest 1, a random or weighted random walk would sometimes result in the robot is inclined to explore unknown regions, it will explore more of the environment and find the victims faster than the other methods.

In the simulation, the robot does not locate victims using its sensors. Instead, it identifies or senses victims when it moves within a certain set distance away from them, approximately 1m for the team's practice runs. This is due to the way the contest is set up by the teaching team. In a real world scenario, victims could be located by looking for humanoid figures or faces and then approaching them to confirm. Once the victim is located, a dataset of images is directly provided to the robot of the same emotion being expressed. Therefore, sensors are again not used for this portion of the contest. The CNN and NN are then used to process the emotions and attempt to classify it into one of the seven victims' emotions. In a real world scenario, the RGB camera sensor would be used to scan for faces and capture images to then process.

The robot's emotional responses are based on behavioural control. When the robot encounters a victim and processes their emotion, it will then react according to a set of behaviours the team programmed into the algorithm. Each victim's emotion has a distinct robot emotional reaction. The team was required to use 3 primary emotions and 3 secondary emotions, with the last one being of either category. The justification for the use of each emotion and its classification as either primary or secondary is explained below.

1. Victim's Emotion: Anger Robot's Emotional Response: Resentment Type of Emotion: Secondary

The team classified resentment as a secondary emotion, as it is quite complex and is usually a mixture of other emotions. One experiences resentment when they feel that they have been treated unfairly by others, meaning that one must think/reflect before experiencing this emotion. The team used resentment as a reaction to an angry victim, since they felt that this would be the most natural response of a rescuer attempting to help this type of victim. A person may feel resentment when trying to help someone who is angry, especially because angry people tend to be stubborn and do not listen (due to their emotional state). The team also believed that resentment would be a good response to change the victim's emotion away from anger and more towards concern for their own safety, so much so that they are willing to listen to instructions and comply. When met with resentment, a person tends to evaluate themself and how they treated others in an attempt to adjust their behaviour.

2. Victim's Emotion: **Neutral** Robot's Emotional Response: **Pride** Type of Emotion: **Secondary**

The team classified pride as a secondary emotion, as one must reflect on achievements (theirs or others) to derive that deep sense of satisfaction. This is another emotion that is complex and tends to include other emotions (like happiness, excitement, etc.). The team used pride as a reaction to a neutral victim as this was one of the few positive emotions available to the team. They believed this would be a better reaction to finding a neutral person, since a negative emotion displayed by the robot could then upset the victim. Pride is a typical emotion one would feel when finding a survivor and assisting them, especially because they may have saved that person's life, which is a major accomplishment one would be proud of. In addition, people with pride tend to display a high level of confidence, which also helps to communicate information to others. People who are confident tend to be listened to more than people who are not.

3. Victim's Emotion: Fear Robot's Emotional Response: Embarrassment Type of Emotion: Secondary

The team classified embarrassment as a secondary emotion because one must think about their actions, reflect on what occurred, and then proceed to feel shame or awkwardness that leads to the emotion. The team used embarrassment as a reaction to a fearful victim for two particular reasons. Firstly, stumbling upon a person who is displaying fear may make one feel shameful or awkward for possibly inducing that fear. Thus, it may naturally follow that one would feel embarrassment for causing that kind of emotion in someone. In stressful situations, such as encountering someone who is fearful, a person may feel a sense of awkwardness on how to handle the situation, and sometimes they may resort to feeling embarrassed when trying to help. In addition, the team also believed a safe reaction to fear would be embarrassment in order to not scare the victim further, and thus have them no longer willing to listen to instructions. When a person is embarrassed, people tend to feel a sense of pity and try to help or listen to what that person may need. In a similar manner, the team is trying to ensure that the fearful victim listens to the instructions to evacuate.

Victim's Emotion: Sadness Robot's Emotional Response: Anger Type of Emotion: Primary

The team classified anger as a primary emotion, given that it is a very primal and inherent emotion among people and animals. Although anger can be a secondary emotion, at certain times it does not require one to think and is, instead, closer to a reflex (e.g., becoming angry when in pain). The team chose anger as a reaction to a victim displaying

sadness as they believed a first reaction to seeing a sad individual during an emergency would be anger. If a rescuer encountered a person who was sad, and thus possibly unwilling to move because of it, they may become angry at the individual for not caring about the situation and trying to get out safely. In addition, a rescuer may become angry that they are risking their own life to save someone who seems to not care. The team also thought that the use of anger would push the victim out of their sadness enough to evacuate safely. For example, a rescuer has to tell someone to "snap out of it" to bring that person back to the reality of the critical situation. In this way, the team would hope that the robot displaying an angry emotion would motivate the sad person to listen to the robot.

5. Victim's Emotion: **Happiness** Robot's Emotional Response: **Positively Excited** Type of Emotion: **Primary**

The team classified positively excited as a primary emotion because it is one that is quickly experienced when something good happens. A person rarely has to think in order to experience this emotion and it is very similar to the feeling of joy (another primary emotion). The team chose positively excited as a reaction to happiness since they believed that positive victim emotions should be responded to positively. This is because a negative reaction may cause the victim to then feel fear or sadness, which is counterintuitive to helping them (as they may no longer want help). Therefore, the team believed "matching" the energy of the victim would be helpful in communicating important evacuation information. In addition, rescuers who encounter happy survivors tend to display similar emotions of happiness as they are not only relieved to find the person, but they also do not wish to upset the person if there are other issues present (for example, if the person is pinned or harmed, negative emotions can send them into shock).

6. Victim's Emotion: **Surprise** Robot's Emotional Response: **Surprise** Type of Emotion: **Primary**

The team classified surprise as a primary emotion, since it is a typical reaction to external stimuli that catch people off guard. In addition, it is a feeling that people can rarely control, yet it is simple in understanding its nature. The team used surprise as a reaction to a surprised victim as a way to ease the tension in the situation. A person who is experiencing the emotion of surprise is put into a vulnerable state, as the shock of the situation could wear off and they could return to a neutral/positive state (for example, being surprised by a party) or the situation could worsen, such as the person developing fear or terror. Therefore, responding to a surprised person in a negative manner could be inappropriate. Instead, the team decided that the robot could reciprocate the emotion and act surprised as well to have found the person. This type of reaction could de-escalate the situation, as the person who is surprised would acknowledge that they surprised the robot in return. Thus, they may no longer be surprised and return to a calmer state in order to receive instructions.

Victim's Emotion: Disgust Robot's Emotional Response: Discontent Type of Emotion: Secondary

The team classified discontent as a secondary emotion since one needs to consider their current situation in order to feel a sense of dissatisfaction with it. It is one of those emotions that, although seems simple, is actually quite complex and can stem from multiple feelings. The team used discontent as a reaction to a disgusted person in a similar manner to using resentment towards anger. The team thought that a rescuer who encountered a disgusted person may feel uncomfortable and unhappy about the situation, especially if the disgust stems from the interaction. Given that discontent is similar to disgust, but at a lesser intensity, the team would hope that the victim would not be further upset by the situation and thus unwilling to cooperate. Instead, the idea is that the robot would display a similar emotion in an attempt to make the victim understand their own emotions. Therefore, they would be more likely to receive help from the robot, especially if they see that their emotions are beginning to upset the entity that is trying to help them.

3.0 Detailed Robot Design and Implementation

3.1 Sensory Design

The simulated Turtlebot platform is equipped with a variety of sensors to aid in navigation, including touch bumpers, a depth camera, cliff sensors, and odometry data. As in Contest 1, the environment the robot is exploring in Contest 3 is flat and enclosed. Cliff sensors do not produce much useful information; there are no edges or changes in elevation for the robot to detect. Similar to Contest 1, the robot does not have a map of its environment prior to initialization. The laser scanner, odometry, touch bumpers, and gyroscope were used throughout the *move_base* and *gmapping* ROS nodes, feeding into the *explore_lite* node to aid in exploration and obstacle avoidance. During the robot's emotional interactions, the odometry is used directly to aid in the control of the robot. The other sensors are handled directly by the aforementioned ROS topics without any action required of the team's code to aid in their roles.

3.1.1 Bumpers

The Turtlebot has a set of three bumpers located on the front, left, and right sides. These bumpers publish their state (1 when pressed, 0 otherwise), which in turn is fed to the *gmapping* node. The *gmapping* package adds bumper contact to the map by default, improving the fidelity of the map in situations where obstacles are not detected by the depth camera (i.e when moving backwards during an interaction).

3.1.2 Depth Camera/Laser Scanner

The Turtlebot is equipped with a Microsoft Kinect camera system, which integrates both an RGB and depth camera. The depth camera can be used to approximate the functionality of a laser scanner, which generates a distance at each point in a one-dimensional array from the depth camera input. The motivation for using the depth camera in this way was to provide sensory input to the *gmapping* and *move_base* nodes regarding the presence and proximity of obstacles before the robot. This depth data is used by *gmapping* to build a map of the environment, and by *move_base* to aid in obstacle avoidance and to build a costmap of the environment for movement planning.

3.1.3 Odometry

Encoders on the motors of the turtlebot provide information on the rotational velocity of each wheel, which is used with the forward differential kinematics model to produce linear and rotational velocities. Integrating these velocities over time produces an estimate of the robot's position and heading; this is the odometry data supplied by the *odom* topic.

Odometry data is used by the *move_base* node as an input to the local planner so it is aware of the rough distance travelled. It is also used by the *gmapping* node to stitch together a map of the environment from depth camera/laser scanner data, estimating the robot's pose with respect to the map. Odometry data was also utilized directly during the robot's emotional interactions with victims in the environment to close the control loop with respect to movement. Inputting movement commands with motor velocities and durations alone is an example of open-loop control. The robot does not track how far it has actually travelled, nor can it easily estimate its position in the world frame. Checking the odometry data during movements allows the robot to more accurately assess its motion when performing the multimodal emotional response.

3.1.4 Gyroscope

The Turtlebot's gyroscope provides information on the robot's angular velocity and orientation. This data is supplied to the *move_base* node as a sensor transform, and assists in movement and local planning.

3.1.5 RGB Camera

The Turtlebot's Kinect's integrated RGB camera provides a full colour video feed of the robot's surroundings. The team did not make use of the RGB camera during Contest 3, because the victims the robot needed to find were not physical objects in the environment. Instead, the victims were represented as a set of coordinates in the environment (stored in *victims.txt*). When the robot moves within a certain radius of these coordinates, the *victimLocator.py* function is triggered. This function publishes a set of images to the robot for emotion classification, simulating the use of the RGB camera for emotional analysis.

3.2 Controller Design

The overarching control architecture used for the third contest was a behaviour-based controller, implemented using a finite state machine. Combinations of a set of behaviours were encoded in various states ("explore", "emote", etc.), with transitions between states triggered by sensory input and the code of other states. The robot can only exist in a single state at a given time, but this state may encompass multiple sub-behaviours, called substates, within the team's code. The robot's behaviours are a combination of reactive and deliberative actions. Exploring the environment is a deliberate process, where the robot considers where it has and has not been. Emotional responses are reactive to external stimuli.

The main behaviour states are "Explore" (the main/default state) and the seven emotional reactions (Pride, Positive Excitement, Embarrassment, Surprise, Discontent, Anger, Resentment). The Explore State is the main state the robot takes while navigating the environment. In this state, the robot utilises the *explore_lite* package to explore the environment using a greedy frontier-based method. The environment is mapped using the *gmapping* package, while path planning, movement, and obstacle avoidance are handled by the *move_base* node. Occasionally, the robot will conduct a scan to reset the frontiers for exploration. When the *victimLocator:py* function publishes a victim emotion (indicating a victim has been found), the "explore" behaviour terminates and an emote behaviour is triggered. The seven emotional response states correspond to seven unique behaviours; each of these behaviours is associated with the emotional state of the victim. A state is selected based on the output of the *emotionalClassifier* function, and the associated multimodal response behaviour is performed. Once this response behaviour is complete, the robot re-enters the explore state.

3.2.1 Exploration and Mapping

Exploration is the primary state the robot is in when completing this challenge, as it tries to explore the entirety of the environment to find all seven victims within it. Exploration is largely handled using ROS's own built-in algorithms, functions, and services, as well as those provided to the team for use, namely the *Explore* class from the *explore* namespace. The team used this class to automatically explore the environment by simply calling *Explore.start()* to send it exploring or *Explore.stop()* when a controlled motion is needed again (e.g. a specific reaction to a victim).

The automated exploration is guided by a greedy frontier-based exploration approach. A frontier is defined as a division between mapped (visited) and unmapped regions of the environment, that a robot may cross. This is in contrast to a barrier which is an impassable division between regions of the map. Frontier exploration has the robot seek out and approach these frontiers to map what lies beyond them. To select which frontier of the many potentially present in an unfinished map, the robot uses a "greedy" strategy. Specifically the robot records the location of all known frontier points in its immediate vicinity (roughly 1.5m along each map axis) and averages them out to find the centroid, it then approaches this centroid and updates it as it moves. This is "greedy" because it is the local optimum, in that it is the boundary which is closest to the robot that is scanned, even if it is much smaller (and thus reveals less) than other frontiers further from the robot.

The exploration is enabled by the mapping services *gmapping* provides as part of the ROS package. As the robot travels through the environment, it uses the sensor data to continuously build a map of the robot's surroundings and estimate its position within this map. It is this *gmapping* that provides the information and guidance *Explore* requires to complete its function. The results of mapping, such as the figure provided below, are viewable using another ROS utility, *RVIZ*.



Figure 4: An example of a partially explored map. The rover lies in the center of the coloured section, the colours indicate the proximity to the robot's current exploration goal with a path drawn in red.

The team did not alter the behavior of *Explore* in any notable way, as it performed well as provided to the team. However, the team did implement supporting functions that would intervene and attempt to recover the robot from a position where *Explore* on its own would be stuck. *Explore*, due to its "greedy" algorithm, has the potential to get stuck when it arrives at a frontier centroid that does not reveal enough of the map to determine a distinct and new local frontier centroid compared to the one it just reached, and thus determines there is nowhere to go. To detect this condition, the team monitors the robot's distance travelled in the last period (roughly 10 seconds) to see if the robot has stopped moving any reasonable amount (1m). If it fails to surpass this target, the team has its code attempt a recovery, detailed more in section 3.2.2, Movement.

3.2.2 Movement

Although the majority of the robot's motion is left to be autonomously determined and handled by ROS, there are two primary states where the team's code opts to exert direct control over the robot's motion: reacting to victims and exploration recovery. To achieve direct control of the robot, the team reused the code they created for this purpose in Contest 1 as a baseline and developed it to face the unique challenges Contest 3 presented. This code is contained primarily in *movement.h* and *movement.cpp*.

Direct control is exerted on the rover by stopping the exploration (*Explore.stop(*)) and then starting a publisher to the teleoperation topic within *contest3.cpp*. Any data published to this topic will emulate that of a user, and thus override any other sections of code attempting to steer the robot's motion. This allows a linear and rotational velocity to be set for the base by the code; however, most movements for the rover are meant to be measured (e.g. going 0.5m backwards). To enact such prescribed movements, odometry feedback is used to monitor the robot's motion to ensure that it moves the required distance and not to depend on maintaining course for a set duration of time. A callback function is prepared in *movement* to be used to subscribe to the odometry so it can be utilized by the rest of the program.

Upon those foundations, the main function in these files, *travel*, is built. This function is called repeatedly to execute a desired motion defined by displacements and velocities, linear and rotational. This function keeps a record of the most recent targets provided and uses the odometry to track the progress. Once the motion is complete, it returns a value of true to signal to the program it may proceed to the next task. This design made it very well suited to the state-machine approach of the team's control system because it allows motions to easily be interrupted and change. In addition, it does not hold execution away from the main loop for a significant portion of time.

The main addition to the *movement* files was the function *checkIfMoved* used to check if the rover has been moving recently. This is used to determine if the robot's exploration code has gotten "stuck" in a location and cannot determine where to go next (explained in section 3.2.1). To accomplish this, the code regularly records the robot's positional information to a circular buffer to determine the overall travel between the ten most recent recordings (covering roughly ten seconds of real-time motion). If this distance travelled falls below a set threshold (currently set to 1m), the robot is determined as stationary and this is returned to the code that called it to signal a recovery attempt is required.

The first recovery method for a stuck robot is having it complete a whole rotation about its current position to try and reveal any potential frontiers and/or barriers. Any new features will influence the calculation of the centroid and typically be enough to "dislodge" the robot. Should the robot centroid remain as it is regardless of this rotation (for example being in the center of a room so no new barriers are found) the team will then clear the frontier blacklist to try and alter the centroid's calculation.

3.2.3 Emotion Detection

Emotion detection in this competition is achieved using a convolutional neural network model trained and operated within Python (*emotionClassifier.py* and *emotionTrainingSample.py*, respectively). As the robot approaches victims it will be provided a series of images of their face and, using *emotionClassifier* with the trained model, it will process them and estimate the emotion portrayed by the victim's face. It then compares these results and selects the most common result from the processed series of images as the victim's emotion and publishes it as a ROS topic. The team's code uses a subscriber to *emotionClassifier* to collect these results and pass them onto the remainder of the code, namely to react to the victim and instruct them to evacuate properly (explained in greater detail in section 3.2.4).

The first step in accomplishing emotion detection is to design the neural network used to classify the emotions. This was done using the *emotionTrainingSample* file. The neural network has two main parts: a series of convolution blocks forming the convolutional neural network (CNN) followed by a series of fully connected layers forming a neural network (NN). The images are fed through the CNN and its four blocks of convolutions, then into the NN, which resolves it into buckets for each of the seven responses and the confidence that emotion is present. The structure of the overall network was kept the same with the exception of the addition of a fully connected layer in the NN.

This structure was trained using a dataset to try and best identify emotions in images. It was trained using a supervised learning method. A data set of pictures with corresponding labels for the emotion in each is split into a training set and validation set when developing the model. The model is trained on the training set, using any errors in identification to try and tune the model. Once complete, the model is tested on the test set to see how it does in classifying them before it undergoes another tuning iteration. The model that best classifies the validation set is kept for use in the project.

To make better use of the limited dataset, the team used a "k-fold" approach when training. This operates in much the same way as described above. However, the dataset is split into k (the team used five) groups where k models are trained in parallel using different combinations of the groups for training and validation and keeping the best performing model on it's dataset. The dataset is loaded into *emotionClassifier* which processes victim photos as mentioned previously. The most frequently appearing emotion in the batch of images is labelled as the victim's emotional state and passed on as it gets published.

When the subscriber in the team's code receives the determined emotion from *emotionClassifier*, it is recorded and the robot switches from its exploration state to the victim response state to react accordingly. The code for handling emotions is contained across the *emotionHandling.cpp* and *emotionHandling.h*, primarily by calling *emotionReaction*. The steps that follow are explained in section 3.2.4.

3.2.4 Victim Emotion Responses

The emotional responses the robot utilizes in interactions with victims is a behaviour-based controller implemented using a state-machine with substeps in the function *emotionReaction* of *emotionHandling.cpp*. As mentioned in the previous sections, the robot begins in an "explore" state while it navigates the environment. When it encounters a victim, the robot then enters the "emote" state to first determine the emotion of the victim. Once the emotion has been classified, the robot then checks the state machine to determine the emotional response behaviour that the robot will perform in the interaction.

Each victim's emotion has a different and unique robot emotional response. The robot is programmed to display one of the following emotions: pride, positively excited, embarrassment, surprise, discontent, anger, and resentment. Each emotion acts as a state and has a series of substeps that comprise the complete emotional behaviour. Each step is either a distinct motion or a set of multimedia, such as sounds and images, that display the emotion. The system progresses through all of the behaviours in the emotion state. Once the interaction is completed, the emotion state is cleared. The algorithm then switches the state back to "explore" until a new victim is located. Table 1 below outlines the victims' emotion and the robot's emotion reactions, including the behavioural substeps used in each emotion state. For example, when the robot classifies a victim's emotion as neutral, the robot then interacts by displaying the emotion of pride and completes the actions in the response column.

In addition to using sound files, the team also included an image file to communicate evacuation information to the victims. This was implemented in the event that one or more of the victims was experiencing auditory issues, such as being hearing impaired or having outside noise interrupt the communication. The image file includes a neutral expressing face and text that explains the emergency and what to do. This image is constantly displayed while the robot explores and is only removed during interactions.

Using *travel* allows the team to produce the tailored motions required by emotional responses, such as reversing or rotating at different rates to produce a basic choreographed act.

Victim Emotion	Robot Emotion	Response
Neutral	Pride	 Play sound/show picture 360 degree slow spin Play sound
Happiness	Positively Excited	 Play sound/show picture Move forward quickly Quick spin Play sound
Fear	Embarrassment	 Play sound/show picture Move backwards slowly Turn one direction slowly, slightly (like being shy) Play sound Turn back slowly
Surprise	Surprise	 Show picture (no sound) Move backwards quickly Play sound Slowly move forwards (less distance) Play sound
Disgust	Discontent	 Play sound/show picture Move back slowly a bit slowly turn one way Play sound
Sadness	Anger	 Play sound/show picture Move forward a small distance quickly (threateningly) Play sound Move forward again quickly
Anger	Resentment	 Play sound/show picture Turn quickly 180 degrees Move straight short distance Turn back 100 degrees (as if to look over its shoulder) Play sound

 Table 1: Table summarizing the emotion reactions the robot performs

4.0 Future Recommendations

Given additional time, the team would implement changes to further improve the robot's emotional reactions and emotional classification, and address issues with exploration. One such change would be to utilize other forms of communication when interacting with victims, such as replacing the static images and voice lines with a video of a human rescuer communicating evacuation instructions. Using a video for each emotional response would allow the team to include other elements of emotion not easily portrayed by the Turtlebot. For example, a video of a human rescuer includes human motion/gestures, body language and posture; all aspects of communication not currently captured in the robot's emotional responses. With video, these non-verbal cues are integrated directly with the warnings and instructions communicated by the robot, and help to express the severity of the disaster situation more completely. Time was a limiting factor in the preparation of media for the emotional response. In addition, given that the robot used for the contest is simulated, video files would require more resources to store and run. With more time to test and develop, the team would add a video element to their emotional responses.

Improving the accuracy and robustness of the *emotionClassifier* CNN is another area to which the team would direct future development efforts. Currently, the team aims to mitigate overfitting by implementing dropout in the hidden layers of the CNN. To further reduce overfitting, and also to provide the classifier with a larger set of data to train on, the team could implement data augmentation during the training process. This would entail generating new images in the training set via the application of rotations, zooms, skews, colour filters, and random noise to existing images. This approach would result in a much larger training set, resulting in a more generalized network with increased robustness.

An issue the team faced when exploring the environment was the *explore_lite* package struggling to select a goal frontier in situations where the robot was equidistant from all viable frontiers. This behaviour was most common when the robot was initialized in the middle of an open area, where all walls were outside the range of the frontiers. The team did include a "spin" sub-behaviour that rotated the robot 360° in situations where the robot did not move for an extended period of time. However, this did not eliminate the unwanted idle behaviour in all situations. There are several related solutions to this problem the team could implement as future work. Developing a method to more quickly identify when the robot is struggling to select a frontier would make the overall exploration process more robust, with fewer long pauses. This would reduce the time lost to the unwanted behaviour in situations where the "spin" sub-behaviour was sufficient to help the planner set a goal frontier. To handle situations where a spin alone is insufficient, the team would look to add a random walk component to the sub-behaviour. This additional random displacement would be particularly useful in cases where all frontiers are equidistant, as shown in the figure on the following page.



Figure 5: An example of equidistant frontiers causing failure of the planner

Random walk behaviour could be implemented in a similar manner to the first contest; a random rotation is selected, and then the robot translates in a fixed direction. The team could tune the number of random steps taken before reinitializing the *explore_lite* package to ensure the robot could decide on a frontier for exploration. If the issue persisted, the team could also look to dynamically adjust the *maxUrange* parameter in *gmapping*. This parameter governs the maximum range at which *gmapping* will record objects and add them to the map. Increasing the vision range in situations where frontiers are equidistant from the robot would allow for distant boundaries (e.g. walls) to be detected, helping the planner decide on a viable frontier.

5.0 Contribution Table

Contributions noted by value: 1 - small amount, 3 - majority, blank for none.

Section	Savo Bajic	Maximilian Glidde	en Catherine Kucaba
1.0			3
2.0	2	1	3
3.1		3	
3.2	3	2	1
4.0	1	3	1
Robot Code	3	2	2

Table X: Contribution Table

6.0 Appendices

Appendix A: contest3.cpp

```
#include <ros/ros.h>
#include <ros/package.h>
#include "explore.h"
#include "emotionHandling.h"
#include "movement.h"
#include <time.h>
#include <sound play/sound play.h>
#include <ros/console.h>
int main(int argc, char** argv) {
  time t startTime = time(NULL);
  time t secondsElapsed = 0;
  ros::init(argc, argv, "contest3");
  ros::NodeHandle n;
&emotionCallback);
   clearEmotionState();
  ros::Subscriber odom = n.subscribe("odom", 1, &odomCallback);
  ros::Publisher vel pub;
```

```
geometry_msgs::Twist vel;
```

```
bool spinInPlace = true; // Is rover supposed to be spinning in place
  bool manualOverride = spinInPlace; // Are we manually overriding the
bot's motion
scan
  explore::Explore explore;
  ROS WARN ("STARTING MAIN LOOP!\n");
  while(ros::ok()) {
       ros::spinOnce();
      static bool overridingPrev = false; // Store overriding motion
       if (readEmotion() >= 0) {
#ifdef EMOTION BYPASS
           ROS WARN("BYPASSING EMOTION CODE");
           victimsEncountered++; // Increment victim count as though they
were handled
           clearEmotionState();
#else
           if (manualOverride == false) {
               explore.stop();
               manualOverride = true;
           emotionReaction(sc);
           if (readEmotion() < 0) {</pre>
               manualOverride = false;
               ROS INFO("Done emotional reaction");
#endif // EMOTION BYPASS
```

```
20
```

```
else {
          if (spinInPlace) {
              manualOverride = true;
               spinInPlace = ((travel(0, 0, M PI * 2, FAST SPIN)) ==
false);
              if (spinInPlace == false) {
                  ROS INFO("Done spinning in the spot.");
                  manualOverride = false; // Release control
               if (checkIfMoved(explore) == false) {
                   spinInPlace = true;
      if (manualOverride) {
          if (overridingPrev == false) {
              vel pub =
n.advertise<geometry msgs::Twist>("cmd vel mux/input/teleop", 1);
          vel.angular.z = angular;
          vel.linear.x = linear;
```

```
vel pub.publish(vel);
      if ((overridingPrev) && (manualOverride == false)) {
          travel(0, 0, 0, 0); // Reset motion code
          vel pub.shutdown(); // Stop override when no longer needed
          explore.start(); // Explore again once control is released
      overridingPrev = manualOverride; // Record override state for
      secondsElapsed = time(NULL) - startTime;
      ros::Duration(0.01).sleep();
      if (victimsEncountered == victimsExpected) {
search.", victimsExpected);
      if (secondsElapsed > 1200) {
  explore.stop();
  secondsElapsed = time(NULL) - startTime;
int(secondsElapsed));
  ROS FATAL ("TERMINATING PROGRAM");
```

Appendix B: costmap client.h

Software License Agreement (BSD License)

Copyright (c) 2015-2016, Jiri Horner. All rights reserved.

Redistribution and use in source and binary forms, with or without modification, are permitted provided that the following conditions are met:

- * Redistributions of source code must retain the above copyright notice, this list of conditions and the following disclaimer.
- * Redistributions in binary form must reproduce the above copyright notice, this list of conditions and the following disclaimer in the documentation and/or other materials provided with the distribution.
- * Neither the name of the Jiri Horner nor the names of its contributors may be used to endorse or promote products derived from this software without specific prior written permission.

THIS SOFTWARE IS PROVIDED BY THE COPYRIGHT HOLDERS AND CONTRIBUTORS "AS IS" AND ANY EXPRESS OR IMPLIED WARRANTIES, INCLUDING, BUT NOT LIMITED TO, THE IMPLIED WARRANTIES OF MERCHANTABILITY AND FITNESS FOR A PARTICULAR PURPOSE ARE DISCLAIMED. IN NO EVENT SHALL THE COPYRIGHT OWNER OR CONTRIBUTORS BE LIABLE FOR ANY DIRECT, INDIRECT, INCIDENTAL, SPECIAL, EXEMPLARY, OR CONSEQUENTIAL DAMAGES (INCLUDING, BUT NOT LIMITED TO, PROCUREMENT OF SUBSTITUTE GOODS OR SERVICES; LOSS OF USE, DATA, OR PROFITS; OR BUSINESS INTERRUPTION) HOWEVER CAUSED AND ON ANY THEORY OF LIABILITY, WHETHER IN CONTRACT, STRICT LIABILITY, OR TORT (INCLUDING NEGLIGENCE OR OTHERWISE) ARISING IN ANY WAY OUT OF THE USE OF THIS SOFTWARE, EVEN IF ADVISED OF THE POSSIBILITY OF SUCH DAMAGE.

```
#ifndef COSTMAP_CLIENT_
```

#define COSTMAP_CLIENT_

```
#include <costmap_2d/costmap_2d.h>
#include <geometry_msgs/Pose.h>
#include <map_msgs/OccupancyGridUpdate.h>
#include <nav_msgs/OccupancyGrid.h>
#include <ros/ros.h>
#include <tf/tf.h>
#include <tf/tf.n>
```

geometry msgs::Pose getRobotPose() const;

namespace explore

class Costmap2DClient

public:

from

24

```
costmap 2d::Costmap2D* getCostmap()
 return &costmap_;
from
const costmap 2d::Costmap2D* getCostmap() const
  return &costmap ;
const std::string& getGlobalFrameID() const
  return global frame ;
const std::string& getBaseFrameID() const
  return robot base frame ;
```

```
void updateFullMap(const nav msgs::OccupancyGrid::ConstPtr& msg);
void updatePartialMap(const map msgs::OccupancyGridUpdate::ConstPtr&
msg);
costmap 2d::Costmap2D costmap ;
std::string global frame ; ///< @brief The global frame for the</pre>
std::string robot base frame ; ///< @brief The frame id of the robot</pre>
private:
ros::Subscriber costmap sub ;
ros::Subscriber costmap updates sub ;
#endif
```

Appendix C: costmap_client.cpp

- * Redistributions of source code must retain the above copyright notice, this list of conditions and the following disclaimer.
- * Redistributions in binary form must reproduce the above copyright notice, this list of conditions and the following disclaimer in the documentation and/or other materials provided with the distribution.
- * Neither the name of the Jiri Horner nor the names of its contributors may be used to endorse or promote products derived from this software without specific prior written permission.

THIS SOFTWARE IS PROVIDED BY THE COPYRIGHT HOLDERS AND CONTRIBUTORS "AS IS" AND ANY EXPRESS OR IMPLIED WARRANTIES, INCLUDING, BUT NOT LIMITED TO, THE IMPLIED WARRANTIES OF MERCHANTABILITY AND FITNESS FOR A PARTICULAR PURPOSE ARE DISCLAIMED. IN NO EVENT SHALL THE COPYRIGHT OWNER OR CONTRIBUTORS BE LIABLE FOR ANY DIRECT, INDIRECT, INCIDENTAL, SPECIAL, EXEMPLARY, OR CONSEQUENTIAL DAMAGES (INCLUDING, BUT NOT LIMITED TO, PROCUREMENT OF SUBSTITUTE GOODS OR SERVICES; LOSS OF USE, DATA, OR PROFITS; OR BUSINESS INTERRUPTION) HOWEVER CAUSED AND ON ANY THEORY OF LIABILITY, WHETHER IN CONTRACT, STRICT LIABILITY, OR TORT (INCLUDING NEGLIGENCE OR OTHERWISE) ARISING IN ANY WAY OUT OF THE USE OF THIS SOFTWARE, EVEN IF ADVISED OF THE POSSIBILITY OF SUCH DAMAGE.

#include <costmap client.h>

#include <functional>
#include <mutex>
#include <string>

namespace explore

// static translation table to speed things up
std::array<unsigned char, 256> init_translation_table();
static const std::array<unsigned char, 256> cost_translation_table__ =
 init translation table();

Costmap2DClient::Costmap2DClient(ros::NodeHandle& param_nh, ros::NodeHandle& subscription nh,

const tf::TransformListener* tf)

: tf_(tf)

```
std::string costmap topic;
std::string footprint topic;
std::string costmap updates topic;
param nh.param("costmap topic", costmap topic, std::string("costmap"));
param nh.param("costmap updates topic", costmap updates topic,
               std::string("costmap updates"));
param nh.param("robot base frame", robot base frame ,
               std::string("base link"));
param nh.param("transform tolerance", transform tolerance , 0.3);
costmap sub = subscription nh.subscribe<nav msgs::OccupancyGrid>(
    costmap topic, 1000,
    [this](const nav msgs::OccupancyGrid::ConstPtr& msg) {
      updateFullMap(msg);
    });
         costmap topic.c str());
auto costmap msg = ros::topic::waitForMessage<nav msgs::OccupancyGrid>(
    costmap topic, subscription nh);
updateFullMap(costmap msg);
costmap updates sub =
    subscription nh.subscribe<map msgs::OccupancyGridUpdate>(
        costmap updates topic, 1000,
        [this](const map msqs::OccupancyGridUpdate::ConstPtr& msq) {
          updatePartialMap(msg);
        });
std::string tf prefix = tf::getPrefixParam(param nh);
robot base frame = tf::resolve(tf prefix, robot base frame );
```

```
ros::Time last error = ros::Time::now();
while (ros::ok() &&
        !tf ->waitForTransform(global frame , robot base frame ,
ros::Time(),
                               ros::Duration(0.1), ros::Duration(0.01),
                               &tf error)) {
  ros::spinOnce();
  if (last error + ros::Duration(5.0) < ros::Time::now()) {</pre>
         "Timed out waiting for transform from %s to %s to become
available "
         "before subscribing to costmap, tf error: %s",
         robot base frame .c str(), global frame .c str(),
tf error.c str());
     last error = ros::Time::now();
  tf error.clear();
void Costmap2DClient::updateFullMap(const
nav msgs::OccupancyGrid::ConstPtr& msg)
global frame = msg->header.frame id;
unsigned int size in cells x = msg->info.width;
unsigned int size in cells y = msg->info.height;
double resolution = msg->info.resolution;
double origin x = msg->info.origin.position.x;
double origin y = msg->info.origin.position.y;
```

```
size in cells y);
costmap .resizeMap(size in cells x, size in cells y, resolution,
origin x,
                    origin y);
auto* mutex = costmap .getMutex();
std::lock guard<costmap 2d::Costmap2D::mutex t> lock(*mutex);
unsigned char* costmap data = costmap .getCharMap();
size t costmap size = costmap .getSizeInCellsX() *
costmap .getSizeInCellsY();
ROS DEBUG("full map update, %lu values", costmap size);
for (size t i = 0; i < costmap size && i < msg->data.size(); ++i) {
  unsigned char cell cost = static cast<unsigned char>(msg->data[i]);
  costmap data[i] = cost translation table [cell cost];
ROS DEBUG("map updated, written %lu values", costmap size);
void Costmap2DClient::updatePartialMap(
  const map msgs::OccupancyGridUpdate::ConstPtr& msg)
ROS DEBUG("received partial map update");
global frame = msg->header.frame id;
if (msg - x < 0 | | msg - y < 0) {
  ROS ERROR ("negative coordinates, invalid update. x: %d, y: %d", msg->x,
             msq ->y);
size t x0 = \text{static cast} < \text{size t} (\text{msg} - x);
size t y0 = static cast<size t>(msg->y);
size t xn = msg - width + x0;
size t yn = msg->height + y0;
auto* mutex = costmap .getMutex();
```

```
std::lock guard<costmap 2d::Costmap2D::mutex t> lock(*mutex);
size_t costmap_xn = costmap_.getSizeInCellsX();
size t costmap yn = costmap .getSizeInCellsY();
if (xn > costmap xn || x0 > costmap xn || yn > costmap yn ||
    y0 > costmap yn) {
           x0, xn, y0, yn, costmap xn, costmap yn);
unsigned char* costmap data = costmap .getCharMap();
for (size t y = y0; y < yn \&\& y < costmap yn; ++y) {
  for (size t x = x0; x < xn && x < costmap xn; ++x) {
    size t idx = costmap .getIndex(x, y);
    unsigned char cell cost = static cast<unsigned char>(msg->data[i]);
    ++i;
geometry msgs::Pose Costmap2DClient::getRobotPose() const
tf::Stamped<tf::Pose> global pose;
global pose.setIdentity();
tf::Stamped<tf::Pose> robot pose;
robot pose.setIdentity();
robot pose.frame id = robot base frame ;
robot pose.stamp = ros::Time();
    ros::Time::now(); // save time for checking tf delay later
  tf_->transformPose(global_frame_, robot pose, global pose);
```

```
ex.what());
  return {};
                      ex.what());
 } catch (tf::ExtrapolationException& ex) {
%s\n",
                      ex.what());
  return {};
if (current time.toSec() - global pose.stamp .toSec() >
  ROS WARN THROTTLE (1.0, "Costmap2DClient transform timeout. Current
time: "
                     current time.toSec(), global pose.stamp .toSec(),
  return {};
geometry msgs::PoseStamped msg;
tf::poseStampedTFToMsg(global pose, msg);
return msg.pose;
std::array<unsigned char, 256> init translation table()
std::array<unsigned char, 256> cost translation table;
```

```
cost_translation_table[i] =
    static_cast<unsigned char>(1 + (251 * (i - 1)) / 97);
}
// special values:
cost_translation_table[0] = 0; // NO obstacle
cost_translation_table[99] = 253; // INSCRIBED obstacle
cost_translation_table[100] = 254; // LETHAL obstacle
cost_translation_table[static_cast<unsigned char>(-1)] = 255; // UNKNOWN
return cost_translation_table;
}
// namespace explore
```

Appendix D: costmap_tools.h

```
unsigned int size_x_ = costmap.getSizeInCellsX(),
              size_y_ = costmap.getSizeInCellsY();
if (idx > size_x_ * size_y_ - 1) {
  return out;
  out.push back(idx - 1);
  out.push back(idx + 1);
  out.push back(idx - size x );
if (idx < size_x_ * (size_y_ - 1)) {
  out.push back(idx + size x );
return out;
 @param costmap Reference to map data
std::vector<unsigned int> nhood8(unsigned int idx,
                                const costmap 2d::Costmap2D& costmap)
std::vector<unsigned int> out = nhood4(idx, costmap);
```

```
unsigned int size x = costmap.getSizeInCellsX(),
             size y = costmap.getSizeInCellsY();
  return out;
  out.push back(idx - 1 - size_x_);
if (idx % size x > 0 && idx < size x * (size_y - 1)) {
  out.push back(idx - 1 + size x );
  out.push back(idx + 1 - size x );
if (idx \% size x < size x - 1 && idx < size x * (size y - 1)) {
  out.push back(idx + 1 + size x );
return out;
 @param result Index of located cell
 @param start Index initial cell to search from
 @param costmap Reference to map data
bool nearestCell(unsigned int& result, unsigned int start, unsigned char
val,
               const costmap 2d::Costmap2D& costmap)
const unsigned char* map = costmap.getCharMap();
const unsigned int size x = costmap.getSizeInCellsX(),
                   size y = costmap.getSizeInCellsY();
```
```
return false;
```

```
std::vector<bool> visited_flag(size_x * size_y, false);
bfs.push(start);
visited flag[start] = true;
while (!bfs.empty()) {
  unsigned int idx = bfs.front();
  bfs.pop();
  if (map[idx] == val) {
    result = idx;
  for (unsigned nbr : nhood8(idx, costmap)) {
    if (!visited flag[nbr]) {
      bfs.push(nbr);
      visited flag[nbr] = true;
#endif
```

Appendix E: emotionClassifier.py

```
from future import print function
import torch
import cv2
import torchvision
import numpy as np
import rospy
import roslib # Needed to import ModelStates
import argparse
from gazebo msgs.msg import ModelStates
from geometry msgs.msg import Pose
from std msgs.msg import Int32
from emotionTrainingSample import EmotionClassificationNet
from mie443 contest3.msg import EmotionMsg
from mie443 contest3.msg import EmotionFaceMsg
import matplotlib.pyplot as plt
class EmotionDetector(object):
  def init (self, args):
       super(EmotionDetector, self). init ()
      device = torch.device('cpu')
      if args.gpu:
           device = torch.device('cuda:0')
      self.model = EmotionClassificationNet()
       self.model = self.model.to(device)
       self.model.load state dict(torch.load(args.model file,
map location=device))
      self.model.eval()
       self.vis = args.vis
       print('Setting up subscribers.')
```

```
self.emotion sub = rospy.Subscriber('/emotion img', EmotionFaceMsg,
self.emotionsub)
       self.emotion pub = rospy.Publisher('/detected emotion', Int32,
queue size=1)
       self.emotion file = open('detectedVictim.txt', 'w')
  def showImBatch(self, imgs):
       img grid = torchvision.utils.make grid(imgs)
       self.matplotlib imshow(img grid)
  def matplotlib imshow(self, img):
      npimg = img.numpy().transpose([1,2,0])
       cv2.imshow('Input', npimg)
       cv2.waitKey(0)
  def emotionsub(self, msg):
      with torch.no grad():
          imgs = msg.data
          w = msg.width
          h = msg.height
          b = msg.batch
           imgs = torch.from numpy(np.array(imgs)).view(b, 1, h,
w).float()
           if self.vis:
               print('Showing images.')
               self.showImBatch(imgs)
           emotions = self.model(imgs, True)
           uniqueEmotions, counts = emotions.unique(sorted=True,
return_counts=True)
           print('uniqueEmotions:', uniqueEmotions)
           print('EmotionCounts:', counts)
           print(uniqueEmotions[max idx])
           intmsg = Int32()
           intmsg.data = uniqueEmotions[max idx].item()
           self.emotion pub.publish(intmsg)
```

```
self.emotion file.write(str(uniqueEmotions[max idx].item()))
  def logEmotionHistory(self):
       self.emotion file.close()
def getInputArgs():
  parser = argparse.ArgumentParser('MIE443 contest3 victim emotion
detector.')
  parser.add argument('--gpu', dest='gpu',
default=torch.cuda.is available(), type=bool, help='Use gpu for training')
   parser.add argument('--model', dest='model file',
default='mdl best.pth', type=str, help='NN model to use for emotion
detection.')
   parser.add argument('--vis', dest='vis', default=False,
action='store true', help='Visualize the received images.')
  args = parser.parse args()
   return args
if name == " main ":
   rospy.init node('emotionDetector')
   args = getInputArgs()
  victim locations = EmotionDetector(args)
   rospy.spin()
```

Appendix F: emotionHandling.h

```
#ifndef EMOTION_HANDLING_GROUP22_H
#define EMOTION_HANDLING_GROUP22_H
#include <ros/ros.h>
#include <ros/package.h>
#include <std_msgs/Int32.h> // Emotion messages
#include "movement.h"
#include <sound_play/sound_play.h>
#include <opencv2/core.hpp>
#include <cv.h>
#include <cv_bridge/cv_bridge.h>
```

```
#include "opencv2/highgui.hpp"
void emotionCallback (const std_msgs::Int32::ConstPtr& msg);
int32 t readEmotion(void); // Returns most recent emotion scanned
void clearEmotionState(void); // Clears emotion handler
void emotionReaction(sound play::SoundClient &sc); // Handles emotion
reactions
void showImage(std::string fileLocation); // Show an image on screen
extern const char *emotionName[7]; // Emotion descriptions
extern int32 t emotionValue; // Most recent scanned emotion (-1 if no
emotion found/present)
extern int victimsEncountered; // Number of victims encountered thus far
extern const int victimsExpected; // Expected victim count
#endif
```

Appendix G: emotionHandling.cpp



```
6=Neutral
```

```
int victimsEncountered = 0; // Number of victims encountered thus far
const int victimsExpected = 7; // Expected victim count for environment
const char *emotionName[7] = { "Angry", "Disgust", "Fear", "Happy", "Sad",
"Surprise", "Neutral"}; // Emotion descriptions
int32 t emotionValue = -1; // Stores most recent scanned emotion (-1 if no
emotion found/present)
const std::string pathToSounds = ros::package::getPath("mie443 contest3")
+ "/sounds/";
const std::string pathToImages = ros::package::getPath("mie443 contest3")
void emotionReaction(sound play::SoundClient &soundPlayer) {
  static int emotionStep = 0;
  bool goToNextStep = false;
  const int soundPause = 5; // Length to sleep after each audio track
   if (emotionValue == 0) {
      if (emotionStep == 0) {
           soundPlayer.playWave(pathToSounds + "resentment1.wav");
           showImage(pathToImages + "resentment.png");
          ros::spinOnce();
          sleep(soundPause);
           emotionStep++; // Go to next step
       else if (emotionStep == 1) {
           goToNextStep = travel(0,0,M PI,FAST SPIN);
           if (goToNextStep) emotionStep++; // Go to next step once done
```

```
else if (emotionStep == 2) {
        goToNextStep = travel(0.3, FAST MOVE, 0, 0);
        if (goToNextStep) emotionStep++;
    else if (emotionStep == 3) {
        goToNextStep = travel(0,0,DEG2RAD(-100),-SLOW SPIN);
        if (goToNextStep) emotionStep++;
    else if (emotionStep == 4) {
        soundPlayer.playWave(pathToSounds + "resentment2.wav");
        ros::spinOnce();
        sleep(soundPause);
        clearEmotionState(); // Clear emotion reaction once done
else if (emotionValue == 1) {
    if (emotionStep == 0) {
        soundPlayer.playWave(pathToSounds + "discontent1.wav");
        showImage(pathToImages + "discontent.png");
        ros::spinOnce();
        sleep(soundPause);
        emotionStep++; // Go to next step
    else if (emotionStep == 1) {
        goToNextStep = travel(0.25,-SLOW MOVE,0,0);
        if (goToNextStep) emotionStep++; // Go to next step once done
    else if (emotionStep == 2) {
        goToNextStep = travel(0,0,M PI,SLOW SPIN);
        if (goToNextStep) emotionStep++;
```

```
else if (emotionStep == 3) {
        soundPlayer.playWave(pathToSounds + "discontent2.wav");
        ros::spinOnce();
        sleep(soundPause);
        clearEmotionState(); // Clear emotion reaction once done
else if (emotionValue == 2) {
    if (emotionStep == 0) {
        soundPlayer.playWave(pathToSounds + "embarrassment1.wav");
        showImage(pathToImages + "embarrassment.png");
       ros::spinOnce();
        sleep(soundPause);
        emotionStep++; // Go to next step
    else if (emotionStep == 1) {
        goToNextStep = travel(-0.4,-SLOW MOVE,0,0);
        if (goToNextStep) emotionStep++; // Go to next step once done
        goToNextStep = travel(0,0,DEG2RAD(45),-SLOW SPIN);
        if (goToNextStep) emotionStep++;
    if (emotionStep == 3) {
        soundPlayer.playWave(pathToSounds + "embarrassment2.wav");
        ros::spinOnce();
        sleep(soundPause);
        emotionStep++; // Go to next step
    else if (emotionStep == 4) {
        goToNextStep = travel(0,0,DEG2RAD(135),-SLOW SPIN);
        if (goToNextStep) clearEmotionState();
```

```
if (emotionStep == 0) {
        soundPlayer.playWave(pathToSounds + "excited1.wav");
        showImage(pathToImages + "excited.png");
        ros::spinOnce();
        sleep(soundPause);
        emotionStep++; // Go to next step
    else if (emotionStep == 1) {
        goToNextStep = travel(0.2,FAST MOVE,0,0);
        if (goToNextStep) emotionStep++; // Go to next step once done
    else if (emotionStep == 2) {
        goToNextStep = travel(0,0,M PI * 2,FAST SPIN);
        if (goToNextStep) emotionStep++;
    else if (emotionStep == 3) {
        soundPlayer.playWave(pathToSounds + "excited2.wav");
        ros::spinOnce();
        sleep(soundPause);
        clearEmotionState(); // Clear emotion reaction once done
else if (emotionValue == 4) {
    if (emotionStep == 0) {
        soundPlayer.playWave(pathToSounds + "anger1.wav");
        showImage(pathToImages + "anger.png");
        ros::spinOnce();
        sleep(soundPause);
```

```
emotionStep++; // Go to next step
    else if (emotionStep == 1) {
        goToNextStep = travel(0.2,FAST MOVE,0,0);
        if (goToNextStep) emotionStep++; // Go to next step once done
    else if (emotionStep == 2) {
        soundPlayer.playWave(pathToSounds + "anger2.wav");
        ros::spinOnce();
        sleep(soundPause);
        emotionStep++; // Go to next step
    else if (emotionStep == 3) {
        goToNextStep = travel(0.2,FAST MOVE,0,0);
        if (goToNextStep) clearEmotionState(); // Clear emotion
else if (emotionValue == 5) {
    if (emotionStep == 0) {
        showImage(pathToImages + "surprise.png");
        emotionStep++; // Go to next step
    else if (emotionStep == 1) {
        goToNextStep = travel(-0.3, -FAST MOVE, 0, 0);
        if (goToNextStep) emotionStep++; // Go to next step once done
    else if (emotionStep == 2) {
        soundPlayer.playWave(pathToSounds + "surprise1.wav");
        ros::spinOnce();
        sleep(soundPause);
        emotionStep++; // Go to next step
```

```
else if (emotionStep == 3) {
        goToNextStep = travel(0.2,SLOW MOVE,0,0);
        if (goToNextStep) emotionStep++; // Go to next step once done
    else if (emotionStep == 4) {
        soundPlayer.playWave(pathToSounds + "surprise2.wav");
        ros::spinOnce();
        sleep(soundPause);
        clearEmotionState(); // Clear emotion reaction once done
else if (emotionValue == 6) {
    if (emotionStep == 0) {
        soundPlayer.playWave(pathToSounds + "pride1.wav");
        ros::spinOnce();
        showImage(pathToImages + "pride.png");
        sleep(soundPause);
        emotionStep++; // Go to next step
    else if (emotionStep == 1) {
        goToNextStep = travel(0,0,2*M PI,SLOW SPIN);
        if (goToNextStep) emotionStep++; // Go to next step once done
    else if (emotionStep == 2) {
        soundPlayer.playWave(pathToSounds + "pride2.wav");
        ros::spinOnce();
        sleep(soundPause);
        clearEmotionState(); // Clear emotion reaction once done
```

```
if (emotionValue < 0) {
       emotionStep = 0;
       victimsEncountered++; // Increment victim count once done
void showImage(std::string fileLocation) {
       cv::Mat img = cv::imread(fileLocation, cv::IMREAD UNCHANGED);
       for (int i = 0; i < img.size().width; i++) {</pre>
           for (int j = 0; j < img.size().width; j++) {
               int index = 4 *i + j * 4 * img.size().width + 3;
               if (img.data[index] < 127 ) {</pre>
                   img.data[index] = 255;
                   img.data[index-1] = 255;
                   img.data[index-2] = 255;
                   img.data[index-3] = 255;
       cv::resize(img, img, cv::Size(480, 480)); // Resize to be visible on
       cv::imshow("Current Response", img);
       cv::waitKey(250); // Wait 50 ms so image appears
fileLocation.c str());
```

```
void emotionCallback(const std msgs::Int32::ConstPtr& msg) {
  emotionValue = msg->data;
emotionValue, emotionName[emotionValue], victimsEncountered + 1);
int32 t readEmotion(void) {
  return emotionValue;
void clearEmotionState(void) {
  showImage(pathToImages + "neutral.png"); // Normal image for victims
between reactions
  emotionValue = -1;
```

Appendix H: emotionTrainingSample.py

```
#!/usr/bin/env python3
import torch
import torch.nn as nn
import argparse
from tqdm.auto import tqdm
import matplotlib.pyplot as plt
# Parse the input arguments.
def getInputArgs():
```

```
parser = argparse.ArgumentParser('Sample for training an emotion
classification model.')
   parser.add argument('--gpu', dest='gpu',
default=torch.cuda.is available(), type=bool, help='Use gpu for training')
   parser.add argument('--nepoch', dest='nepoch', default=50, type=int,
help='Number of training epochs')
   parser.add argument('--mdl', dest='mdl', default=None, type=str,
help='Model to load')
  parser.add argument('--val', dest='val', action='store true', help='Get
validation score')
  parser.add argument('-f') # Added fo use in Google Colab
  args = parser.parse args()
  return args
class EmotionClassificationNet(nn.Module):
  def init (self):
       super(EmotionClassificationNet, self). init ()
       self.cnn = nn.Sequential(
           nn.Conv2d(1, 64, 3, padding=1),
          nn.ReLU(),
           nn.BatchNorm2d(64),
          nn.MaxPool2d(2, 2),
          nn.Dropout(0.25),
           nn.Conv2d(64, 128, 3, padding=1),
          nn.ReLU(),
           nn.BatchNorm2d(128),
           nn.MaxPool2d(2, 2),
          nn.Dropout(0.25),
           nn.Conv2d(128, 128, 3, padding=1),
           nn.ReLU(),
           nn.BatchNorm2d(128),
           nn.MaxPool2d(2, 2),
           nn.Dropout(0.25),
           nn.Conv2d(128, 128, 3, padding=1),
           nn.ReLU(),
           nn.BatchNorm2d(128),
```

```
nn.MaxPool2d(2, 2),
          nn.Dropout(0.25),
           nn.Conv2d(128, 128, 3, padding=1),
           nn.ReLU(),
          nn.BatchNorm2d(128),
          nn.Dropout(0.25),
      self.nn = nn.Sequential(
           nn.Linear(1152, 512),
          nn.ReLU(),
          nn.Dropout(0.25),
          nn.Linear(512, 256),
          nn.ReLU(),
          nn.Dropout(0.25),
          nn.Linear(256, 128),
          nn.ReLU(),
          nn.Dropout(0.25),
  def forward(self, x, test mode=False):
      batch_size = x.shape[0]
       feats = self.cnn(x) # [128, 128, 3, 3]
      out = self.nn(feats.view(batch size, -1))
      if test mode:
       return out
def getDataset(splitData):
  import pathlib
3/train split.pth' #Cathy
  pathToData = '/content/drive/MyDrive/MIE443 Robot/Contest
3/train split.pth' #Savo
```

```
if pathlib.Path(pathToData).exists():
       if splitData:
           probs = torch.ones(train imgs.shape[0]) * 0.3
           val set mask = torch.bernoulli(probs).bool()
           val imgs = train imgs[val set mask]
          val labels = train labels[val set mask]
           train imgs = train imgs[~val set mask]
           train labels = train labels[~val set mask]
           return (train imgs, train labels), (val imgs, val labels)
           train imgs, train labels = torch.load(pathToData)
           return train imgs, train labels
       print('The provided dataset does not exist!')
       exit(0)
def getDataloader(args):
  train, val = getDataset(args)
  train dataset = torch.utils.data.TensorDataset(*train)
  val dataset = torch.utils.data.TensorDataset(*val)
  batch size = 128
  weights label = train[1].unique(return counts=True,
sorted=True) [1].float().reciprocal()
   weights = torch.zeros like(train[1], dtype=torch.float)
  for idx, label in enumerate(train[1]):
       weights[idx] = weights label[label]
   sampler = torch.utils.data.sampler.WeightedRandomSampler(weights,
len(weights))
   train loader = torch.utils.data.DataLoader(train dataset,
batch size=batch size,
                                           num workers=2, sampler=sampler)
   val loader = torch.utils.data.DataLoader(val dataset,
batch size=batch_size,
                                           shuffle=False, num workers=2)
```

```
return train loader, val loader
def train loop(mdl, loss fn, optim, dl, device):
  pbar = tqdm(dynamic ncols=True, total=int(len(dl)))
  running loss = 0
  for nex, ex in enumerate(dl):
      ims, labels, = ex
      ims = ims.to(device)
      labels = labels.to(device)
      optim.zero grad()
      outs = mdl(ims)
      loss = loss fn(outs, labels)
      loss.backward(loss)
      optim.step()
      running loss += loss.item()
           status = 'Loss: %.4f '%(loss / n_batch_loss)
          pbar.set description(status)
      pbar.update(1)
  pbar.close()
   return mdl
def calc acc(mdl, dl, dl type, device):
  with torch.no grad():
      pbar = tqdm(dynamic ncols=True, total=int(len(dl)))
      total = 0
      ncorrect = 0
       for nex, ex in enumerate(dl):
          ims, labels, = ex
          ims = ims.to(device)
          labels = labels.to(device)
          predicted = mdl(ims, True)
          total += labels.size(0)
```

```
ncorrect += (predicted == labels).sum().item()
          status = '%s ACC: %.4f '%(dl type, float(ncorrect) / total)
          pbar.set description(status)
          pbar.update(1)
  pbar.close()
   return float(ncorrect) / total
def foldMaker(sizeOfData, numFolds):
  width = int(sizeOfData / numFolds)
  folds = []
  for sliceNum in range(numFolds):
      trainIDs = set(range(sizeOfData))
      validationIDs = set(range(sliceNum * width, (sliceNum + 1) *
width))
      trainIDs = trainIDs - validationIDs # Remove validation entries
      folds.append((list(trainIDs), list(validationIDs)))
  return folds
if name == " main ":
  args = getInputArgs()
  mdl = EmotionClassificationNet()
  ce loss = nn.CrossEntropyLoss()
  optimizer = torch.optim.Adam(mdl.parameters()) #, lr=0.005) # Learning
  if args.gpu:
      device = torch.device('cuda:0')
      device = torch.device('cpu')
  if args.mdl is not None:
      mdl.load state dict(torch.load(args.mdl))
  mdl = mdl.to(device)
```

```
if args.val:
      print('Val ACC loop')
      , val dl = getDataloader(splitData = 1)
      mdl.train(False)
      print('VAL ACC: ', val acc)
      dataImage, dataLabel = getDataset(splitData = 0)
      indecies = torch.randperm(dataImage.size()[0])
      dataImage = dataImage[indecies]
      dataLabel = dataLabel[indecies]
       kFolds = foldMaker(dataImage.size()[0], k)
      bestOverall = 0
       foldAccuracy = [0] * k
       for foldCount, (trainIDs, validationIDs) in enumerate(kFolds):
           print("Fold " + str(foldCount + 1) + " of " + str(k))
           trainingSet =
torch.utils.data.TensorDataset(dataImage[trainIDs], dataLabel[trainIDs])
           validationSet =
torch.utils.data.TensorDataset(dataImage[validationIDs],
dataLabel[validationIDs])
function)
           weights label = dataLabel[trainIDs].unique(return counts=True,
sorted=True)[1].float().reciprocal()
```

```
weights = torch.zeros like(dataLabel[trainIDs],
dtype=torch.float)
           for idx, label in enumerate(dataLabel[trainIDs]):
               weights[idx] = weights label[label]
           sampler =
torch.utils.data.sampler.WeightedRandomSampler(weights, len(weights))
           train dl = torch.utils.data.DataLoader(trainingSet,
batch size=batch size,
                                                    num workers=2,
sampler=sampler)
           val dl = torch.utils.data.DataLoader(validationSet,
batch size=batch size,
                                                   shuffle=False,
num workers=2)
           for epoch in range(args.nepoch):
               loopMessage = 'Train loop %d of %d for fold %d of
%d.'%(epoch + 1, args.nepoch, foldCount + 1, k)
               print(loopMessage)
               mdl.train(True)
               mdl = train loop(mdl, ce loss, optimizer, train dl, device)
               mdl.train(False)
               if val acc > foldAccuracy[foldCount]:
                   foldAccuracy[foldCount] = val acc
               if val acc > bestOverall:
```

```
bestOverall = val_acc
torch.save(mdl.state_dict(), 'mdl_best.pth')

# Print out summary of training
print('Done training the model!')
outputReview = 'Best validation score: %.2f | Average across folds:
%.2f'%(bestOverall * 100.0, sum(foldAccuracy) * 100.0 / k)
print(outputReview)
print('Peak validation accuracy for each fold')
for count, acc in enumerate(foldAccuracy):
temp = '\tFold %d - %.2f %%'%(count + 1, acc*100.0)
print(temp)
```

Appendix I: explore.h

FOR A PARTICULAR PURPOSE ARE DISCLAIMED. IN NO EVENT SHALL THE COPYRIGHT OWNER OR CONTRIBUTORS BE LIABLE FOR ANY DIRECT, INDIRECT, INCIDENTAL, SPECIAL, EXEMPLARY, OR CONSEQUENTIAL DAMAGES (INCLUDING, BUT NOT LIMITED TO, PROCUREMENT OF SUBSTITUTE GOODS OR SERVICES; LOSS OF USE, DATA, OR PROFITS; OR BUSINESS INTERRUPTION) HOWEVER CAUSED AND ON ANY THEORY OF LIABILITY, WHETHER IN CONTRACT, STRICT LIABILITY, OR TORT (INCLUDING NEGLIGENCE OR OTHERWISE) ARISING IN ANY WAY OUT OF THE USE OF THIS SOFTWARE, EVEN IF ADVISED OF THE POSSIBILITY OF SUCH DAMAGE.

#ifndef NAV_EXPLORE_H_
#define NAV EXPLORE H

#include <memory>
#include <mutex>
#include <string>
#include <vector>

```
#include <actionlib/client/simple_action_client.h>
#include <geometry_msgs/PoseStamped.h>
#include <move_base_msgs/MoveBaseAction.h>
#include <ros/ros.h>
#include <visualization msgs/MarkerArray.h>
```

#include <costmap_client.h>
#include <frontier search.h>

```
namespace explore
```

```
/ * *
```

```
@class Explore
@brief A class adhering to the robot actions::Action interface
```

noves the

robot base to explore its environment.

lass Explore

bublic: Explore();

```
~Explore();
void start();
void stop();
void makePlan();
void visualizeFrontiers(
void reachedGoal(const actionlib::SimpleClientGoalState& status,
                 const move base msgs::MoveBaseResultConstPtr& result,
                 const geometry msgs::Point& frontier goal);
bool goalOnBlacklist(const geometry msgs::Point& goal);
ros::NodeHandle private nh ;
ros::NodeHandle relative nh ;
ros::Publisher marker array publisher ;
tf::TransformListener tf listener ;
Costmap2DClient costmap client ;
actionlib::SimpleActionClient<move base msgs::MoveBaseAction>
    move base client ;
ros::Timer exploring timer ;
geometry msgs::Point prev goal ;
double prev distance ;
ros::Time last progress ;
```





Appendix J: explore.cpp

```
#include <explore.h>
#include <thread>
inline static bool operator == (const geometry msgs:: Point& one,
double dy = one.y - two.y;
double dist = sqrt(dx * dx + dy * dy);
return dist < 0.01;
namespace explore
Explore::Explore()
: private nh ("~")
, tf listener (ros::Duration(10.0))
 , costmap client (private_nh_, relative_nh_, &tf_listener_)
, move base client ("move base")
 , prev distance (0)
double timeout;
double min frontier size;
private_nh_.param("planner_frequency", planner_frequency_, 1.0);
private_nh_.param("progress_timeout", timeout, 30.0);
```

```
progress timeout = ros::Duration(timeout);
private nh .param("visualize", visualize , false);
private nh .param("potential scale", potential scale , 1e-3);
private nh .param("orientation scale", orientation scale , 0.0);
private_nh_.param("gain_scale", gain_scale_, 1.0);
private_nh_.param("min_frontier_size", min_frontier_size, 0.5);
search =
frontier exploration::FrontierSearch(costmap client .getCostmap(),
                                                potential scale ,
gain scale ,
if (visualize ) {
  marker array publisher =
      private_nh_.advertise<visualization msgs::MarkerArray>("frontiers",
10);
move base client .waitForServer();
exploring timer =
    relative nh .createTimer(ros::Duration(1. / planner frequency ),
                              [this](const ros::TimerEvent&) { makePlan();
});
Explore::~Explore()
stop();
void Explore::visualizeFrontiers(
blue.r = 0;
```

```
blue.b = 1.0;
std msgs::ColorRGBA red;
red.g = 0;
red.b = 0;
red.a = 1.0;
std msgs::ColorRGBA green;
green.r = 0;
green.g = 1.0;
green.b = 0;
green.a = 1.0;
ROS DEBUG("visualising %lu frontiers", frontiers.size());
visualization msgs::MarkerArray markers msg;
std::vector<visualization msgs::Marker>& markers = markers msg.markers;
m.header.frame id = costmap client .getGlobalFrameID();
m.header.stamp = ros::Time::now();
m.ns = "frontiers";
m.scale.x = 1.0;
m.scale.y = 1.0;
m.scale.z = 1.0;
m.color.r = 0;
m.color.g = 0;
m.color.b = 255;
m.color.a = 255;
m.lifetime = ros::Duration(0);
m.frame locked = true;
double min cost = frontiers.empty() ? 0. : frontiers.front().cost;
for (auto& frontier : frontiers) {
  m.type = visualization msgs::Marker::POINTS;
  m.id = int(id);
```

```
m.pose.position = {};
  m.scale.x = 0.1;
  m.scale.z = 0.1;
  m.points = frontier.points;
  if (goalOnBlacklist(frontier.centroid)) {
    m.color = red;
    m.color = blue;
  markers.push back(m);
  ++id;
  m.type = visualization msgs::Marker::SPHERE;
  m.id = int(id);
  m.pose.position = frontier.initial;
smaller)
  double scale = std::min(std::abs(min cost * 0.4 / frontier.cost), 0.5);
  m.scale.x = scale;
  m.scale.z = scale;
  m.points = \{\};
  m.color = green;
  markers.push back(m);
  ++id;
size t current markers count = markers.size();
  m.id = int(id);
  markers.push back(m);
marker array publisher .publish(markers msg);
```

void Explore::makePlan()

```
auto pose = costmap client .getRobotPose();
auto frontiers = search .searchFrom(pose.position);
ROS DEBUG("found %lu frontiers", frontiers.size());
 ROS DEBUG("frontier %zd cost: %f", i, frontiers[i].cost);
if (frontiers.empty()) {
 stop();
if (visualize ) {
 visualizeFrontiers(frontiers);
    std::find if not(frontiers.begin(), frontiers.end(),
                     [this](const frontier exploration::Frontier& f) {
                       return goalOnBlacklist(f.centroid);
                     });
if (frontier == frontiers.end()) {
  stop();
geometry msgs::Point target position = frontier->centroid;
bool same goal = prev goal == target position;
prev_goal_ = target_position;
if (!same goal || prev distance > frontier->min distance) {
 last progress = ros::Time::now();
 prev distance = frontier->min distance;
```

```
if (ros::Time::now() - last progress > progress timeout ) {
  frontier blacklist .push back(target position);
  ROS DEBUG("Adding current goal to black list");
  makePlan();
if (same goal) {
move base msgs::MoveBaseGoal goal;
goal.target pose.pose.position = target_position;
goal.target pose.pose.orientation.w = 1.;
goal.target pose.header.frame id = costmap client .getGlobalFrameID();
goal.target pose.header.stamp = ros::Time::now();
move base client .sendGoal(
    goal, [this, target_position](
              const actionlib::SimpleClientGoalState& status,
              const move base msgs::MoveBaseResultConstPtr& result) {
      reachedGoal(status, result, target position);
    });
bool Explore::goalOnBlacklist(const geometry msgs::Point& goal)
constexpr static size t tolerace = 5;
costmap 2d::Costmap2D* costmap2d = costmap client .getCostmap();
for (auto& frontier_goal : frontier_blacklist_) {
  double x diff = fabs(goal.x - frontier goal.x);
  double y diff = fabs(goal.y - frontier goal.y);
  if (x_diff < tolerace * costmap2d->getResolution() &&
      y_diff < tolerace * costmap2d->getResolution())
```

```
void Explore::reachedGoal(const actionlib::SimpleClientGoalState& status,
                         const geometry msgs::Point& frontier goal)
ROS DEBUG("Reached goal with status: %s", status.toString().c str());
  frontier blacklist .push back(frontier goal);
oneshot = relative nh .createTimer(
    true);
void Explore::start()
exploring timer .start();
void Explore::stop()
move base client .cancelAllGoals();
exploring timer .stop();
ROS INFO("Exploration stopped.");
```

```
// .
```



Appendix K: frontier_search.h



```
FrontierSearch()
 * @param costmap Reference to costmap data to search.
FrontierSearch(costmap 2d::Costmap2D* costmap, double potential scale,
                double gain scale, double min frontier size);
 * Oparam position Initial position to search from
std::vector<Frontier> searchFrom(geometry msgs::Point position);
adjacent
 * @param initial cell Index of cell to start frontier building
 * Oparam reference Reference index to calculate position from
 * Oparam frontier flag Flag vector indicating which cells are already
marked
Frontier buildNewFrontier (unsigned int initial cell, unsigned int
reference,
                           std::vector<bool>& frontier flag);
```

```
@param frontier flag Flag vector indicating which cells are already
marked
bool isNewFrontierCell(unsigned int idx,
                        const std::vector<bool>& frontier flag);
 * @param frontier frontier for which compute the cost
double frontierCost(const Frontier& frontier);
costmap 2d::Costmap2D* costmap ;
double potential_scale_, gain_scale_;
};
```

Appendix L: frontier_search.cpp



```
namespace frontier_exploration
```

```
using costmap_2d::LETHAL_OBSTACLE;
using costmap_2d::NO_INFORMATION;
using costmap 2d::FREE SPACE;
```

```
FrontierSearch::FrontierSearch(costmap_2d::Costmap2D* costmap,
double potential_scale, double gain_scale,
double min_frontier_size)
```

- : costmap_(costmap)
- , potential scale (potential scale)
- , gain_scale_(gain_scale)
- , min_frontier_size_(min_frontier_size)

```
std::vector<Frontier> FrontierSearch::searchFrom(geometry_msgs::Point
position)
```

```
std::vector<Frontier> frontier list;
```

```
// Sanity check that robot is inside costmap bounds before searching
unsigned int mx, my;
if (!costmap_->worldToMap(position.x, position.y, mx, my)) {
    ROS_ERROR("Robot out of costmap bounds, cannot search for frontiers");
    return frontier_list;
```

```
}
```

```
// make sure map is consistent and locked for duration of search
std::lock_guard<costmap_2d::Costmap2D::mutex_t>
lock(*(costmap ->getMutex()));
```

```
map_ = costmap_->getCharMap();
size_x_ = costmap_->getSizeInCellsX();
size y = costmap ->getSizeInCellsY();
```

```
// initialize flag arrays to keep track of visited and frontier cells
std::vector<bool> frontier_flag(size_x_ * size_y_, false);
std::vector<bool> visited_flag(size_x_ * size_y_, false);
```

```
unsigned int clear, pos = costmap ->getIndex(mx, my);
if (nearestCell(clear, pos, FREE SPACE, *costmap )) {
 bfs.push(clear);
  bfs.push(pos);
  ROS WARN ("Could not find nearby clear cell to start search");
visited flag[bfs.front()] = true;
while (!bfs.empty()) {
  unsigned int idx = bfs.front();
  bfs.pop();
  for (unsigned nbr : nhood4(idx, *costmap )) {
    if (map [nbr] <= map [idx] && !visited flag[nbr]) {</pre>
      visited flag[nbr] = true;
      bfs.push(nbr);
free
    } else if (isNewFrontierCell(nbr, frontier flag)) {
      frontier flag[nbr] = true;
      Frontier new frontier = buildNewFrontier(nbr, pos, frontier flag);
      if (new frontier.size * costmap ->getResolution() >=
        frontier list.push back(new frontier);
```
```
frontier.cost = frontierCost(frontier);
std::sort(
     frontier_list.begin(), frontier_list.end(),
     [](const Frontier& f1, const Frontier& f2) { return f1.cost <
f2.cost; });
return frontier list;
Frontier FrontierSearch::buildNewFrontier(unsigned int initial cell,
                                         unsigned int reference,
                                         std::vector<bool>& frontier flag)
Frontier output;
output.centroid.x = 0;
output.centroid.y = 0;
output.size = 1;
output.min distance = std::numeric limits<double>::infinity();
unsigned int ix, iy;
costmap ->indexToCells(initial cell, ix, iy);
costmap ->mapToWorld(ix, iy, output.initial.x, output.initial.y);
std::gueue<unsigned int> bfs;
bfs.push(initial cell);
unsigned int rx, ry;
double reference x, reference y;
costmap ->indexToCells(reference, rx, ry);
costmap ->mapToWorld(rx, ry, reference x, reference y);
while (!bfs.empty()) {
  unsigned int idx = bfs.front();
```

```
bfs.pop();
```

```
for (unsigned int nbr : nhood8(idx, *costmap )) {
     if (isNewFrontierCell(nbr, frontier flag)) {
      frontier flag[nbr] = true;
      unsigned int mx, my;
      double wx, wy;
      costmap ->indexToCells(nbr, mx, my);
      costmap ->mapToWorld(mx, my, wx, wy);
      geometry msgs::Point point;
      point.x = wx;
      point.y = wy;
      output.points.push back(point);
      output.size++;
       output.centroid.x += wx;
      output.centroid.y += wy;
       double distance = sqrt(pow((double(reference x) - double(wx)), 2.0)
                              pow((double(reference y) - double(wy)),
2.0));
       if (distance < output.min distance) {</pre>
         output.min distance = distance;
        output.middle.x = wx;
        output.middle.y = wy;
      bfs.push(nbr);
```

```
output.centroid.x /= output.size;
output.centroid.y /= output.size;
return output;
bool FrontierSearch::isNewFrontierCell(unsigned int idx,
                                      const std::vector<bool>&
frontier flag)
if (map [idx] != NO INFORMATION || frontier flag[idx]) {
neighbourhood
for (unsigned int nbr : nhood4(idx, *costmap )) {
  if (map_[nbr] == FREE_SPACE) {
double FrontierSearch::frontierCost(const Frontier& frontier)
return (potential scale * frontier.min distance *
        costmap ->getResolution()) -
        (gain_scale_ * frontier.size * costmap_->getResolution());
```

Appendix M: movement.h

```
#ifndef TEAM MOVEMENT HEADER
#define TEAM MOVEMENT HEADER
#include <ros/console.h>
#include "ros/ros.h"
#include <geometry msgs/Twist.h>
#include <nav msgs/Odometry.h>
#include <tf/transform datatypes.h>
#include <time.h>
#include "explore.h"
#define RAD2DEG(rad) ((rad) * 180. / M_PI)
#define DEG2RAD(deg) ((deg) * M PI / 180.)
extern float posX, posY, yaw;
extern float angular;
extern float linear;
extern const float SLOW SPIN;
extern const float FAST SPIN;
extern const float SLOW MOVE;
extern const float FAST MOVE;
extern const float MAX LIN;
extern const float MAX ROT;
void odomCallback (const nav msgs::Odometry::ConstPtr& msg);
bool monitorMotion ();
void setMotion(double dist, double linSpeed, double rot, double rotSpeed);
bool checkIfMoved(explore::Explore &explorer); // Monitors if the robot is
```

```
* @param linSpeed Linear velocity (determines forwards or back)
* @param rot Rotational displacement in radians (magnitude)
* @param rotSpeed Rotational velocity rad/s (+ive is left)
* @return If the movement is complete, true.
*
* @note It knows when you call it the first time or change course, so
long as the parameters change.
* (Calling travel(1,1,0,0) and then travel(1,1,0,0) again after that is
completed will not register
* as a new "move", these need to be seperated by a unique call such as
travel(0,0,0,0) to repeat)
*/
bool travel(double dist, double linSpeed, double rot, double rotSpeed);
/** @name setHeading
* @brief Will align the robot optimally with some absolute heading.
* @param heading The absolute heading (rad) to aim for
* @param speed Magnitude of rotation (rad)
* @return If the movement is complete, true.
* */
bool setHeading(float heading, float speed);
#endif
```

Appendix N: movement.cpp

#include "movement.h"

```
// Movement values
float angular = 0.0; // Global
float linear = 0.0; // Global
// Constants
const float SLOW_SPIN = (M_PI/12);
const float FAST_SPIN = (M_PI/6);
const float SLOW_MOVE = 0.1;
const float FAST_MOVE = 0.25;
const float MAX_LIN = 0.25;
const float MAX_ROT = (M_PI/6);
// Odometery values
```

```
float posX = 0.0, posY = 0.0, yaw = 0.0;
float distanceRemaining, prevX, prevY;
float rotationRemaining, prevYaw;
float rotMaintain, linMaintain;
void odomCallback (const nav msgs::Odometry::ConstPtr& msg)
  posX = msg->pose.pose.position.x;
  posY = msg->pose.pose.position.y;
  yaw = tf::getYaw(msg->pose.pose.orientation);
bool monitorMotion ()
  angular = rotMaintain;
  linear = linMaintain;
  float dx = posX - prevX;
  float dy = posY - prevY;
  float displacement = sqrt((dx * dx) + (dy * dy));
  prevX = posX;
  prevY = posY;
  distanceRemaining = distanceRemaining - displacement;
  if (distanceRemaining <= 0.0) linear = 0; // Stop
  displacement = std::abs(yaw - prevYaw);
  if (displacement > M PI) {
```

```
displacement = (M PI - abs(yaw)) + (M PI - abs(prevYaw));
  rotationRemaining = rotationRemaining - displacement;
  if (rotationRemaining <= 0.0) angular = 0; // Stop when we've rotated
  prevYaw = yaw;
  rotMaintain = angular;
  linMaintain = linear;
  bool doneMotion = (angular == 0) && (linear == 0);
  if (doneMotion) ROS DEBUG("Reached destination.");
       ROS DEBUG("In motion: D:%.2f S:%.2f | A:%.0f S:%.0f",
distanceRemaining,
       linear, RAD2DEG(rotationRemaining), RAD2DEG(angular));
  return doneMotion;
void setMotion(double dist, double linSpeed, double rot, double rotSpeed)
  distanceRemaining = std::abs(dist);
  rotationRemaining = std::abs(rot);
  prevX = posX;
  prevY = posY;
  prevYaw = yaw;
  linear = linSpeed;
  angular = rotSpeed;
  rotMaintain = angular;
  linMaintain = linear;
```

```
distanceRemaining, linear,
      RAD2DEG(rotationRemaining), RAD2DEG(angular));
bool travel(double dist, double linSpeed, double rot, double rotSpeed)
   static double pd = 0, pls = 0, pr = 0, prs = 0; // Used to store
previous state of inputs
  bool repeatedCall = false;
   if ((dist == pd) && (linSpeed == pls) && (rot == pr) && (rotSpeed ==
prs)) repeatedCall = true;
  pd = dist;
  pls = linSpeed;
  pr = rot;
  prs = rotSpeed;
  bool doneMotion = false;
   if (repeatedCall == true) {
      doneMotion = monitorMotion();
      setMotion(dist, linSpeed, rot, rotSpeed);
   return doneMotion;
bool setHeading(float heading, float speed)
  static float lastHeading = 0, lastSpeed = 0; // Used to monitor
  static float change = 0, rotVelocity = 0; // Internal parameters to
describe the change
```

```
bool doneAlignment = false;
  if ((lastHeading == heading) && (lastSpeed == speed)) {
      doneAlignment = travel(0, 0, change, rotVelocity);
      lastHeading = heading;
      lastSpeed = speed;
       float changeForward, changeBackward, tempHeading;
      if (heading < yaw) tempHeading = heading + 2 * M PI; // Loop around
      else tempHeading = heading;
       changeForward = tempHeading - yaw;
       if (heading > yaw) tempHeading = heading - 2 * M PI; // Loop back
       changeBackward = yaw - tempHeading;
RAD2DEG(changeForward), RAD2DEG(changeBackward));
      if (changeBackward < changeForward) {</pre>
          rotVelocity = -speed;
          change = changeBackward;
Change of %.0f right.",
               RAD2DEG(heading), RAD2DEG(yaw), RAD2DEG(change));
           rotVelocity = speed;
```

```
change = changeForward;
               RAD2DEG(heading), RAD2DEG(yaw), RAD2DEG(change));
      doneAlignment = travel(0, 0, change, rotVelocity);
  if (doneAlignment) ROS INFO("Aligned with new heading.");
  return doneAlignment;
bool checkIfMoved(explore::Explore &explorer) {
  const float minMotion = 1.0;
  const int recordingStep = 1; // Time in seconds between location
  const int numRecordings = 10; // Number of recordings to store
  static time t targetTime = time(NULL);
  static bool didItMove = true;
  static bool alreadyStuck = false;
  static int curRecording = 0; // Current point in buffer
  static float xRec[numRecordings], yRec[numRecordings];
  time t presentTime = time(NULL);
  if (targetTime < presentTime) {</pre>
       targetTime = presentTime + recordingStep; // Set next time
       xRec[curRecording] = posX;
       yRec[curRecording] = posY;
```

```
float travelDist = 0.0;
    float temp, dx, dy;
       dx = xRec[numRecordings - 1] - xRec[0];
       dy = yRec[numRecordings - 1] - yRec[0];
       i = numRecordings; // Move i to end
       dx = xRec[i] - xRec[i - 1];
       dy = yRec[i] - yRec[i - 1];
   temp = dx * dx + dy * dy;
   travelDist += sqrt(temp);
curRecording++;
if (curRecording == numRecordings) {
   curRecording = 0;
    ROS INFO ONCE ("Done first pass of movement buffer");
    if (travelDist > minMotion) {
       didItMove = true;
       alreadyStuck = false;
       didItMove = false;
        if (alreadyStuck) {
```

Appendix O: victimLocator.py

```
#!/usr/bin/env python
from future import print function
import os
import torch
import rospy
import roslib # Needed to import ModelStates
import random
import argparse
import numpy as np
import pandas as pd
from geometry_msgs.msg import Pose
from gazebo msgs.msg import ModelState
from gazebo_msgs.msg import ModelStates
from visualization msgs.msg import Marker
from visualization msgs.msg import MarkerArray
from mie443 contest3.msg import EmotionFaceMsg
def getInputArgs():
  parser = argparse.ArgumentParser('MIE443 contest3 victim detector.')
  parser.add argument('--victim file', dest='victim file',
default='victims.csv', type=str, help='Locations of the victims in the
```

```
parser.add argument('--emotion file', dest='emotion file',
default='train split.pth', type=str, help='Emotion detection file')
   parser.add argument('--n face samples', dest='n face samples',
default=10, type=int, help='Emotion detection file')
   args = parser.parse args()
   return args
class VictimLocations(object):
  def init (self, args):
      victims = pd.read csv(os.path.abspath(args.victim file))
       victims.columns = victims.columns.str.strip()
       self.pub victim count = 0
radius']).issubset(set(victims.columns.astype(str).tolist()))
       self.results file = open('detectedVictim.txt', 'w')
       self.victim pose = victims.as matrix(['pos x', 'pos y'])
       self.victim det rad = victims.as matrix(['radius']).reshape(-1)
       self.publish victims(self.victim pose, self.victim det rad)
       self.n face samples = args.n face samples
       self.emotion lists, self.possible emotions =
self.loadEmotion(args.emotion file)
       random.shuffle(self.possible emotions)
       rospy.on shutdown(self.logVictimHistory)
       self.gazebo sub = rospy.Subscriber('/gazebo/model states',
ModelStates, self.stateSub)
       self.emotion pub = rospy.Publisher('/emotion img', EmotionFaceMsg,
queue size=1)
       while self.emotion pub.get num connections() == 0:
           print('Waiting for connection for emotion pub.')
           rospy.sleep(0.5)
```

def publish_victims(self, pose, rad):

```
markerPub = rospy.Publisher('/gazebo/set model state', ModelState,
queue size=10)
       while markerPub.get num connections() == 0:
           print('Waiting for connection for marker pub.')
           rospy.sleep(0.5)
      msg = ModelState()
      msg.reference frame = 'world'
       for v idx in range(pose.shape[0]):
           msg.model name = 'box contest3 clone ' + str(v idx)
          msg.pose.position.x = pose[v idx][0]
          msg.pose.position.y = pose[v idx][1]
          msg.pose.position.z = -0.999
          msg.pose.orientation.x = -0.5
          msg.pose.orientation.y = -0.5
          msg.pose.orientation.z = -0.5
          msg.pose.orientation.w = 0.5
          markerPub.publish(msg)
  def loadEmotion(self, emotion file path):
       imgs, labels = torch.load(os.path.abspath(emotion file path))
       unique emotions = labels.unique()
       emotion imgs = \{\}
       for 1 in unique emotions:
           emotion mask = labels == l
           emotion_imgs[l.item()] = imgs[emotion mask].numpy()
       return emotion imgs, unique emotions.numpy().tolist()
  def robotLocation(self, msg):
       assert isinstance(msg, Pose)
       return np.array([msg.position.x, msg.position.y])
  def pubEmotionSet(self, gt emotion idx):
      print('Publishing emotion images')
       emo imgs = self.emotion lists[gt emotion idx]
       selected emo imgs = emo imgs[np.random.choice(emo imgs.shape[0],
self.n face samples, replace=False), :]
       print(selected emo imgs.shape)
       emo msg = EmotionFaceMsg()
       emo msg.width = 48
       emo msg.height = 48
```

```
emo msg.batch = self.n face samples
      emo msg.data = selected emo imgs.reshape(-1,1,1)
      while self.emotion pub.get num connections() == 0:
          print('Waiting for connection for emotion pub.')
          rospy.sleep(0.05)
      self.emotion pub.publish(emo msg)
      print('Published emotion images')
  def stateSub(self, msg):
      if self.pub victim count < 200:
          self.publish victims(self.victim pose, self.victim det rad)
          self.pub victim count += 1
      for idx, name in enumerate(msg.name):
      rpose = self.robotLocation(msg.pose[idx])
      vic dist = np.sqrt(np.sum((rpose - self.victim pose) ** 2,
-1)).reshape(-1)
      vic detected = vic dist < self.victim det rad</pre>
      detected poses = self.victim pose[vic detected]
      self.victim det rad = self.victim det rad[~vic detected]
      self.victim pose = self.victim pose[~vic detected]
      for pose in detected poses:
          gt emotion idx = self.possible emotions.pop(0)
          print('Sending emotion:', gt emotion idx)
          detected victim = [pose, gt emotion idx]
          self.pubEmotionSet(gt emotion idx)
          self.results file.write(str(detected victim))
```

```
def logVictimHistory(self):
```

self.results_file.close()

if __name__ == "__main__":
 rospy.init_node('victimLocator')
 args = getInputArgs()
 victim_locations = VictimLocations(args)
 rospy.spin()

7.0 References

[1] *MIE443H1S: Contest 3: Finding and Interacting with Emotional People in an Unknown Environment*, University of Toronto, Toronto, ON, 2021, pp. 1-3.