<pre>lamda = 0.07; delta = 1.2; angle_start = atan2(0,1); position = [x_start, y_start]; time = 1;</pre>	
<pre>%Equations for partial derivative of Z [u, v] = gradient(Z); pub = rospublisher('raw_vel'); % stop the robot if it's going right now stopMsg = rosmessage(pub); stopMsg.Data = [0 0]; send(pub, stopMsg);</pre>	
<pre>placeNeato(x_start, y_start, 1,0); % wait a bit for robot to fall onto the bridge pause(2); % time to drive!!</pre> <pre>%making the Neato go based off of calculated velocities</pre>	
<pre>%setting up message that send left and right wheel velocities velMsg = rosmessage(pub); distance_traveled = 0; start = rostime('now'); tic while 1 current = rostime('now'); elapsed = current - start;</pre>	
<pre>% get the current time from ROS if elapsed.seconds >= time start = current; %Equation for the direction of travel used_indices=[used_indices, nrstx]; [-,nrstx] = min(abs(X(:)-x_start)); [-,nrsty] = min(abs(Y(:)-y_start)); f_grad=[fx(nrsty,nrstx) fy(nrsty,nrstx)];</pre>	
<pre>f_grad=f_grad./norm(f_grad); %finding change in position move = f_grad*lamda; %Calculate Total Distance Traveled delta_p = sqrt(move(1).^2+move(2).^2) distance_traveled = distance_traveled + delta_p</pre>	
<pre>%finding speed r_speed = norm(move/(time)); %calculating direction of gradient new_angle = atan2(f_grad(2), f_grad(1)); %finding angular velocity delta_w = (new_angle - angle_start); w_speed = delta_w/time; angle_start = new_angle;</pre>	
%finding linear velocities of each wheel V_L = r_speed - (w_speed*(d/2)); V_R = r_speed + (w_speed*(d/2)); velMsg.Data = [V_L, V_R]; % set wheel velocities at specific time send(pub, velMsg); % send new wheel velocities	
<pre>%calculating new position position = position + move; x_start = position(1); y_start = position(2); lamda = delta *lamda; if norm(f_grad*lamda) > 0.63 %stoping the Neato at the end of the path send(pub, stopMsg); total_time_elasped = toc</pre>	
break %leave this loop once we have reached the stopping time end end distance_traveled; total_time_elasped;	
<pre>load gauntlet.mat timeframe=dataset(:,1); leftpos=dataset(:,2); rightpos=dataset(:,3); %Using the equation of velocity=position/time, finding the velocities of %the left and right wheel. leftvelo=diff(leftpos)./diff(timeframe); rightvelo=diff(rightpos)./diff(timeframe);</pre>	
%With the velocities of each wheel, we can calculate the linear and angular %velocity velo_exp=(leftvelo+rightvelo)./2; angular_exp=(rightvelo-leftvelo)./0.235; %Setting the starting position and heading of the NEATO bridgeStart = [0,0,0];	
startinggrad = [1,0,0]; %Heading matrix will collect the heading throughout the simulation and %Position matrix will collect the position throughout the simulation heading=startinggrad; position=bridgeStart; %new_position will consistently update the position and angle variable will %consistently update the angle	
<pre>new_position=position; angle=0; n=1; while n < 154 if n ==1; new_angle=angle+angular_exp(n,:)*timeframe(1);</pre>	
<pre>new_heading=[startinggrad(:,1)*cos(new_angle)-startinggrad(:,2)*sin(new_angle), startinggrad(:,1)*sin(new_angle)+startinggrad(:,2)*cos(new_angle),0]; new_position=new_position+velo_exp(n,:)*new_heading*timeframe(1); else new_angle=angle+angular_exp(n,:)*(timeframe(n)-timeframe(n-1)); new_heading=[startinggrad(:,1)*cos(new_angle)-startinggrad(:,2)*sin(new_angle), startinggrad(:,1)*sin(new_angle)+startinggrad(:,2)*cos(new_angle),0]; new_position=new_position+velo_exp(n,:)*new_heading*(timeframe(n)-timeframe(n-1)); end position=cat(1,position,new_position); heading=cat(1,heading,new_heading); angle=new_angle:</pre>	
<pre>angle=new_angle; n=n+1; end %3D plot of the experimental path p=position; figure() plot3(p(:,1),p(:,2),p(:,3))</pre>	
1 0.5	
%Plotting the predicted and actual path with several unit tangent vectors	
<pre>figure() plot(p(:,1),p(:,2),''), hold on for i=1:size(everyposition,1) quiver(everyposition(i,1),everyposition(i,2),everymove(i,1),everymove(i,2)) end hold off title('Predicted vs Actual Path')</pre>	
xlabel('X position') ylabel('Y position') legend('Actual Path', 'Predicted Path') Predicted vs Actual Path O.5 Predicted vs Actual Path — Actual Path — Predicted Path — Predicted Path	
-0.5 Formula 1 Solution 1 Solution 2 -1.5	
-2.5 -3 0.1 0.2 0.3 0.4 0.5 0.6 0.7 0.8 0.9 1 X position	
otal time elapsed for robot to reach BoB: 14.65 seconds otal distance traveled: 3.47m overall, the system worked well. The gradient was set out very clean because of connecting te obstructions in one single pen, and the NEATO never had to move to unnessesary locations due to a mishap between obstacles. The contour looks very nice, a experimental data above shows that the gradient and the actual path opf the NEATO is a little different, but this is because the gradient never considered the distance between wheels or even the size of the robot - this all lead to actual physics being a little	
function [fitline_coefs,bestInlierSet,bestOutlierSet,bestEndPoints]= robustLineFit(r,theta,d,n,visualize) %The [fitline_coefs,bestInlierSet,bestOutlierSet,bestEndPoints]= robustLineFit(r,theta,d,n) %function runs the RANSAC algorithm for n candidate lines and a threshold of d. The inputs r and %theta are polar coordinates. The output fitline_coefs are the coefficientsmatlab:matlab.internal.language.introspective.errorDocCallback('TheGauntlet>robustLineFit', 'C:\Users\mkang1\De %of the best fit line in the format [m b] where y=m*x+b. If you want %to visualize, set visualize flag to 1, off is 0. Default is true.	esktop\QEA\Robo\Gauntlet\TheGauntlet.mlx'
<pre>if -exist('visualize','var') % visualize parameter does not exist, so default it to 1 visualize = 1; end %eliminate zeros index=find(r~=0 & r<3); r_clean=r(index); theta_clean=theta(index);</pre>	
%convert to Cartesian and plot again for verification [x,y]=pol2cart(deg2rad(theta_clean),r_clean); points=[x,y]; %now let's actually implement the RANSAC algorithm bestcandidates = []; bestInlierSet = zeros(0,2); bestOutlierSet = zeros(0,2);	
<pre>bestEndPoints = zeros(0,2); for k=1:n %number of candidate lines to try %select two points at random using the 'datasample' function to define %the endpoints of the first candidate fit line candidates = datasample(points, 2, 'Replace', false); %Find the vector that points from point 2 to point 1</pre>	
<pre>v=(candidates(1,:)-candidates(2,:))'; %Check the length of the vector v. If it is zero, the datasample %function chose the same point twice, and we need to resample. The %continue command will pass to the next iteration of the for loop. if norm(v) == 0 continue; end</pre>	
%Determine whether points are outliers, we need to know the %perpendicular distance away from the candidate fit line. To do this, %we first need to define the perpendicular, or orthogonal, direction. orthv= [-v(2); v(1)]; orthv_unit=orthv/norm(orthv); %make this a unit vector %Here, we are finding the distance of each scan point from one of the %endpoints of our candidate line. At this point this is not the %distance perpendicular to the candidate line.	
diffs = points - candidates(2,:); %Next, we need to project the difference vectors above onto the %perpendicular direction in 'orthv_unit'. This will give us the %orthogonal distances from the canidate fit line. orthdists=diffs*orthv_unit; %To identify inliers, we will look for points at a perpendicular %distance from the candidate fit line less than the threshold value.	
%The output will be a logic array, with a 1 if the statement is true %and 0 if false. inliers=abs(orthdists) < d; %we also want to check that there are no big gaps in our walls. To do %this, we are first taking the distance of each inlier away from an %endpoint (diffs) and projecting onto the best fit direction. We then %sort these from smallest to largest and take difference to find the %spacing between adjacent points. We then identify the maximum gap.	
<pre>biggestGap = max(diff(sort(diffs(inliers,:)*v/norm(v)))); %Now, we check if the number of inliers is greater than the best we %have found. If so, the candidate line is our new best candidate. We %also make sure there are no big gaps. if biggestGap < 0.2 && sum(inliers) > size(bestInlierSet,1) if sum(inliers) > size(bestInlierSet,1) bestInlierSet=points(inliers,:); %points where logical array is true bestOutlierSet = points(~inliers,:); %points where logical array is not true</pre>	
bestcandidates=candidates; %these two lines find a nice set of endpoints for plotting the best %fit line projectedCoordinate = diffs(inliers, :)*v/norm(v); bestEndPoints = [min(projectedCoordinate); max(projectedCoordinate)]*v'/norm(v) + repmat(candidates(2, :), [2, 1]); end end	
<pre>if isempty(bestEndPoints) m= NaN; b= NaN; bestEndPoints=[NaN, NaN; NaN, NaN]; fitline_coefs=[m b]; return; end</pre>	
<pre>%Find the coefficients for the best line m=diff(bestEndPoints(:,2))/diff(bestEndPoints(:,1)); b=bestEndPoints(1,2)-m*bestEndPoints(1,1); fitline_coefs=[m b]; if visualize==1 %plot the polar data as verification hold off;</pre>	
figure(1) polarplot(deg2rad(theta_clean),r_clean,'ks','MarkerSize',6,'MarkerFaceColor','m') title('Visualization of Polar Data') figure(2) plot(x,y,'ks') title('Scan Data- Clean') xlabel('[m]')	
<pre>ylabel('[m]') %Now we need to plot our results figure(3) plot(bestInlierSet(:,1), bestInlierSet(:,2), 'ks') hold on plot(bestOutlierSet(:,1), bestOutlierSet(:,2), 'bs') plot(bestEndPoints(:,1), bestEndPoints(:,2), 'r')</pre>	
<pre>legend('Inliers', 'Outliers', 'Best Fit', 'location', 'northwest') title(['RANSAC with d=' num2str(d) ' and n=' num2str(n)]) xlabel('[m]') ylabel('[m]') % Create textbox annotation(figure(3), 'textbox', [0.167071428571429 0.152380952380952 0.25 0.1], 'String', {'Number of Inliers:' num2str(size(bestInlierSet,1))}, 'FitBoxToText', 'off');</pre>	
end function collectDataset_sim(datasetname) % This script provides a method for collecting a dataset from the Neato % sensors suitable for plotting out a 3d trajectory. To launch the	
% application run: % % collectDataset_sim('nameofdataset.mat') % % where you should specify where you'd like to the program to save the % the dataset you collect. % % The collected data will be stored in a variable called dataset. Dataset % will be a nx6 matrix where each row contains a timestamp, the encoder	
<pre>% values, and the accelerometer values. Specifically, here is the row % format. % % [timestamp, positionLeft, positionRight, AccelX, AccelY, AccelZ]; % % To stop execution of the program, simply close the figure window. function myCloseRequest(src,callbackdata) % Close request function</pre>	
<pre>% to display a question dialog box % get rid of subscriptions to avoid race conditions clear sub_encoders; clear sub_accel; delete(gcf) end function processAccel(sub, msg) % Process the encoders values by storing by storing them into</pre>	
<pre>% the matrix of data. lastAccel = msg.Data; end function processEncoders(sub, msg) % Process the encoders values by storing by storing them into % the matrix of data. if ~collectingData return;</pre>	
<pre>end currTime = rostime('now'); currTime = double(currTime.Sec)+double(currTime.Nsec)*10^-9; elapsedTime = currTime - start; dataset(encoderCount + 1,:) = [elapsedTime msg.Data' lastAccel']; encoderCount = encoderCount + 1;</pre>	
<pre>function keyPressedFunction(fig_obj, eventDat) % Convert a key pressed event into a twist message and publish it ck = get(fig_obj, 'CurrentKey'); switch ck case 'space' if collectingData collectingData = false; dataset = dataset(1:encoderCount, :); save(datasetname, 'dataset'); disp(!Stopping dataset, collection!);</pre>	
<pre>disp('Stopping dataset collection'); else start = rostime('now'); start = double(start.Sec)+double(start.Nsec)*10^-9; encoderCount = 0; dataset = zeros(100000, 6); collectingData = true; disp('Starting dataset collection'); end end</pre>	
<pre>end global dataset start encoderCount lastAccel; lastAccel = [0; 0; 1];</pre>	
<pre>title('Dataset Collection Window'); set(f,'WindowKeyPressFcn', @keyPressedFunction); end function [xs,ys]=connect_dots(x,y,i,j) x0=[x(i) x(j)];y0=[y(i) y(j)]; line=create_line_function([x(i) y(i)],[x(j) y(j)]); m=line(1);b=line(2);</pre>	
<pre>if abs(1/m) < 1 % Vertical line ys=min(y0):0.01:max(y0); xs=(ys-b)./m; else % Horizontal line xs=min(x0):0.01:max(x0); ys=xs.*m+b; end end</pre>	
<pre>function [r_clean, theta_clean]=trim_data(r, theta) % Filtering out 0's r_clean = r(r(:,1)>0 & r(:,1)<3,:); theta_clean = theta(r(:,1)>0 & r(:,1)<3,:); end function y=create_line_function(p1,p2) m = (p1(2) - p2(2)) / (p1(1) - p2(1)); b = p1(2) - (m * p1(1));</pre>	
<pre>y = [m, b]; end function placeNeato(posX, posY, headingX, headingY) svc = rossvcclient('gazebo/set_model_state'); msg = rosmessage(svc); msg.ModelState.ModelName = 'neato_standalone'; startYaw = atan2(headingY, headingX);</pre>	
<pre>startYaw = atan2(headingY, headingX); quat = eul2quat([startYaw 0 0]); msg.ModelState.Pose.Position.X = posX; msg.ModelState.Pose.Position.Y = posY; msg.ModelState.Pose.Position.Z = 1.0; msg.ModelState.Pose.Orientation.W = quat(1); msg.ModelState.Pose.Orientation.X = quat(2); msg.ModelState.Pose.Orientation.Y = quat(3); msg.ModelState.Pose.Orientation.Z = quat(4);</pre>	

The Gauntlet Challenge (Part 2, Navigating the Gauntlet)

For our Gauntlet Challenge, we choice Level 3 as our problem to tackle. First off, we identified what obstacles there were and where the position of the NEATO, pre-determined by the assignment. Afterwards, the scanned data was turned into line

segments that was connected together for simpler obstruction recognition. The circle, however was identified using the circle fitting function and we recieved the centerpoint and the radius for the BoB. With these data, we created the contour lines by making a scalar feild with the obstructions being a sink and the BoB a source. With this, we were able to create a 3-D contour that showed where the NEATO would travel. In this case, the NEATO traveled using the steepest gradient ascent function, where it identifies the quickest way to clinb to the top of the contour. Since we set the obstruction as a sink and the BoB as a source, the we were able to create a path for the NEATO using the gradients. Then the NEATO was simulated using multipled by the step, which was put into a while loop until it finished the course. The data of the wheels were then collected and put

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