Real-time implementation of an asynchronous vision-based target tracking system in an unmanned aerial vehicle

Schenk, Michael A.

Monterey  California. Naval Postgraduate School
REAL-TIME IMPLEMENTATION OF AN ASYNCHRONOUS VISION-BASED TARGET TRACKING SYSTEM IN AN UNMANNED AERIAL VEHICLE

by

Michael A. Schenk

June 2007

Thesis Advisor: Isaac I. Kaminer
Second Reader: Vladimir N. Dobrokhodov

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**Author(s)**: Michael A. Schenk  

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**Abstract**: Currently, small unmanned aerial vehicles developed by NPS have been able to locate and track stationary and moving targets on the ground. New methods of continuous target tracking are always being developed to improve speed and accuracy, ultimately aiding the user of the system. This thesis describes one such method, utilizing an open loop filter as well as an external correction source: Perspective View Nascent Technologies (PVNT). While the PVNT correction can theoretically improve the accuracy from 20-30 meters to 1-2 meters, it does have a disadvantage in that the target position updates are delayed anywhere from 1-10 seconds. In order to account for the delay, an asynchronous filter is used to update the target position data given the external position correction from PVNT. Two cases have been tested including the general filter and one that utilizes a road model in the calculations. While an earlier thesis developed the basic simulation for the system, this thesis discusses improvements and corrections to the simulation model as well as the necessary steps for real-time implementation.  

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Michael A. Schenk
Ensign, United States Navy
B.S., United States Naval Academy, 2007

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June 2007

Author: Michael A. Schenk

Approved by: Isaac I. Kaminer
Thesis Advisor

Vladimir N. Dobrokhodov
Second Reader

Anthony J. Healey
Chairman, Department of Mechanical and Astronautical Engineering
ABSTRACT

Currently, small unmanned aerial vehicles developed by NPS have been able to locate and track stationary and moving targets on the ground. New methods of continuous target tracking are always being developed to improve speed and accuracy, ultimately aiding the user of the system. This thesis describes one such method, utilizing an open loop filter as well as an external correction source: Perspective View Nascent Technologies (PVNT). While the PVNT correction can theoretically improve the accuracy from 20-30 meters to 1-2 meters, it does have a disadvantage in that the target position updates are delayed anywhere from 1-10 seconds. In order to account for the delay, an asynchronous filter is used to update the target position data given the external position correction from PVNT. Two cases have been tested including the general filter and one that utilizes a road model in the calculations. While an earlier thesis developed the basic simulation for the system, this thesis discusses improvements and corrections to the simulation model as well as the necessary steps for real-time implementation.
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I. INTRODUCTION

A. OVERVIEW

The goal of the work in this thesis is to contribute to continuous improvements that are being made in the area of vision-based target tracking and motion estimation. Changes to the current system allow the entire process to be faster, more accurate, and more user-friendly. Improvements to the technology can be simulated using computer programs such as MATLAB and Simulink, implemented, and then field tested during scenarios run by NPS during the Tactical Network Topology (TNT) sessions. Ultimately, the systems developed here can make their way to military use in surveillance and reconnaissance missions. Concepts examined in this thesis include the development and implementation non real-time and real-time target motion estimation systems as well as asynchronous target tracking filters with and without road following capabilities.

B. BACKGROUND

Several important tools are discussed that play prominent roles in the developed systems in this thesis. The PVNT position update system is described first, followed by the background on the asynchronous constant gain Kalman filter.

1. Perspective View Nascent Technologies (PVNT)

One of the problems with incorporating detailed terrain maps into real-time systems is the large amount of required data storage and equally large amount of necessary computing power needed to deal with the loading and retrieval of map sections. Developed by Dr. Wolfgang Baer, the PVNT system offers a low-cost alternative available on a personal computer. The PVNT system begins with terrain data collected by the National Imagery and Mapping Agency (NIMA) and contains tools that allow updates to be included from local measurement devices and other sensors [1]. The inclusion of this new data results in a more accurate terrain mapping system than can be more efficiently updated to reflect terrain variations rather than creating entirely new maps.
Another major advantage PVNT has over other scene-visualization programs is that the terrain data are stored using raster formats (pixels) instead of using a polygon database [1]. This makes implementation of the system using PVNT combined with remote sensor arrays in real-time much more practical.

Tests conducted at Camp Roberts, CA, depict how PVNT works hand-in-hand with a vision-based target tracking system. Initially, the target is acquired and the gimbal-mounted camera passes data to image processing software. Then, open-loop, non-linear filters are able to estimate the target position and resulting velocity. After around 20 seconds of tracking, the accuracy for this portion of the system is within about 10-20 meters. The PVNT software then compares data coming in from GPS, camera angle values, and the images from the UAV camera to the terrain database for the area. Since the accuracy of the database has roughly a one meter resolution, the accuracy of the position update from PVNT can be ten times more accurate than the non-linear filter estimation. However, because of the multiple data inputs and necessary image comparison between the camera and terrain database, the required processing time results in a delay up to ten seconds before the position update is delivered to the system [6].

2. Asynchronous Constant Gain Kalman Filter

It is in target motion estimation that the Kalman filter can be employed. One of the reasons that the Kalman filter works well with target tracking applications is its ability to compare and integrate data from multiple sensors (such as a position update with estimated target velocity and estimated position) to give the most accurate result. However, standard Kalman filters are hindered by the fact that they must have evenly-spaced data inputs and updates for maximum effectiveness. The filter runs into accuracy problems when data arrives at different sampling rates or delayed times.

The asynchronous constant gain Kalman filter was developed because of the need for an accurate estimation tool in a system with delayed data inputs. It is the preferred filter to ensure better system robustness and overall result accuracy because the asynchronous version of the filter is able to accept out of synchronization data entries from sensors. Thus, the asynchronous constant gain Kalman filter is a better match for
this target tracking system since the data from the PVNT update is delayed anywhere from one to ten seconds before being entered into the filter [6].

C. THESIS DESCRIPTION

Chapter I presents a general overview of the work of the thesis with respect to UAV target tracking capabilities as well as background for two of the main tools utilized in the thesis: the PVNT update system and asynchronous constant gain Kalman filter. The next chapter will outline, step-by-step, the process that takes place during target motion estimation with and without position updates. The chapter also discusses the general and road following filters; the two different styles of filters that are employed in the real- and non real-time models. Chapter III will briefly review the current non real-time general filter model for target tracking presented in an earlier thesis and then develop a road following version of the model for non real-time simulation in Simulink. Additionally, the chapter will discuss the steps needed to convert the non real-time models into real-time models along with the actual modeling of the real-time general and road following filters in Simulink. Chapter IV will go over the system parameters and actual simulation of the non real-time and real-time models. While the non real-time road following model will be tested using a single road model to ensure proper function, the real-time general and road following models will be tested with numerous simulated roads and varying input errors. All of the necessary results will follow in the chapter along with explanations for the response of the system to different scenarios. Finally, Chapter V will present the conclusions from the thesis results and recommendations for future work in the field of study.
II. SYSTEM STRUCTURE

This chapter provides flow charts and diagrams describing the processes that take place in the filter operations with and without PVNT updates. The final section also explains the objectives for the thesis.

A. SEQUENCE OF OPERATION

The first task is to organize the order of the processes within the target tracking system. Then, the improved system with the general linear filter must be altered to incorporate the PVNT position updates in parallel with the normal operation.

1. Linear Filter with PVNT Update

![Diagram of Linear Filtering with PVNT Update]

*Figure 1. Linear filtering with PVNT update*
Figure 1 shows the asynchronous filtering system with the addition of the PVNT position update. After the necessary image processing by the PVNT software, the new estimated target position is fed into the asynchronous constant gain Kalman filter. The asynchronous filter then performs the required calculations, rewriting over the data previously stored during the delay, and outputs a new estimated target position to the original non-linear filter. Figure 2 below shows how the PVNT processing and asynchronous constant gain Kalman filter relates to the time interval for the system.

![Diagram of PVNT processing and Kalman filtering with respect to time](image)

**Figure 2.** PVNT processing and Kalman filtering with respect to time

**B. GENERAL AND ROAD FOLLOWING FILTER DIFFERENCES**

The two different style filters tested in this thesis are the general and road following filters. Both filters receive PVNT updates and perform target motion estimation along road models during the simulations. The difference, however, is that the road following filter uses the road model equations in the target motion estimation calculations while the general filter does not. This allows the $x, y, z$ coordinates used by the general filter model to be simplified in the road following model by a road parameter, $\rho$. This concept is discussed in greater detail in Chapter III.
The figure below depicts the effects of the road following parameter by comparing the position estimates of an open loop (OL) filter and an asynchronous road following filter (AF) with PVNT along a sample road profile. The filter estimates are identical during straight portions of the road, but the road following filter provides much better position estimates during areas of greater curvature. The position error is due to velocity estimation error, but the estimate still lies along the road profile.

**Figure 3. Target Estimation Comparison for Open Loop and Road Following Filters**

C. OBJECTIVES

While the basic non real-time filter design is discussed in the master’s thesis in Reference 6, numerous improvements and corrections were needed to make the non real-time road following filter perform correctly. The generation and storage of the PVNT update will be changed and the asynchronous constant gain Kalman filter will be modeled in Simulink. Additionally, the non real-time general and road following filter models will be converted into systems capable of real-time implementation. The two real-time filter systems will also be extensively simulated and the results analyzed to determine conditions of peak and poor performance.
III. MODELING

This chapter details the modeling process for the non real-time and real-time systems. Included in the sections are models for the general and road following filters for each style system as well as development of the road equations. The process of converting the non real-time models into models that can be implemented in real-time systems is also cited with major focus placed on the S-function and its capabilities.

A. NON REAL-TIME MODELING

Before models can be implemented in a real-time system, non real-time models had to be produced. These non real-time models serve as a starting point for the development of the real-time filter.

1. General Filter

The general asynchronous filter and MATLAB code is found in Reference 6 and models the general target tracking filter incorporating a delayed PVNT update. Many of the components for the road following version of the filter are similar and will be discussed in the next section.

2. Road Following Filter with Separate Model File Integration

The main addition to the road following asynchronous filter is the parameter \( \rho \), which defines the road along which the target is moving. Since the target plane is assumed to be two dimensional, \( \rho \) relates to the x and y coordinates of the target while z relates to local altitude. The road following asynchronous filter was corrected from Reference 6 due to errors in the MATLAB code concerning data storage and retrieval. Portions of the Simulink model were also adjusted to correctly generate and pass on the PVNT position update to later tasks in the system. The updated road following filter system is shown in Figure 4.
Figure 4. **Road following asynchronous filter**

Next, the Simulink diagram is broken down with a description of the function of each component in the overall system.

*a. True Target Motion with Road Following Characteristics*

Figure 5. **True target position and velocity generation with road following**
The non real-time system begins with the target of constant velocity and integrating to calculate $\rho$. The new $\rho$ value is then inputted into a MATLAB function block that calculates $x$, $y$, and $z$ target position based on a predetermined equation for the road. For the preliminary simulation tests, the equation for the road based on the parameter $\rho$ was taken from Reference 6:

$$
\begin{align*}
  x &= \rho \\
  y &= 0.0000192 \rho^3 - 0.025 \rho^2 + 9.74 \rho \\
  z &= 0
\end{align*}
$$

(1) [Ref. 6]

The above system of equations produces a road profile that is depicted in the next figure.

![Simulated Road Profile](image)

**Figure 6.** Simulated Road Profile [After Ref. 6]

This target data along with the $\rho$ values are stored for later use as the simulated PVNT update. For the Simulink model, it is more practical to assume that the PVNT estimate, with one to two meter accuracy, can be simulated by taking an actual target position from the true target model at time $\tau$ instead of trying to run PVNT software.
b. **PVNT Model**

![Diagram of PVNT update generator]

Figure 7. **PVNT update generator**

Figure 7 depicts the PVNT update block and its components for the non real-time road following model. The block receives and takes in the true target position and velocity from the target model along with the system time and a random PVNT update delay. The fifth input is an oscillating step signal that indicates when the PVNT update is active (signaling an update is ready to be sent to the asynchronous constant gain Kalman filter). Additionally, a PVNT input error can be included in the system to simulate the expected one to two meter accuracy of the device.

c. **Optimization Function**

The optimization function block is used to determine the parameter \( \rho \), and was not changed from the earlier thesis. The method of optimization that is used in the non real-time road following filter simply finds \( \rho \) that minimizes the distance from the inputted PVNT position update to the road. Once the minimum distance over a maximum range is found, the corresponding \( \rho \) value for the \( x \) and \( y \) road coordinates is passed on to the asynchronous filter.
d. **Asynchronous Filter**

The asynchronous filter portion of the diagram has multiple triggered subsystems.

![Asynchronous filter - Subsystem 1](image)

**Figure 8. Asynchronous filter – Subsystem 1**

The first subsystem calculates time $\tau$ once given the current system time $t$ and the PVNT delay time. The time $\tau$ is then passed on to the second subsystem along with the PVNT delay and the PVNT update $\rho^*$.

![Asynchronous filter - Subsystem 2](image)

**Figure 9. Asynchronous filter – Subsystem 2**
Figure 9 shows the data being passed through a mux block and into a MATLAB function block, which outputs the estimated position \( \hat{p} \) and velocity \( \hat{v} \) data for the target at time \( t \). It is important to note that this subsystem only runs when the PVNT update is present. The MATLAB function block refers to a function written in MATLAB code that actually performs the asynchronous double integration. The filter is actually contained in a separate Simulink model file and called by the MATLAB function.

![Simulink model](image)

**Figure 10. Separate Simulink model containing asynchronous filter**

The filter operates independently of the time of the overall system, allowing it to be asynchronous. The MATLAB code stores the new estimated position and velocity data as the asynchronous filter integrates from time \( \tau \) to \( t \) and outputs the last position and velocity values to initialize the open-loop filter. While the asynchronous filter for the target tracking system without road following capabilities must integrate
variables for velocity (in the $x$, $y$, and $z$ directions) and position ($x$, $y$, $z$), the asynchronous filter for the road following system only integrates the scalar velocity and road parameter, $\rho$. Since the road equation is known before tracking begins, the $\rho$ value is easily converted into $x$, $y$, and $z$ coordinates.

e. **Open-Loop Filter**

![Figure 11. Single integration open-loop filter](image)

The final step in the simulation is the open-loop filter subsystem. This block essentially performs a dead-reckoning position calculation based on the inputted estimated velocity and contains a single integrator that operates during the periods of time with no PVNT update.

3. **Road Following Filter with Numerical Euler Integration**

The main difference between the road following filter developed in Section 2 above and the one described in Section 3 of this chapter is the conversion of the Simulink model file containing the asynchronous filter in Figure 10 to a set of numerical equations. These equations are used for forward Euler integration, allowing the system to quickly determine position estimates from time $\tau$ to time $t$ (see Figure 2). It is necessary to
implement the integration model in MATLAB code to decrease computation time during simulation. Additionally, a real-time system cannot operate properly using multiple Simulink models.

\subsection*{a. Integration Equations}

In order to implement double integrator into MATLAB code, the diagram must be represented numerically. By taking the asynchronous filter model from Figure 10, removing the scope blocks, and adding state variables before and after the integrator blocks, a set of equations can be developed.

\begin{figure}[h]
\centering
\includegraphics[width=\textwidth]{asynchronous_filter_model_with_state_variables.png}
\caption{Asynchronous filter model with state variables}
\end{figure}

Using the state variables shown in Figure 12, the state equations are as follows:

\begin{align}
\dot{x}_2 &= K_2 (\hat{\rho} - \rho) \\
\dot{x}_1 &= x_2 + K_1 (\hat{\rho} - \rho) \\
\end{align}

and

\begin{align}
x_2(k+1) &= x_2(k) + \dot{x}_2 \, dt \\
x_1(k+1) &= x_1(k) + \dot{x}_1 \, dt \\
\end{align}

(2)
These state equations for Euler integration can easily be implemented in MATLAB code utilizing the data already stored in arrays and loops within the script.

B. REAL-TIME MODELING

The non real-time models can be used as a starting point for developing the general and road following models that can be implemented in real-time. Problems encountered by shifting to a real-time model and their solutions are first discussed followed by the actual design of the general and road following real-time models.

1. Problems

Many of the problems encountered during the conversion from a non real-time to real-time model dealt with synchronized data storage and retrieval. The computation speed of different methods of modeling the system is also analyzed and discussed.

a. Data Storage

One of the problems that arose when modifying the MATLAB code to allow the system to run in real-time was the method of data storage. The simulation of the system that had been created in Simulink simply wrote all of the data to arrays in the function code and by using “to workspace” blocks in the simulation model. There is no problem with this method when the system only runs for 180 seconds, as in the tests for the non real-time road following model in Chapter IV. A system that is actually implemented in hardware, however, may run from just a few minutes up to several hours. Hours of run time can result in massive amounts of data from the programs being executed. Additionally, it is not practical to increase the step size of the program to reduce the amount of data collected as accuracy will suffer as a result of the decreased number of inputs.

The solution, in the case of this system, is to only hold the minimum amount of data required before releasing it. The next decision to be made is how much data actually needs to be stored. The system being studied has a delay associated with the PVNT processing time, assumed to be anywhere from one to ten seconds. Once the
PVNT position update is calculated for time $\tau$ (equal to the current time, $t$, minus the PVNT delay), it is compared to the estimated target position at time $\tau$. This is the first portion of the asynchronous filter implemented through the Euler integration embedded in the MATLAB function. Therefore, the minimum amount of data that can be stored without affecting system function is the maximum PVNT delay divided by the step size or,

$$\frac{delay_{\text{max}}}{dT} = \frac{t - \tau}{dT}$$

(3)

From this equation, the code can be adjusted to allocate only enough data to account for the maximum expected PVNT delay. The resulting program will help avoid data overflow and storing huge amounts of data over prolonged run times.

b. The S-Function

The next problem is how to simulate the system with the necessary speed for real-time implementation. The solution to the dilemma is found in the S-function block in Simulink. The S-function block is linked to an S-Function file containing C code (in this case) that can carry out all the necessary tasks of a system with the speed needed for real-time simulation. Therefore, every MATLAB function shown in the earlier sections of this chapter such as the asynchronous filter, optimization, and data storage functions needed to be implemented in C code. After the code compiles without error, it is converted into a MATLAB .mex file, which allows it to be used by the Simulink model. The S-function can receive inputs, send out outputs, and make function calls, combining the relative simplicity of a Simulink model with the speed and capabilities of C code.

c. Arrays vs. Buffers

A problem encountered with the conversion process from MATLAB code to C code involved the fact that the size of an array in MATLAB does not have to be preset, but an array in C code does. The maximum size of an array in C code is set by using an integer to define the number of data storage spaces. Unfortunately, this means
that global variable or a parameter cannot be used to initialize the maximum array size. Therefore, the method of data storage in C code would have to be different from the methods used in MATLAB code. Since the delays of the PVNT update vary anywhere from one to ten seconds and the code had to be robust enough to handle a larger delay if the user required it, an array-based data storage system would not be practical in C code. While arrays are typically simpler to write code-wise, overflow of an array can cause errors that may prematurely end the simulation. Another problem that was encountered by using an array is that the data stored inside an array is reset following each iteration of the program. Thus, the program is not able to access information stored during previous runs, which is necessary for the asynchronous Euler integration process. As a result of these shortfalls, it was determined that another means of storing the accumulated data was needed.

While several means of data storage were tested, the only method that overcame the disadvantages of arrays and met all the requirements needed for data storage was to use buffers. While the size of a buffer does need to be preset, parameters in the S-function can be used to perform the task, even though they couldn’t be used to preset the sizes of arrays. This means that the user does not have to open the actual C program and change lines of code if the maximum expected PVNT delay were to change. The user can simply change one number in the parameter input of the S-function block and have the maximum buffer size reset automatically by the code.

Additionally, buffers are a type of persistent memory, meaning that the data stored inside remains saved until a command to clear the buffer is given. Therefore, the data from the open loop filter can be stored in buffers and recalled at the start of the asynchronous Euler integration. While the coding of the buffers is more involved than setting up a group of arrays, the requirements of the system make buffers the ideal method of data storage.

2. General Filter

Since a model or equation for the road may not be known ahead of time in most real-time situations, the general non real-time filter simulation from the beginning of this
chapter was first prepared for implementation in the S-function. The asynchronous filter portion along with all data storage from the asynchronous filter needed to be implemented in C code and moved inside the S-function.

\textbf{a. The True Target Model and PVNT Update Generator}

The simulation of the real-time system will not use an actual tracked ground target, so a model needs to be used during the testing process.

\begin{figure}[h]
\centering
\includegraphics[width=\textwidth]{figure13.jpg}
\caption{True target model and PVNT update generator for general filter}
\end{figure}

The diagram in Simulink is nearly identical to the one used in Figure 5 in the system simulation from earlier in the chapter. The target model begins with a velocity that is integrated and then sent to target position generating subsystem to determine the target position coordinates. The subsystem defines a preset road model on which the
target will travel. In the case of this system, only the $x$ and $y$ coordinates are used for the 2-D target tracking, while the $z$ coordinate is set to zero, allowing it to be included for possible future use.

Additionally, the true target model doubles as a generator for the PVNT position updates. The outputs from the position generation subsystem are utilized as a portion of the PVNT update as well. Since the assumed accuracy of the position update is $\pm 1$ meter, the model incorporates this deviation by means of a random number generator (with a mean value of zero and a range of $\pm 1$) through a summing junction. The modified PVNT position update is passed to the S-function for later use.

A small difference in the true target model from the original non real-time subsystem block shown in Figure 5 is the removal of all continuous state blocks. The integrator block is one such tool that had to be altered during the transition to a system capable of real-time calculations. Since samples are only taken every $dT$ seconds by the system during simulation and a fixed step solver is used by the Simulink model, only discrete state blocks can be used. Therefore, every continuous time integrator block used in the true target model and filter subsystems had to be swapped with discrete state integrators.

![PVNT update signal subsystem](image)

**Figure 14.** PVNT update signal subsystem
Based on previous information regarding the PVNT computation time, it is known that the delay associated with the position update can range from one to ten seconds. Therefore, a simple pulse generator could not be used due to the need for a varying delay time. The PVNT update subsystem shown above solves the problem by offering a pseudo-random sequence covering the full range of delay times.

\textit{b. The Open-Loop Filter}

During periods of operation when a PVNT update is not present, the open loop, single integration filter performs the dead reckoning calculations for target position. Unlike its road following counterpart, the general filter system does not have a $\rho$ value based on the known road model with which it can simplify calculations. Therefore, each individual coordinate has to be passed through its own open loop integrator in order to compute the updated position. So, even though all the open loop filters are connected to the same reset trigger, the $x$, $y$, and $z$ open loop filters receive their own respective position and velocity initial conditions. The entire subsystem is contained in a function call block that can be initiated by the S-function.
c. Overall Real-Time Design and Function with S-Function Block

There are more inputs and outputs to the individual subsystem blocks without the added simplicity of the $\rho$ road equation variable. The real-time general filter model really begins with the true target model subsystem and PVNT update signal blocks. The PVNT update signal is a repeating sequence that simulates a varying PVNT update delay from one to ten seconds. Until the signal goes high, indicating a PVNT position update is available; the data from the PVNT update portion of the true target model is ignored. The last known $x$, $y$, and $z$ coordinates and velocities are passed to the open loop filter where they are sent through the single integration system. The updated positions are then fed back into the S-function to be stored in the proper buffers before repeating the process.

Figure 15. General filter real-time model
However, upon the receipt of a PVNT update, the path of data slightly changes. The position update for time \( \tau \) arrives at the inputs of the S-function block and is taken into the C code. Then, the time from the last PVNT update is calculated (delay) and used to determine the value for time \( \tau \). Using the persistent memory characteristic of the buffers, the estimated position of the target at time \( \tau \) is then compared to the PVNT update and sent to a C function that performs Euler integration up to the current time \( t \), storing the new position and velocity data in buffers after each iteration. The final estimated position and velocity data from the Euler integration function for time \( t \) are passed on as the initial conditions for the open loop filter subsystem. The integrator reset is also triggered before the open loop filter calculations continue until the next PVNT position update.

During the model simulation, a storage block is used to send all the pertinent data to a .mat file. A separate script file in MATLAB loads the .mat file and automatically plots the actual target data versus the estimated target data from the filter. The data storage section of the model diagram is for testing purposes only as this process would be altered in an actual real-time simulation to avoid errors associated with storing of the immense amount of data.

3. Road Following Filter

Compared with the general filter design described above, the real-time road following filter design was greatly simplified by the pre-known road equation. This equation allowed the \( x, y, \) and \( z \) position coordinates to be combined into one parameter: \( \rho \). Not only did the road following model appear less cluttered, the C code was also somewhat simpler since only one calculation was needed in most cases where three were required before.

a. The True Target Model and PVNT Update Generator

The true target model, PVNT update generator, and PVNT update signal generator for the road following filter system are identical to the subsystem for the
general filter design shown in Figure 13. The only difference concerning the PVNT position update is in the optimization function contained in the S-function’s C code:

(1). Optimization Method. While the PVNT update that is entered into the S-function does not change from the real-time general filter model to the real-time road following filter model, there is an additional set of calculations that takes place afterwards. Located in a function declaration in the C code of the S-function, the optimization loop finds the closest point on the known road equation to the given PVNT update and sets that point as the new $\rho$ update. The optimization method for the real-time model replaces the rf_optimise.m script file used in the MATLAB function block from the non real-time road following model. In order to keep computation time to a minimum, the optimization function in the S-function C code calculates the distance from the PVNT coordinate update to set points on the road utilizing a dichotomy algorithm. This set of equations controls the adjustments made to the boundaries of search for the minimum distance, proving to be much faster than computing the distance equation for each point along the road within a set range. This direct search method’s results have a high order of accuracy while requiring a minimal amount of computation steps. Following completion of the optimization loop, the new $\rho$ value is outputted to the rest of the S-function code.

b. The Open Loop Filter

![Open loop filter diagram](image)

**Figure 16.** Open loop filter for the road following model
Figure 16 shows the simple open loop integration that calculates the update for $\rho$ and outputs the results back into the S-function for storage and further use.

**c. Overall Real-Time Design and Function with S-Function Block**

![Road following filter real-time model](image)

Figure 17. Road following filter real-time model

There are fewer inputs and outputs for the real-time model than the non real-time model. While the $x$, $y$, and $z$ coordinates are combined into the $\rho$ variable, the system function is nearly identical to the general filter. The open loop filter function call block still performs the dead reckoning integration until a PVNT position update is received and passed through the optimization function. The asynchronous forward Euler integration takes place in the C code inside the S-function but now with only the $\rho$
variable requiring integration from time $\tau$ to time $t$. As a result, only the $\rho$ variable is stored in the buffers before being sent out as the initial condition to the open loop filter as the process repeats itself.
IV. SIMULATION AND RESULTS

The purpose of this chapter is to test and compare the results from the developed real- and non real-time systems. First, different road models used in the simulations are defined. Next, the simulation parameters for each test set are defined as well as a short description of the gain values used in the asynchronous constant gain Kalman filters. The results portion of the chapter begins with simulation data from the non real-time road following filter using the two different types of integration (using an external Simulink model file vs. numerical forward Euler integration) discussed at the beginning of Chapter III. Finally, the real-time general and road following models are tested under a variety of conditions before the data is plotted and discussed.

A. SIMULATION

Two different road models were developed for simulation to determine the effects of different road characteristics on the performance of the general and road following filters. Additionally, the simulation parameters are defined as different values for certain parameters are needed for different road models.

1. Road Models

Using two road models allows a better comparison between the general and road following filters on a case-by-case basis. Each road model is created by a system of equations in the $x$ and $y$ planes, while $z$ is set to zero.

a. Third Order Road Model

The first road model is a third order system based on the set of equations below:

\[
\begin{align*}
x &= \rho \\
y &= 0.0000192\rho^3 - 0.025\rho^2 + 9.74\rho \\
z &= 0
\end{align*}
\] (4)
While the non real-time system simply uses an embedded MATLAB function (as seen in Figure 5) to simulate the road model, the real-time systems are not able to employ these embedded functions. To reduce the amount of computation time required during simulation, the road equation is created using Simulink blocks instead. The subsystem is found in the true target model for both the general and road following models. Additionally, the subsystem also calculates the derivative of each equation for use in the $T_{star}$ variable, which is used in the computation of true and estimated velocities in the road following filter model.

Figure 18. Third order road equation in Simulink subsystem for real-time simulation

When simulated for a three minute test, these equations resulted in the road model depicted in Figure 6.

b. Circular Road Model

It was decided that the second road model should be of a closed loop style similar to a rectangle or circle. Since the vehicle model uses a constant velocity during the simulation, a system of equations for a circle of constant radius was developed:
\[ r = \text{radius} \]
\[ x = r + r \sin \left( \frac{\rho}{r} + \frac{3\pi}{2} \right) \]
\[ y = r \sin \left( \frac{\rho}{r} \right) \]
\[ z = 0 \]

Identical to the third order road model, the circular road model equations had to be created in Simulink without the use of embedded MATLAB functions. The circular road equations are not used in the non real-time simulations.

![Circular road equation in Simulink subsystem for real-time simulation](image)

**Figure 19.** Circular road equation in Simulink subsystem for real-time simulation

The radius of the circle could be set to complete one loop during the simulation. A longer simulation time was chosen to display how the real-time filter does not produce errors associated with data overflow during extended tests. In the trials for this thesis, an hour long simulation was chosen, resulting in a circle radius of 2865 meters and a road model shown in the figure below:
2. Simulation Parameters

a. Simulation Time

The simulation time is set to 180 seconds for the third order road model and 3600 seconds for the circular road model. NOTE: The simulations for the non real-time road following filter only use the third order road model.

b. Sample Time

The sample time used during both the non real-time and real-time simulations for the Simulink model is 0.1 seconds. Additionally, the sample time for the general and road following S-function blocks in the real-time simulations is 0.1 seconds.

c. Asynchronous Kalman Filter Gains

The gains $k1$ and $k2$ are both set equal to 0.5. While the initial response time is slightly slower than the response time for higher gain values, trial-and-error
testing for both filters in the non real-time and real-time systems has shown that the lower gain values are more robust during periods of high PVNT noise or longer PVNT time delays.

d. **Reference Frame**

The frame reference used for all simulations is Local Tangent Plane (LTP).

e. **PVNT Parameters**

The non real-time road following model used the randomized PVNT delay time shown in Figure 4 for all simulations.

The real-time general and road following simulations vary the PVNT parameters over the series of tests. The PVNT position noise is tested at three different values: ±1, 5, and 10 meters. The PVNT delay time is also tested for three different scenarios: a simulated pseudo-random delay covering 1-10 seconds, a repeating 5 second delay, and a repeating 10 second delay. The simulated pseudo-random delay is shown in the figure below.

![Simulated pseudo-random PVNT update delay](image)

**Figure 21.** Simulated pseudo-random PVNT update delay
B. RESULTS

The results section is subdivided into the data from the non real-time simulations followed by the data from the real-time simulations. The non real-time simulations contain the road following model with the asynchronous integration performed by the external Simulink model compared to the numerical forward Euler integration method. The real-time simulations include the general and road following models. Each real-time model is also put through a series of tests in which certain PVNT parameters are altered, such as PVNT delay and input noise.

1. Non Real-Time Models

The results for the two non real-time models are divided into three comparisons each: position, velocity, and \( \rho \). Both models need to show that they are incorporating the PVNT updates into the estimated target data and effectively tracking the target throughout the simulation.
a. Road Following Model with Separate Simulink Model Integration

(1) Position Comparison

Figure 22. Comparison of actual vs. estimated target position – Simulink integration

Figure 22 depicts the results of the actual target position plotted against the estimated target position from the non real-time road following filter with external Simulink model integration. It is apparent that the position data is accurate and the model does not lose track of the target during the simulation.
Figure 23. Comparison of actual vs. estimated target velocity – Simulink integration

Figure 23 depicts the results of the actual target position plotted against the estimated target position from the non real-time road following asynchronous filter with external Simulink model integration. Since the initial velocity of the target is assumed to be 0 m/s, the estimated target velocity does not respond until the first PVNT update. The plot is zoomed in around 5 m/s (the true target velocity) to show how the estimated target velocity obtains the correct value with the help of the PVNT updates.
(3) Comparison of $\rho$ Values

The final comparison was between the actual and estimated $\rho$ values for the target. Correlating with the accuracies found on the position and velocity comparison plots, the $\rho$ comparison plot shows the same high degree of accuracy.

b. Road Following Model with Numerical Forward Euler Integration

The results for the non real-time system with the numerical integration technique are nearly identical to the method using the external Simulink model file.
(1) Position Comparison.

Figure 25. Comparison of actual vs. estimated target position – Euler integration

The figure above shows the actual target position plotted against the estimated target position for the non real-time road following filter using numerical forward Euler integration. The plot shows nearly identical results to the simulation with the double integration performed in the separate Simulink model. The estimated target position matches the actual target position with a satisfactory degree of accuracy.
(2) Velocity Comparison.

Figure 26. Comparison of actual vs. estimated target velocity – Euler integration

Figure 26 depicts the actual target velocity of 5 m/s compared with the estimated target velocity from the numerical Euler integration. While the response is not as fast as the separate Simulink model double integration, the results show that the steady state error remains at zero and the model effectively computes the estimated target velocity. If the response were deemed too slow for the environment in which the system was placed, the gain values (specifically K2) in the integration loop could be adjusted to compensate.
(3) Comparison of ρ Values.

Figure 27. Comparison of actual vs. estimated ρ velocity – Euler integration

Figure 27 shows the actual target ρ value plotted against the estimated target ρ value for the simulation using numerical forward Euler integration. Further confirming that the Euler integration contained in the MATLAB function code is accurate, the data shows nearly identical results.

Overall, the previous three figures show that the simulation can be accurately run using numerical forward Euler integration instead of the double integration process being contained in a separate Simulink model.
2. Real-Time Models

\textit{a General Filter}

(1) Ideal Conditions. Ideal conditions are defined as a PVNT noise value covering a range of ±1 meter and a simulated random PVNT delay.

\textit{(a) Third Order Road Model}

The first test for the general filter uses the third order road model under ideal conditions with a 180-second simulation time. After completion of the simulation, the results are loaded from the .mat file and comparison plots are created.

![True vs. estimated position - General filter](image)

**Figure 28. General filter position comparison plot – Third order road model – Ideal conditions**

Figure 28 shows the comparison of the target’s true position versus the general filter’s estimation for the real-time general filter model. The estimated position from the filter is quite accurate for the straighter portions of the road model and less accurate for the curved sections. A reason for the decrease in estimation
accuracy is due to the lack of an optimization function in the general filter's function. Since the general filter design does include a known road model on which to base the incoming PVNT position updates, the resulting estimated position is heavily reliant on PVNT noise. A plot of position estimation error vs. time is shown below:

![Position error vs. time - General filter](image)

**Figure 29.** General filter position error vs. time – Third order road model

Further confirming the position comparison plot in Figure 28, the error is greatest at the curved sections of the road and least during the straighter portions.
Figure 30. General filter velocity comparison plot – Third order road model – Ideal conditions

The velocity comparison plot shown above shows a small estimation error following acquisition of the target coinciding with the position plot. The overall velocity estimation accuracy is good as it stays at or near the target true velocity of 5 m/s.

Figure 31. General filter velocity error vs. time – Third order road model
The above figure shows the relationship between estimated velocity error from the real-time general filter and simulation time. After the initial target acquisition, the overall RMS error remains below 0.5 m/s.

\[(b)\] **Circular Road Model**

The circular road model simulation is run for one hour of simulation time, allowing the target model to complete one loop of the circular track.

![True vs. estimated position - General filter](image)

**Figure 32.** General filter position comparison plot – Circular road model – Ideal conditions

The position comparison plot for the circular road model appears to be much better than the third order road model. One reason for this involves the fact that the target is following a path that does not include any abrupt changes in curvature. Instead the target is engaged in one constant, gradual turn and the dead-reckoning portion of the real-time general filter is able to accurately follow the vehicle’s movement.
The real-time general filter is much more accurate for the circular road model than it is for the third order road model as shown in the above figure. The RMS position error is rarely above five meters and is centered at around one meter error due mainly to PVNT noise.
The velocity comparison plot for the circular road model is very similar to the velocity plot for the third order road model. The results from the real-time general filter show a fairly accurate estimated velocity that never strays above 5.5 m/s or below 4.75 m/s.
Figure 35. General filter velocity error vs. time – Circular road model

The estimated velocity error plot coincides with the velocity comparison plot for the circular road model. The velocity estimation performed by the real-time general filter is slightly more accurate for the circular road model than it is for the third order road model with a lower RMS error value over the system simulation time.

(2) PVNT Update Delay Variance. The next testing phase for the real-time general filter is to alter the delay time from the PVNT update signal subsystem to view the effects on target motion estimation. Instead of using the simulated pseudo-random update signal, a signal generator block is used to simulate a repeating five and ten second PVNT position update delay.

(a) Third Order Road Model

The first simulation run involved a repeating PVNT update delay of five seconds.
Figure 36. General filter position comparison plot – Third order road model – 5 second PVNT delay

Figure 36 shows the effects of a repeating five second PVNT delay on the general filter model. The result of PVNT updates arriving once every five seconds slightly decreases the estimated target position accuracy, especially around the areas of greater curvature in the road model. The error increases near the end of the simulation due to the exponential road profile equations. Typically, an update with a shorter delay time allows the model to correct itself to be closer to the actual road model in between the larger delay times of five seconds or greater. The position accuracy therefore suffers without the less delayed PVNT updates to fill in the gaps.
The velocity estimation plot shows little or no change from the random PVNT delay times. The real-time general filter remains fairly accurate with a slight bias due to the inputted PVNT noise.

Next, the real-time general filter using the third order road model is subjected to a repeating ten second PVNT update delay. Based on the PVNT background information, a ten second delay is the longest expected delay associated with the PVNT computation time.
The repeating ten second PVNT update delay greatly decreases the accuracy of the general filter model. The trend of the filter accuracy declining during areas of increased curvature turns along the road continues here as the greatest variances in estimated position accuracy are at the first turn in the third order road model.

The velocity comparison plot for the ten second PVNT update delay shows similar results when compared to the five second delay test. After the target acquisition, the filter shows good velocity estimation close to the target’s true velocity of five meters per second.

Figure 39. General filter velocity comparison plot – Third order road model – 10 second PVNT delay
(b) Circular Road Model

The varying simulation parameters that were used for the third order road model are also used for the circular road model. The repeating five second PVNT delay results are discussed first.

**Figure 40. General filter position comparison plot – Circular road model – 5 second PVNT delay**

The position comparison plot for the repeating five second PVNT delay simulation shows an estimated position that closely matches the true target position. It is necessary to view the position error vs. time plot, though, to see the true relationship due to the large sample time and axes scales.
As shown by the plot, the estimated position error from the real-time general filter remains largely unchanged with a five second PVNT delay when compared to the same trial under ideal conditions. After the initial acquisition period, the RMS error pertaining to the estimated position remains under five meters.
The velocity comparison plot shows an estimated velocity that is only slightly off of the true target’s five meter per second velocity. Based on the data from Figures 40, 41, and 42 and the circular road model simulation under ideal conditions, the real-time general filter does not lose any accuracy with the repeating five second PVNT delay.

Like the third order road model, the system using the circular road model is also tested at the upper limit of the expected PVNT delay:
The position error plot for the ten second PVNT delay simulation shows a slight increase in the estimated position error from the five second delay test. The plot in Figure 43 shows a peak error value of just less than eight meters compared to a maximum error of five meters for the five second PVNT delay error plot.
The velocity comparison plot actually shows a slightly better target velocity estimate than the repeating five second delay simulation. This is one example of how the shape of the road affects the results of the simulation. While the third order road model showed no change in velocity estimation between the five and ten second PVNT delay tests, the circular road model actually showed an improvement due to its shape.

(3) PVNT Noise Variance. The final testing phase for the real-time general filter involved setting the PVNT delay back to the simulated pseudo-random delay time and adjusting the random number generator block controlling PVNT noise in the true target model subsystem block. While the ideal conditions had a PVNT noise value of ±1 meter, the noise would be increased to ±5 and ±10 meters between the simulations.

(a) Third Order Road Model

![Figure 45. General filter position comparison plot – Third order road model – ±5 m PVNT noise](image)
First, the real-time general filter model is tested with a ±5 meter PVNT noise and the results appear quite similar to the ideal conditions test. When comparing the position plots, a slight decrease in estimation accuracy is noticed as the position updates do not match up with the target’s true position due to the extra PVNT noise.

![True vs. estimated velocity - General filter](image_url)

**Figure 46. General filter velocity comparison plot – Third order road model – ±5 m PVNT noise**

The effects of the added PVNT noise are more noticeable in the velocity comparison plot due to the larger axes in the position comparison plot. The deviation between the true and estimated velocity is greater than the velocity difference found in the ideal conditions test.

The PVNT noise is then doubled to ±10 m for the final set of tests for the third order road model using the real-time general filter design. This is very impractical as other filter designs with PVNT updates can boast ten meter accuracy, but it is important to show how much the filter can attempt to compensate to the inputted error [6].
Figure 47.  General filter position comparison plot – Third order road model – $\pm 10$ m PVNT noise

Even with a PVNT noise value having a ten meter variance in either direction, the real-time general filter shows little change from the five meter PVNT noise simulation. While the overall accuracy does have room for improvement, there is minimal change in position estimation accuracy between the five and ten meter PVNT noise tests.
The velocity comparison plot shows an estimated steady state velocity that is always within 0.6 m/s of the true target velocity. While the ± 5 meter test had a maximum error of 0.35 m/s, the test with the doubled PVNT input error shows less than a twofold increase in velocity estimation error.

(b) Circular Road Model

The circular road model is put through the same tests for PVNT noise variance as the third order road model for the real-time general filter system.
The added PVNT noise seems to have a minimal effect on the real-time general filter running the circular road model but a look at the position error plot is required due to the large axes scale.
Figure 50 shows that the added noise from the PVNT input results in an estimated position error along the circular road model that is more than double that of the ideal conditions test. The real-time general filter shows just how dependent it is on the accuracy of the PVNT position update since it does not utilize the road equation in its calculations.

![True vs. estimated velocity - General filter](image)

Figure 51. **General filter velocity comparison plot – Circular road model – ± 5 meter PVNT noise**

The velocity comparison plot shows an increase as well due to the extra PVNT input noise. There is quite a large change when compared to the velocity plot for the circular model under ideal conditions. While the ideal test resulted in a maximum estimated velocity of 5.2 m/s, the test with the PVNT noise pushed the maximum estimated velocity to over 5.5 m/s. The modular shape of the velocity estimation is due to the shape of the road profile. There are certain points in the model where there is only velocity error in the $x$ or $y$ direction as opposed to both the $x$ and $y$ directions.
Finally, the real-time general filter design using the circular road model is simulated with a ±10 meter PVNT input noise and the results are analyzed.

![Position error vs. time - General filter](image)

**Figure 52. General filter position error plot – Circular road model – ±10 meter PVNT noise**

As expected, the RMS error increased with the doubled PVNT noise as shown in Figure 52. The peak error is just over 27 meters at around 1660 seconds into the simulation.
Figure 53. General filter velocity comparison plot – Circular road model – ± 10 meter PVNT noise

The velocity comparison plot for the real-time general filter during this trial shows a maximum absolute error of around one meter per second. The figure further shows that the velocity error has a direct relationship to PVNT input error.

b. Road Following Filter

The system parameters for the tests involving the real-time road following filter are identical to those performed with the real-time general filter

(1) Ideal Conditions. Ideal conditions are defined as a PVNT noise value covering a range of ± 1 meter and a simulated random PVNT delay.
The results for the real-time road following filter using the third order road model under ideal conditions appear exponentially more accurate than the position comparison plot for the real-time general filter design under the same conditions. To confirm these results, the position error vs. time plot is examined:
The position error plot shows that after the initial target acquisition time, the real-time filter is able to estimate a target position that has less than a ten meter deviation from the actual position. During the straighter sections of the road, the position estimation is even more accurate with the error dropping to less than one meter.
The velocity comparison plot is actually quite similar to the results from the real-time general filter. The low steady state error for estimated velocity is confirmed by Figure 57:

Figure 57. Road following filter velocity error plot – Third order road model – Ideal conditions

The velocity error plot is nearly identical to the real-time general filter velocity error plot for the same conditions. Following the target acquisition period, the absolute velocity error remains less than 0.2 m/s.

Additionally, the road following filter utilizes the road parameter \( \rho \) in the equations that track target movement. This allows the variance in the estimated and actual \( \rho \) value to be plotted as well. Throughout all of the tests, the true target \( \rho \) value linearly increases with time as seen in the next figure.
The difference in the estimated and actual $\rho$ value is kept to a minimum by the real-time road following filter. The system is able to accurately estimate the $\rho$ value through the asynchronous forward Euler integration process coupled with the optimized PVNT position input.

(c) Circular Road Model

Figure 58. Road following filter $\rho$ comparison plot – Third order road model – Ideal conditions

Figure 59. Road following filter position comparison plot – Circular road model – Ideal conditions

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The position estimation for the real-time road following filter using the circular road model appears very accurate and the position error vs. time plot is examined:

![Position error vs. time - Road following filter](image)

**Figure 60.** Road following filter position error plot – Circular road model – Ideal conditions

The real-time road following filter with its added PVNT optimization function decreases the estimated error during the simulation. There is a noticeable difference when compared to the general filter simulation using the circular road model under ideal conditions. While the real-time general filter had a maximum absolute error of five meters, the real-time road following filter only had a maximum absolute error of three meters.
True vs. estimated velocity - Road following filter

The velocity comparison plot shows the estimated velocity using the real-time road following filter to be very similar to the results from the real-time general filter. It is noticed that both filters have a slight steady state velocity error during the simulations even though the precision is good.
The velocity error plot for the real-time road following filter using the circular road model is nearly identical to the third order road model. The maximum absolute estimated velocity error is never more than 0.2 m/s following the target acquisition time.
The $\rho$ comparison plot coincides with the estimated position and velocity plots, showing minimal estimation error throughout the hour long simulation.

(2) PVNT Update Delay Variance. The next testing phase for the real-time road following filter is to adjust the delay time from the PVNT update signal subsystem. Instead of using the simulated pseudo-random update signal, a signal generator block is used to simulate a repeating five and ten second PVNT position update delay.

(a) **Third Order Road Model**

![Figure 64. Road following filter position comparison plot – Third order road model – 5 second PVNT delay](image)

The expected PVNT delay can be as long as ten seconds so the five second PVNT delay does not affect the position tracking results. The velocity comparison plot below depicts similar results:
Figure 65. Road following filter velocity comparison plot – Third order road model – 5 second PVNT delay

Figure 65 shows the estimated target velocity from the real-time road following filter plotted against the actual target velocity. The results are nearly identical to the ideal conditions plot shown in Figure 56.

Figure 66. Road following filter \( \rho \) comparison plot – Third order road model – 5 second PVNT delay
The $\rho$ comparison plot depicts an estimated $\rho$ value that achieves a near zero steady state error. The repeating five second PVNT delay can be viewed during the first 30 seconds of the test as each update brings the estimated $\rho$ value closer to the target’s true $\rho$ value.

The next step involves doubling the PVNT delay to ten seconds for the third order road model.

![Figure 67. Road following filter position comparison plot – Third order road model – 10 second PVNT delay](image)

The position comparison plot for the repeating ten second PVNT delay test shows the robustness of the real-time road following filter with longer delay times. As long as the inputted PVNT update has little noise, the optimization function ensures the accuracy of the new $\rho$ value sent to the asynchronous forward Euler integration function. This results in a more accurate target tracking model even though the frequency of the updates has decreased.
Figure 68. Road following filter velocity comparison plot – Third order road model – 10 second PVNT delay

The velocity comparison plot, like the position comparison plot, shows little or no change from the additional five seconds of PVNT update delay. The maximum absolute error of the velocity estimation remains the same following target acquisition.

Figure 69. Road following filter $\rho$ comparison plot – Third order road model – 10 second PVNT delay
The ρ comparison plot in the figured above is best viewed next to the ρ comparison plot for the repeating five second PVNT delay test. Even though the delay for the position updates is twice as long, the accuracy of the ρ estimate is aided by the optimization routine in the S-function. While the settling time increases slightly, the overall steady state accuracy is not affected by the increased PVNT delay.

\( (b) \quad \text{Circular Road Model} \)

The same PVNT delay trials are performed with the real-time road following filter using the circular road model.

Initial impressions of the position comparison plot for the circular road model trial with a repeating five second PVNT update delay are good but the depiction of the estimated position error plot is shown below due to the large sample time and figure axes.

\[ \text{Figure 70. Road following filter position error plot – Circular road model – 5 second PVNT delay} \]
The plot of estimated position error vs. time shows the results of the simulation with a repeating five second PVNT update delay are no different from the simulation under ideal conditions. The maximum absolute position errors are identical between the two trials as the repeating delay shows no effect on the road following filter using the circular road model.

Figure 71. Road following filter velocity comparison plot – Circular road model – 5 second PVNT delay

The velocity comparison plot shown above for the real-time road following filter mimics the results for the position error comparison plot. No change is seen between the velocity estimation accuracy between the repeating five second delay and ideal conditions trials.

The next step involves doubling the PVNT delay to ten seconds for the circular road model.
Figure 72. Road following filter position error plot – Circular road model – 10 second PVNT delay

Figure 72 depicts the difference between the estimated and actual target position with a repeating ten second PVNT delay. Even when the PVNT delay is set at its upper expected limit, the real-time model only loses one meter of accuracy during the hour long trial. Once again, the addition of the optimization function to the real-time road following filter’s code aids the robustness of the system with respect to longer periods of time between PVNT position updates.
The velocity comparison plot for the real-time road following filter with a repeating ten second delay also shows no decrease in accuracy throughout the trial. It is interesting to note that, similar to the same trial for the real-time general filter, the steady state error of the estimated velocity actually decreases with the increase in PVNT delay time. While each real-time filter’s performance is mainly due to the code within their respective S-functions, the shape of the road model also plays a role in the accuracy of the target motion estimation.

(3) PVNT Noise Variance. The final testing phase for the real-time road following filter involves setting the PVNT delay back to the simulated pseudo-random update and adjusting the random number generator block controlling PVNT noise in the true target model subsystem block. While the ideal conditions had a PVNT noise value of ±1 meter, the noise would be increased to ±5 and ±10 meters.
Figure 74. Road following filter position comparison plot – Third order road model – ± 5 m PVNT noise

Figure 74 shows that the real-time road following filter is still able to quite accurately track the target model with the additional PVNT input noise. The five meter variance is not enough to see any noticeable differences in position estimation precision.

Figure 75. Road following filter velocity comparison plot – Third order road model – ± 5 m PVNT noise
Similar to the position comparison plot, the figure above shows a very slight variance in estimated velocity error during the curved portions of the road. The small change in velocity estimation accuracy, however, does not seriously affect the performance of target tracking.

![Figure 76. Road following filter ρ comparison plot – Circular road model – ± 5 m PVNT noise](image)

The ρ comparison plot shows that the overall accuracy of the real-time filter remains unchanged except for slight errors around one and two minutes into the simulation. When compared to the position plot, it is found that these times correspond with the major areas of greatest curvature in the road model.

The PVNT noise is then doubled to ± 10 m for the final set of tests for the third order road model using the real-time road following filter design.
Figure 77. Road following filter position comparison plot – Third order road model – ±10 m PVNT noise

The effects of the added PVNT noise still not quite noticeable in the position comparison plot even after the maximum variance of the PVNT input error is doubled. The system still appears to track the target without a significant drop in accuracy.

Figure 78. Road following filter velocity comparison plot – Third order road model – ±10 m PVNT noise

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The velocity comparison plot with the results from the trial with ±10 meters of PVNT noise finally shows the effects on the system. The optimization function located in the road following filter’s C code takes the \( x, y, z \) coordinate input from the PVNT update and finds the closest point on the pre-known road model to the PVNT input. The optimization loop ensures that the new position update lies along the road model by converting the new \( x, y, z \) coordinates into a \( \rho \) value, but it cannot guarantee the accuracy of the new estimated \( \rho \) value. Therefore, while the robustness of the real-time road following filter with respect to PVNT noise is better than the real-time general filter, the target motion estimation accuracy still decreases with larger amounts of input noise.

![True vs. estimated \( \rho \) value - Road following filter](image)

**Figure 79. Road following filter \( \rho \) comparison plot – Circular road model – ±10 m PVNT noise**

The findings from the previous figures are confirmed with the \( \rho \) comparison plot for the ±10 m PVNT noise trial. The deviations in estimated and true target \( \rho \) values are more noticeable than in the previous test. The optimization loop in the S-function code is able to greatly reduce error, but it cannot eliminate all of the variation between the actual \( \rho \) value and the resulting \( \rho \) value from the PVNT position update.
(b) Circular Road Model

The same test parameters are used with the circular road model as with the third order road model.

Figure 80. Road following filter position error plot – Circular road model – ± 5 m PVNT noise

The position error plot for the circular road model shows a large increase in RMS error from the ideal conditions test. While the RMS error for the ideal conditions trial is around two meters, the RMS error shown in Figure 81 is roughly seven meters.
The position comparison plot directly coincides with the velocity comparison plot. The precision of the velocity estimates increase at around 1250 and 3000 seconds into the simulation, resulting in better position estimation.

The PVNT noise is then doubled to $\pm 10$ m for the final set of tests for the circular road model using the real-time road following filter design.
The extra five meters of PVNT deviation greatly affect the position estimation results of the real-time road following filter for the circular road model. The peak absolute error value is only two meters less than the peak absolute error value for the same test parameters using the real-time general filter design. The RMS error, however, is much less for the real-time road following filter.
The velocity comparison plot for the real-time road following filter is nearly identical to the velocity plot for the real-time general filter shown in Figure 54. The likeness of the two plots is a perfect example of how real-time road following filter’s robustness depends not only on inputted errors, but the road model as well.

\textbf{c. Additional Road Models and Worst Case Scenarios}

The previous examples of the real-time models show that the accuracy of the target motion estimation is greatly affected by the amount of curvature present in the road model. The road following filter design is able to compensate for higher order road models and greater curvatures than the general filter design due to the fact that the road equations are used in the filter code. To show just how much of a difference there is
between the road following and general filters, the real-time simulations are run under ideal conditions using four road models with increasing amounts of curvature. The equations for the road models are shown below:

\[ x = \rho \]
\[ z = 0 \quad \text{for all road models} \]

Road model 1: \[ y = 2.7922222x \quad (6) \]

Road model 2: \[ y = 0.0000192x^3 - 0.025x^2 + 9.74x \quad (7) \]

Road model 3: \[ y = 0.000033642291x^3 - 0.0444961x^2 + 15.5884511x \quad (8) \]

Road model 4:
\[ y = -1.98707 \times 10^{-10}x^5 + 3.0300064 \times 10^{-7}x^4 - 9.832032 \times 10^{-5}x^3 - 0.0219297x^2 + 11.64832x \quad (9) \]

The four road models are plotted in the following figure to show the amount of curvature for each set of equations.

![Figure 84. Road model comparison](image-url)
The real-time road following and general filters are both run for three minute simulations and their position estimation, position error, and velocity error plots are directly compared.

(1) Road Model 1. The first road model depicts a linearly dependent first order plot where there is no curvature in the shape of the road.

Figure 85. Filter position estimation comparison – Model 1

As expected, there is no difference between the estimated positions from the real-time road following and general filters. This is a rare situation in which the dead reckoning style integration is enough to provide an accurate position estimate for both filters throughout the simulation.
The error comparison plot between the two filters confirms the results from the position estimation plot. The estimated velocity and position values from the filters are nearly identical throughout the simulation.

(2) Road Model 2. The second road model is the same system of equations used as the “third order road model” in the previous chapters. It is a third order system with a modest amount of curvature throughout the simulation run time.

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**Figure 86. Filter estimation error comparison – Model 1**

The error comparison plot between the two filters confirms the results from the position estimation plot. The estimated velocity and position values from the filters are nearly identical throughout the simulation.

(2) Road Model 2. The second road model is the same system of equations used as the “third order road model” in the previous chapters. It is a third order system with a modest amount of curvature throughout the simulation run time.

**Figure 87. Filter position estimation comparison – Model 2**
The figure above shows the amount of deviation the third order equation has between the results of the two filters. During the periods of maximum curvature, specifically around $x=250$ and $x=700$, the real-time general filter is noticed lagging in its position estimates. The real-time road following filter, on the other hand, displays an excellent target approximation compared to the actual road model throughout the test.

![Figure 88. Filter estimation error comparison – Model 2](image)

The comparison of estimated velocity error is quite similar between the two filters while the estimated position error shows a huge difference. Following the target acquisition portion of the run, the errors in estimated velocity stay below 0.25 m/s.

(3) Road Model 3. The next road model tested is a third order system that has an increased amount of curvature when compared to the second road model.
The third road model shows the real-time road following filter still performing quite well when compared to the actual target track. The dead reckoning estimation from the real-time general filter, however, is worse than the previous road model test. The sections of sharp curvature in the road create large position errors in the real-time general filter’s estimation.
Despite the large differences in the position error plot, the velocity error comparison figure shows that the real-time road following filter provides only a slightly better estimation than the real-time general filter.

(4) Road Model 4. The final road model tested is meant to greatly increase the amount of curvature seen in the previous models. This last road model features a fifth order system of equations to compare the results from the two filters.

![Filter position estimation comparison – Equation 4](image)

**Figure 91. Filter position estimation comparison – Model 4**

The extreme amount of curvature present in the fifth order road model clearly decreases the accuracy of the position estimates from the real-time general filter. The estimated target track is far outside the actual target track, especially noticeable around the final, sharp turn at \( x=850 \). The real-time road following filter, on the other hand, seems to be completely unaffected by the additional turns in the road model as the robustness of the filter to road model curvature is displayed.
Figure 92. Filter estimation error comparison – Model 4

Similar to the previous trials, the velocity estimation errors are nearly equal between the two filters while the position error plots display large variances. Despite the similarities on the velocity plot, though, the real-time road following filter proves that it is a much better predictor of target motion than the real-time general filter for road models with varying amounts of curvature.

(5) Worst case scenario. A worst case scenario is chosen to show that even though the real-time road following model shows overall great target motion estimation, there are still limits to the amount of PVNT input noise and delay that it can overcome. To show a true failure of the filter, the PVNT input delay is set to 50 seconds for the real-time road following model using the circular road model.
As shown in Figure 93, the estimated position of the target model is very poor. The 50 second PVNT delay is simply too long for the filter to accurately predict target motion.

Figure 94. Filter estimation error comparison – Worst case scenario
The velocity error plot shows a steady state error of around 130 m/s while the position error plot depicts nearly a five kilometer position estimation error. Remembering that the circular path is only 2865 meters in radius, these results illustrate a complete failure of the real-time road following model.
V. CONCLUSION AND RECOMMENDATIONS

A. CONCLUSIONS

The overall goals set for this thesis were accomplished. The non real-time road following model was successfully developed and tested to ensure proper function. The problems associated with converting the non real-time systems to real-time were solved with the use of buffers and the implementation of S-function C code. Finally, the real-time general and road following systems were successfully modeled and simulated. All of the results for the real-time models were then compiled and analyzed to provide a definite set of conclusions.

Throughout the simulations in the thesis, the road following filter design shows that it is a better target motion estimator than the general filter design. The simulations display the relative robustness of the real-time road following model to several forms of PVNT input errors while the real-time general filter model results were less accurate. Additionally, the ability of the real-time road following model to provide accurate position and velocity estimation results along simulated roads of increasing curvature were shown. The real-time general filter faltered on road models containing larger amounts of curvature as the dead reckoning integration without the optimization technique was not enough to give accurate results. Finally, while the real-time road following filter performed well in all the practical simulations put forth in the thesis, it was shown that the filter can fail in a worst case scenario involving exceptionally large PVNT input errors.

B. RECOMMENDATIONS

There are quite a few opportunities for further work on the subject of this thesis. The methods used are a solid foundation on which improvements can be made. One simple test that can be worked on includes adjusting the $k_1$ and $k_2$ gain values for the asynchronous integration loop. Similar to a proportional compensator, lower gain values result in smaller overshoot with slower response time while higher gain values improve
response time but increase overshoot. Some tweaking may be required to find the best compromise for filter performance that will more accurately represent a field testing environment.

Another possible improvement concerning filter accuracy can be made by examining the asynchronous filter integration process. Currently, the model utilizes forward Euler integration to cycle back from the delayed PVNT update to the current simulation time. Future work may involve using trapezoidal or higher order of integration to see if this improves overall target motion estimation accuracy.

Other areas for immediate work include improvements to the S-function code to ensure the minimal amount of required computation time along with storing and plotting the data from each iteration of the asynchronous filter. Further testing can provide results with different types of road models to see what direct relationships exist between road models and filter performance.

Eventually, the simulated real time system can be loaded into hardware and bench-tested. The final goal is to have a program that is able to run in real time on an unmanned aerial vehicle during field testing.
APPENDIX

This appendix presents the ANSI C code for the general and road following filter S-functions as well as a manual explaining the programs’ operation.

The purpose of the S-function is to provide an alternative method to MATLAB functions that will allow the system to perform real-time simulations. Each filter design performs a number of different operations with the goal of providing accurate target motion estimation. The filters receive delayed PVNT updates, perform asynchronous forward Euler integration from the update time to current time, and then output the results to the open loop filter. The open loop filter then runs until the next PVNT update arrives.

A. GENERAL FILTER

1. Manual

File: s_filter_general.c

Lines 26-49
Complete basic program initializations, library calls, and global variable input.

Lines 26 and 27
Designate the file name and indicate that the file is in C code, to be converted into MEX format and run in MATLAB.

Lines 29-34
Make all the necessary library calls that are required in the program.

Lines 40 and 41
Take in the two S-function parameters, MAX_DELAY and TIME_STEP, from the S-function block in the Simulink model. MAX_DELAY is the maximum amount of expected delay in between PVNT updates while TIME_STEP is the time step to be used by the C code. NOTE: The time step parameter value must match the discrete time step value found on the simulation parameters menu in Simulink.
**Lines 45 and 46**
Convert the parameters into “real_T” format for use in numerical calculations later.

**Line 49**
Defines the global variable `MAX_INDEX`, used to ensure that buffer overflow does not occur.

_Euler_integration_ function

**Line 58**
Lists the inputs to the function along with buffers marked by an asterisk in front of their names.

**Lines 65-72**
Perform forward Euler integration for the x, y, and z variables, assigning the new position and velocity values to the buffers beginning with “temp.”

_mdllInitializeSizes_
Sets up the sizes of the various vectors used in the code.

**Line 82**
Means that there will be two parameters inputted into the S-function block in Simulink.

**Lines 83-86**
Return an error to MATLAB if the incorrect number of parameters is found.

**Line 88**
Defines zero continuous states since the model is running with a preset, fixed step time.

**Line 89**
Defines eleven discrete states which must match the number of input ports found in line 91.

**Lines 92-102**
Set the size of each input port
Denote each input port as a direct feed through port.

Defines eight output ports from the S-function.

Define the width of each port.

Defines one sample time to be used.

Define the number of real, integer, pointer, and mode work vectors to be used in the program. The work vectors can be thought of as a value of a certain type (real, integer, pointer, etc.) that is stored in persistent memory. This means that the value will be stored even while the program is called multiple times.

Defines the number of zero crossings to be zero as it is not used in the filter program.

Defines the program’s sample time to be set to $dT$, which come from the second parameter input to the S-function block in line 45.

Indicates a 0.0 second offset time

Indicates that a function call is made on the first element of the first output port.

Defines all of the variables that need to be initialized only once, i.e. the very first time the program is run in the simulation.
Line 161

Predefines the integer work vectors for the index counter and the integrator flag that indicates when the discrete integrator blocks in the open-loop filter subsystem need to be reset.

Lines 162 and 163

Predefine the real work vectors for the initial x, y, and z positions and velocities.

Lines 168-170

Predefine the buffers that are used in the code for data storage.

Lines 175-189

Initialize the buffers to a number of positions equal to MAX_INDEX (from line 49) with each position having enough memory to store a piece of data with the size real_T. The calloc command also initializes every position in the buffers to zero.

Lines 191-199

Define the first value for the index counter, integrator reset flag, and position and velocity initial conditions to be zero.

Lines 203-219

Set the pointer work vectors to point to the first position of each of the buffers.

Lines 221-230

Set and store the initial integer and real work values.

mdlOutputs function

Lines 242-270

Contain the input and output declarations.

Lines 242-247 and 253-258

Define the pointers and values of the position and velocity estimates coming from the open loop filter function call.

Lines 248 and 259

Define the pointer and value coming in from the PVNT update delay subsystem

Lines 249-251 and 260-263
Define the actual PVNT update \((x,y,z)\) from the target model subsystem.

**Lines 252 and 263**

Designate a port for the clock input.

**Lines 264-269**

Define output ports for the position and velocity initial conditions to the open-loop filter function call.

**Line 270**

Defines the integrator reset signal, which is also fed into the open-loop filter function call.

**Lines 272-280**

Contain declarations for the work values and the buffers which match the declarations found in the `mdlStart` function.

**Lines 286-290**

Define and initialize the non-persistent variables that are used only in the `Euler_integration` and `mdlOutputs` function.

**Lines 302-318**

Retrieves the values that were stored in the pointer work vectors.

**Lines 321-328**

Retrieves the values that were stored in the integer and real work vectors.

**Lines 338-388**

Contained in an `if` loop that executes only if the index counter is less than or equal to the preset `MAX_INDEX` value. This ensures that no data is written to the buffers beyond their maximum preset number of storage positions, reducing the risk of buffer overflow.

**Line 345**

Sets the `integrator_reset` output to the integer work value `integrator_flag`.

**Lines 354-359**

Take in the estimated position and velocity values from the first six inputs (arriving from the outputs of the open-loop filter function call).

**Lines 362-367**
Set the respective buffer values to the inputted position and velocity estimates. These values are then also stored in the real work vectors designating position and velocity initial conditions.

**Line 378**

Resets the integrator flag integer work value to zero (if it was set to one following the Euler integration loop, see line 473).

**Lines 380-382**

Take in the PVNT position update \((x,y,z)\) from the true target model subsystem in the Simulink diagram

**Lines 385-387.**

Assign the values from the PVNT position update to their respective buffers.

**Lines 392-475**

Contained in an if loop that is only triggered if the input from the PVNT delay subsystem is set high, indicating that a PVNT update is available.

**Lines 395-400**

Adjust the pointers to each position and velocity buffer so that they now refer to time \(\tau\), the time to which the PVNT update refers. This is controlled by the index integer work vector which is incremented after each iteration of the `mdlOutputs` function (see line 479).

**Lines 402-404**

Perform the same operation for the buffers that contain the PVNT position update data.

**Lines 408-410**

Calculate the difference between the estimated position data at time \(\tau\) and the PVNT position update at time \(\tau\) for \(x, y,\) and \(z\).

**Lines 414-419**

Set up the values for the first position of the buffers that are used in the `Euler_integration` function and to pass on the updated position and velocity data to the open-loop filter function call.
Line 425

Begins the asynchronous portion of the S-function. The *for* loop runs enough times to move the new estimated position and velocity values from time $\tau$ to time $t$ (current system time), which is controlled by the *index* integer work vector value.

Lines 430-432

Set the *delta* variable values originally set in lines 408-410 to zero after the first iteration of the *for* loop, allowing for normal, dead-reckoning style integration.

Line 435

Passes the required variables to the *Euler_integration* function in lines 58-73. Additionally, the “&” in front of the *temp* buffers indicate that their changed values from the *Euler_integration* function will be saved after the function executes.

Lines 439-444

Increment the pointer values for the buffers that will contain the updated position and velocity estimates.

Lines 448-453

Actually set the buffers equal to the updates.

Lines 463-468

After the *for* loop runs the appropriate number of times to arrive at time $t$, the final value from each of the buffers containing the updated position and velocity estimates are passed to the initial condition real work vectors in these lines.

Additionally, two integrator reset values are set. The first is the *reset_index* variable on line 470 set equal to one and used inside the S-function program on line 477. The second is the *integrator_flag* integer work value on line 473 that is outputted to the open-loop function call outside the S-function block.

The remainder of the buffer pointer incrementation/resets take place in the *if/else* loop in lines 477-501. The *if* loop portion checks to see if the current *index*
variable value is less that the preset $MAX\_INDEX$ value and if the $reset\_index$ variable value is equal to zero (indicating that a PVNT update did not arrive during the current $mdl\_outputs$ function iteration. If so, the $index$ integer work value is incremented along with the pointers to the position and velocity data buffers.

If the criterion for the $if$ loop are not met, meaning that a PVNT update has occurred, the buffer pointers are all reset back to their first position and the $index$ integer work value is set to zero. This ensures that the buffers are simply overwritten with the new data until the next PVNT update and buffer overflow does not occur. Finally, the pointer work values are updated to now designate the new pointer values for the position and velocity data buffers.

$mdl\_Update$ function

This would be the function in which states would be incremented if they were used in the program. Since the filter design does not use these states, however, the $mdl\_Update$ function is only left in the program as a formality.

$mdl\_Terminate$ function

In this case, all of the data from the buffers must be cleared to avoid errors when re-running the simulation multiple times.

**Lines 537-553**

Designate each of the buffers that were originally defined in the $mdl\_Start$ function.

**Lines 560-574**

Actually release the data stored in the buffers.
# 2. Code

```c
/* File    : s_filter_general.c
 * Abstract:
 *
 * This S-function is a combination of an open-loop filter using a
 * function call subsystem and an asynchronous filter contained in the
 * C code of the S-function. The model is used for a target tracking
 * system, utilizing a delayed position update at different time
 * intervals. When the position update (labeled PVNT) is not
 * available, the S-function calls the open-loop filter and stores the
 * results. When the delayed position update arrives, the loop
 * containing the asynchronous filter is run to update the previous
 * data from time tau (corresponding to the PVNT update) to time t
 * (corresponding to the current time) using buffers to store all
 * data. The model takes in parameters from the S-function block in
 * the Simulink model for the maximum amount of delay (seconds) and
 * the desired time step (seconds). The user can easily manipulate
 * these parameters without having to change C code in the S-function
 *
 * For more details about S-functions, see
 * matlabcot/simulink/src/sfunimpl_doc.c
 *
 * Copyright 1990-2006 The MathWorks, Inc.
 * $Revision: 1.15.4.3 $
 */

#define S_FUNCTION_NAME  s_filter_general
#define S_FUNCTION_LEVEL 2

#include "simstruc.h"

#include <stdlib.h>
#include <stdio.h>
#include <string.h>
#include <math.h>

/* Input Arguments */
/*takes in parameters that define a max value for the PVNT update delay and
 *the desired time step*/
#define MAX_DELAY               ssGetSFcnParam(S,0)
#define TIME_STEP               ssGetSFcnParam(S,1)

/*converts the above parameters from structs to allow them to be used in
 *computations*/
#define dT                    ((real_T) mxGetPr(TIME_STEP)[0])
#define DELAY_MAX           ((real_T) mxGetPr(MAX_DELAY)[0])

/*defines and global constant that is used to prevent buffer overflow*/
#define MAX_INDEX           (DELAY_MAX/dT)
```
/* Function: Euler_integration ============================================
 * Abstract:
 * Performs asynchronous forward Euler integration once the PVNT update is
 * received in order to rewrite over the previous data from time tau to
 * time t.
 */
void Euler_integration(double k1, double k2, float delta_x_tou, float delta_y_tou, float
delta_z_tou, float time_step, real_T *new_Px_est_tou, real_T *new_Vx_est_tou, real_T
*new_Py_est_tou, real_T *new_Vy_est_tou, real_T *new_Pz_est_tou, real_T *new_Vz_est_tou,
real_T *temp_new_Px_est_tou, real_T *temp_new_Vx_est_tou,
real_T *temp_new_Py_est_tou, real_T *temp_new_Vy_est_tou, real_T
*temp_new_Pz_est_tou, real_T *temp_new_Vz_est_tou)
{
    /*performs asynchronous double integration with a time step
     *equal to dT seconds and stores the results in a temp variable
     *to be transferred to the buffers after they have been
     *incremented*/

    *temp_new_Px_est_tou = *new_Px_est_tou+ (*new_Vx_est_tou +
k1*delta_x_tou)*time_step;
    *temp_new_Vx_est_tou = *new_Vx_est_tou+ (k2*delta_x_tou)*time_step;

    *temp_new_Py_est_tou = *new_Py_est_tou+ (*new_Vy_est_tou +
k1*delta_y_tou)*time_step;
    *temp_new_Vy_est_tou = *new_Vy_est_tou+ (k2*delta_y_tou)*time_step;

    *temp_new_Pz_est_tou = *new_Pz_est_tou+ (*new_Vz_est_tou +
k1*delta_z_tou)*time_step;
    *temp_new_Vz_est_tou = *new_Vz_est_tou+ (k2*delta_z_tou)*time_step;
}

/* Function: mdlInitializeSizes ===========================================
 * Abstract:
 * Setup sizes of the various vectors.
 */
static void mdlInitializeSizes(SimStruct *S)
{
    ssSetNumSFcnParams(S, 2); /* Number of expected parameters */
    if (ssGetNumSFcnParams(S) != ssGetSFcnParamsCount(S))
    {
        return; /* Parameter mismatch will be reported by Simulink */
    }
    ssSetNumContStates(S, 0); /*defines 0 continuous states*/
    ssSetNumDiscStates(S, 11); /*defines 11 discrete states*/

    if (!ssSetNumInputPorts(S, 11)) return; /*defines 11 input ports*/
    ssSetInputPortWidth(S, 0, 1); /*sets input 1 port size to 1*/
    ssSetInputPortWidth(S, 1, 1); /*sets input 2 port size to 1*/
    ssSetInputPortWidth(S, 2, 1); /*sets input 3 port size to 1*/
    ssSetInputPortWidth(S, 3, 1); /*sets input 4 port size to 1*/
    ssSetInputPortWidth(S, 4, 1); /*sets input 5 port size to 1*/
    ssSetInputPortWidth(S, 5, 1); /*sets input 6 port size to 1*/
    ssSetInputPortWidth(S, 6, 1); /*sets input 7 port size to 1*/
}
ssSetInputPortWidth(S, 7, 1); /*sets input 8 port size to 1*/
ssSetInputPortWidth(S, 8, 1); /*sets input 9 port size to 1*/
ssSetInputPortWidth(S, 9, 1); /*sets input 10 port size to 1*/
ssSetInputPortWidth(S, 10, 1); /*sets input 11 port size to 1*/
ssSetInputPortDirectFeedThrough(S, 0, 1);
ssSetInputPortDirectFeedThrough(S, 1, 1);
ssSetInputPortDirectFeedThrough(S, 2, 1);
ssSetInputPortDirectFeedThrough(S, 3, 1);
ssSetInputPortDirectFeedThrough(S, 4, 1);
ssSetInputPortDirectFeedThrough(S, 5, 1);
ssSetInputPortDirectFeedThrough(S, 6, 1);
ssSetInputPortDirectFeedThrough(S, 7, 1);
ssSetInputPortDirectFeedThrough(S, 8, 1);
ssSetInputPortDirectFeedThrough(S, 9, 1);
ssSetInputPortDirectFeedThrough(S, 10, 1);

if (!ssSetNumOutputPorts(S, 8)) return;
ssSetOutputPortWidth(S, 0, 1); /*sets output port 1 width to 1*/
ssSetOutputPortWidth(S, 1, 1); /*sets output port 2 width to 1*/
ssSetOutputPortWidth(S, 2, 1); /*sets output port 3 width to 1*/
ssSetOutputPortWidth(S, 3, 1); /*sets output port 4 width to 1*/
ssSetOutputPortWidth(S, 4, 1); /*sets output port 5 width to 1*/
ssSetOutputPortWidth(S, 5, 1); /*sets output port 6 width to 1*/
ssSetOutputPortWidth(S, 6, 1); /*sets output port 7 width to 1*/
ssSetOutputPortWidth(S, 7, 1); /*sets output port 8 width to 1*/
ssSetNumSampleTimes(S, 1);
ssSetNumRWork(S, 6); /*real vector*/
ssSetNumIWork(S, 2); /*integer vector*/
ssSetNumPWork(S, 15); /*pointer vector*/
ssSetNumModes(S, 0); /*mode vector*/
ssSetNumNonsampledZCs(S, 0); /*number of zero crossings*/

/* Take care when specifying exception free code - see sfuntmpl_doc.c */
ssSetOptions(S, SS_OPTION_EXCEPTION_FREE_CODE);
}

/* Function: mdlInitializeSampleTimes ===================================
 * Abstract:
 *    Discrete sample time of dT seconds and specify that we are doing
 *    function-call's on the 1st element of the 1st output port.
 */
static void mdlInitializeSampleTimes(SimStruct *S)
{
  ssSetSampleTime(S, 0, dT); /*sets sample time to dT seconds*/
  ssSetOffsetTime(S, 0, 0.0);  /*indicates 0 offset time*/
  ssSetCallSystemOutput(S, 0);    /* call on first element */
  ssSetModelReferenceSampleTimeDefaultInheritance(S);
}

/*Function: mdlStart ====================================================
 *Abstract:
This function sets up the variables passed between the function and the s-function.

```c
#define MDL_START

static void mdlStart(SimStruct *S) {
    int_T index, integrator_flag;
    real_T initial_x_position, initial_y_position, initial_z_position;
    real_T initial_x_velocity, initial_y_velocity, initial_z_velocity;

    /*The four real_T variables below denote the buffers used to store the real_T variables*/
    real_T *velocity_x_data, *position_x_data, *velocity_y_data, *position_y_data,
    *velocity_z_data, *position_z_data;
    real_T *new_Vx_est_tou, *new_Px_est_tou, *new_Vy_est_tou, *new_Py_est_tou,
    *new_Vz_est_tou, *new_Pz_est_tou;
    real_T *x_PVNT_data, *y_PVNT_data, *z_PVNT_data;

    /*The buffers are allocated enough memory to store 'MAX_INDEX' data with each data space being 'real_T' size. The 'calloc' command also initializes the buffers*/
    velocity_x_data = (real_T*) calloc(MAX_INDEX, sizeof(real_T));
    position_x_data = (real_T*) calloc(MAX_INDEX, sizeof(real_T));
    velocity_y_data = (real_T*) calloc(MAX_INDEX, sizeof(real_T));
    position_y_data = (real_T*) calloc(MAX_INDEX, sizeof(real_T));
    velocity_z_data = (real_T*) calloc(MAX_INDEX, sizeof(real_T));
    position_z_data = (real_T*) calloc(MAX_INDEX, sizeof(real_T));
    new_Vx_est_tou = (real_T*) calloc(MAX_INDEX, sizeof(real_T));
    new_Px_est_tou = (real_T*) calloc(MAX_INDEX, sizeof(real_T));
    new_Vy_est_tou = (real_T*) calloc(MAX_INDEX, sizeof(real_T));
    new_Py_est_tou = (real_T*) calloc(MAX_INDEX, sizeof(real_T));
    new_Vz_est_tou = (real_T*) calloc(MAX_INDEX, sizeof(real_T));
    new_Pz_est_tou = (real_T*) calloc(MAX_INDEX, sizeof(real_T));
    x_PVNT_data = (real_T*) calloc(MAX_INDEX, sizeof(real_T));
    y_PVNT_data = (real_T*) calloc(MAX_INDEX, sizeof(real_T));
    z_PVNT_data = (real_T*) calloc(MAX_INDEX, sizeof(real_T));

    index = 0;  //initializes index to 0*
    integrator_flag = 0;  //sets integration reset flag to 0

    initial_x_velocity = 0.0;  //initializes position and velocity
    initial_y_velocity = 0.0;
    initial_z_velocity = 0.0;
    initial_x_position = 0.0;
    initial_y_position = 0.0;
    initial_z_position = 0.0;

    /*Sets the pointer work variables for the buffers*/
    ssSetPWorkValue(S, 0, (real_T*)velocity_x_data);
    ssSetPWorkValue(S, 1, (real_T*)position_x_data);
    ssSetPWorkValue(S, 2, (real_T*)velocity_y_data);
    ssSetPWorkValue(S, 3, (real_T*)position_y_data);
}````
ssSetPWorkValue(S, 4, (real_T *)velocity_z_data);
ssSetPWorkValue(S, 5, (real_T *)position_z_data);
ssSetPWorkValue(S, 6, (real_T *)new_Vx_est_tou);
ssSetPWorkValue(S, 7, (real_T *)new_Px_est_tou);
ssSetPWorkValue(S, 8, (real_T *)new_Vy_est_tou);
ssSetPWorkValue(S, 9, (real_T *)new_Py_est_tou);
ssSetPWorkValue(S, 10, (real_T *)new_Vz_est_tou);
ssSetPWorkValue(S, 11, (real_T *)new_Pz_est_tou);
ssSetPWorkValue(S, 12, (real_T *)x_PVNT_data);
ssSetPWorkValue(S, 13, (real_T *)y_PVNT_data);
ssSetPWorkValue(S, 14, (real_T *)z_PVNT_data);
ssSetIWorkValue(S, 0, index);          /*sets the first integer work
value to the index variable*/
ssSetIWorkValue(S, 1, integrator_flag);  /*sets the second integer work
value to the integrator flag*/
ssSetRWorkValue(S, 0, initial_x_velocity);  /*sets the real work values*/
ssSetRWorkValue(S, 1, initial_x_position);
ssSetRWorkValue(S, 2, initial_y_velocity);
ssSetRWorkValue(S, 3, initial_y_position);
ssSetRWorkValue(S, 4, initial_z_velocity);
ssSetRWorkValue(S, 5, initial_z_position);
}

*/ Function: mdlOutputs ==
Abstract:
Issue ssCallSystemWithTid on 1st output element of 1st output port
and then update 2nd output port with the state.
*/
static void mdlOutputs(SimStruct *S, int_T tid)
{
    /*S-function input and output declarations*/
    real_T *Vx_est          = ssGetRealDiscStates(S,0);
    real_T *Px_est          = ssGetRealDiscStates(S,1);
    real_T *Vy_est          = ssGetRealDiscStates(S,2);
    real_T *Vz_est          = ssGetRealDiscStates(S,4);
    real_T *Pz_est          = ssGetRealDiscStates(S,5);
    real_T *PVNT           = ssGetRealDiscStates(S,6);
    real_T *x_ro            = ssGetRealDiscStates(S,7);
    real_T *y_ro            = ssGetRealDiscStates(S,8);
    real_T *z_ro            = ssGetRealDiscStates(S,9);
    real_T *clock          = ssGetRealDiscStates(S,10);
    InputRealPtrsType Vx_est_Ptrs      = ssGetInputPortRealSignalPtrs(S,0);
    InputRealPtrsType Px_est_Ptrs      = ssGetInputPortRealSignalPtrs(S,1);
    InputRealPtrsType Vy_est_Ptrs      = ssGetInputPortRealSignalPtrs(S,2);
    InputRealPtrsType Py_est_Ptrs      = ssGetInputPortRealSignalPtrs(S,3);
    InputRealPtrsType Vz_est_Ptrs      = ssGetInputPortRealSignalPtrs(S,4);
    InputRealPtrsType Pz_est_Ptrs      = ssGetInputPortRealSignalPtrs(S,5);
    InputRealPtrsType PVNT_Ptrs       = ssGetInputPortRealSignalPtrs(S,6);
    InputRealPtrsType x_ro_Ptrs        = ssGetInputPortRealSignalPtrs(S,7);
    InputRealPtrsType y_ro_Ptrs        = ssGetInputPortRealSignalPtrs(S,8);
    InputRealPtrsType z_ro_Ptrs        = ssGetInputPortRealSignalPtrs(S,9);
InputRealPtrsType clock_Ptrs = ssGetInputPortRealSignalPtrs(S, 10);
real_T *TgtVx_IC = ssGetOutputPortRealSignal(S, 1);
real_T *TgtPx_IC = ssGetOutputPortRealSignal(S, 2);
real_T *TgtVy_IC = ssGetOutputPortRealSignal(S, 3);
real_T *TgtPy_IC = ssGetOutputPortRealSignal(S, 4);
real_T *TgtVz_IC = ssGetOutputPortRealSignal(S, 5);
real_T *TgtPz_IC = ssGetOutputPortRealSignal(S, 6);
real_T *integrator_reset = ssGetOutputPortRealSignal(S, 7);

int_T index, integrator_flag;
real_T initial_x_position, initial_y_position, initial_z_position;
real_T initial_x_velocity, initial_y_velocity, initial_z_velocity;
real_T temp_new_Vx_est_tou, temp_new_Px_est_tou, temp_new_Vy_est_tou,
temp_new_Py_est_tou, temp_new_Vz_est_tou, temp_new_Pz_est_tou;

/*buffer declarations for mdlOutputs*/
real_T *velocity_x_data, *position_x_data, *velocity_y_data,
*position_y_data, *velocity_z_data, *position_z_data;
real_T *new_Vx_est_tou, *new_Px_est_tou, *new_Vy_est_tou, *new_Py_est_tou,
*new_Vz_est_tou, *new_Pz_est_tou;
real_T *x_PVNT_data, *y_PVNT_data, *z_PVNT_data;

/*defines pointer to output file for forward Euler integration results*/
FILE *Euler_output_data;

/*defines intermediate position and velocity matrices*/
float delta_x_tou = 0.0, delta_y_tou = 0.0, delta_z_tou = 0.0;
float time_index = 0.0, delay = 0.0, time_step = dT;
int i = 0;        /*counter*/
int reset_index = 0;    /*flag indicating and index reset to 0*/
double k1=0.5, k2=0.5;   /*sets integrator gains*/

/* ssCallSystemWithTid is used to execute a function-call subsystem. The
 * 2nd argument is the element of the 1st output port index which
 * connected to the function-call subsystem. Function-call subsystems
 * can be driven by the first output port of s-function blocks.
 */

UNUSED_ARG(tid);    /* not used in single tasking mode */

/*Retrieves the pointer work values for the buffers*/
velocity_x_data = (real_T *)ssGetPWorkValue(S, 0);
position_x_data = (real_T *)ssGetPWorkValue(S, 1);
velocity_y_data = (real_T *)ssGetPWorkValue(S, 2);
position_y_data = (real_T *)ssGetPWorkValue(S, 3);
velocity_z_data = (real_T *)ssGetPWorkValue(S, 4);
position_z_data = (real_T *)ssGetPWorkValue(S, 5);
new_Vx_est_tou = (real_T *)ssGetPWorkValue(S, 6);
new_Px_est_tou = (real_T *)ssGetPWorkValue(S, 7);
new_Vy_est_tou = (real_T *)ssGetPWorkValue(S, 8);
new_Py_est_tou = (real_T *)ssGetPWorkValue(S, 9);
new_Vz_est_tou = (real_T *)ssGetPWorkValue(S, 10);
new_Pz_est_tou = (real_T *)ssGetPWorkValue(S, 11);
x_PVNT_data = (real_T *)ssGetPWorkValue(S, 12);
y_PVNT_data = (real_T *)ssGetPWorkValue(S, 13);
z_PVNT_data = (real_T *)ssGetPWorkValue(S, 14);

/*Retrieves integer and real work values*/
index = ssGetIWorkValue(S,0);
integrator_flag = ssGetIWorkValue(S,1);
initial_x_velocity = ssGetRWorkValue(S,0);
initial_x_position = ssGetRWorkValue(S,1);
initial_y_velocity = ssGetRWorkValue(S,2);
initial_y_position = ssGetRWorkValue(S,3);
initial_z_velocity = ssGetRWorkValue(S,4);
initial_z_position = ssGetRWorkValue(S,5);

/*creates .txt file for output results*/
Euler_output_data = fopen("Euler_data_general.txt", "w");

/*Entire sequence is in an 'if' loop to ensure that there is no
overflow for the position and velocity arrays (defined with a maximum
of MAX_INDEX data points.)*/
if(index <= (int)MAX_INDEX)
{
    TgtPx_IC[0] = initial_x_position; /*sets outputs to initial V and P*/
    TgtVx_IC[0] = initial_x_velocity;
    TgtPy_IC[0] = initial_y_position;
    TgtVy_IC[0] = initial_y_velocity;
    TgtPz_IC[0] = initial_z_position;
    TgtVz_IC[0] = initial_z_velocity;

    integrator_reset[0] = integrator_flag; /*sets output 3 to integration
                                          *reset flag*/

    if(!ssCallSystemWithTid(S,0,tid)) /*calls system with task ID 1*/
    {
        /* Error occurred which will be reported by Simulink */
        return;
    }

    Vx_est_Ptrs = ssGetInputPortRealSignalPtrs(S,0); /*Gets inputs*/
    Px_est_Ptrs = ssGetInputPortRealSignalPtrs(S,1);
    Vy_est_Ptrs = ssGetInputPortRealSignalPtrs(S,2);
    Py_est_Ptrs = ssGetInputPortRealSignalPtrs(S,3);
    Vz_est_Ptrs = ssGetInputPortRealSignalPtrs(S,4);
    Pz_est_Ptrs = ssGetInputPortRealSignalPtrs(S,5);

    /*assigns the position and velocity data to the buffers*/
    *position_x_data = (real_T)*Px_est_Ptrs[0];
    *velocity_x_data = (real_T)*Vx_est_Ptrs[0];
    *position_y_data = (real_T)*Py_est_Ptrs[0];
    *velocity_y_data = (real_T)*Vy_est_Ptrs[0];
    *position_z_data = (real_T)*Pz_est_Ptrs[0];
    *velocity_z_data = (real_T)*Vz_est_Ptrs[0];

    /*resets the initial velocity and position values*/
}
initial_x_velocity = ssSetRWorkValue(S, 0, (real_T)*Vx_est_Ptrs[0]);
initial_x_position = ssSetRWorkValue(S, 1, (real_T)*Px_est_Ptrs[0]);
initial_y_velocity = ssSetRWorkValue(S, 2, (real_T)*Vy_est_Ptrs[0]);
initial_y_position = ssSetRWorkValue(S, 3, (real_T)*Py_est_Ptrs[0]);
initial_z_velocity = ssSetRWorkValue(S, 4, (real_T)*Vz_est_Ptrs[0]);
initial_z_position = ssSetRWorkValue(S, 5, (real_T)*Pz_est_Ptrs[0]);

/*resets the integrator reset to 0*/
integrator_flag = ssSetIWorkValue(S, 1, 0);

x_ro_Ptrs = ssGetInputPortRealSignalPtrs(S,7); /*takes in ro_star value*/
y_ro_Ptrs = ssGetInputPortRealSignalPtrs(S,8);
z_ro_Ptrs = ssGetInputPortRealSignalPtrs(S,9);

/*assigns coordinates to buffers*/
*x_PVNT_data = (real_T)*x_ro_Ptrs[0];
*y_PVNT_data = (real_T)*y_ro_Ptrs[0];
*z_PVNT_data = (real_T)*z_ro_Ptrs[0];

if ((real_T)*PVNT_Ptrs[0] >= 0.99) /*indicates pulse is high (PVNT update present)*/
{
    /*calls the estimated position and velocity values at time tou from
    *the buffers*/
    position_x_data = position_x_data - index;
    velocity_x_data = velocity_x_data - index;
    position_y_data = position_y_data - index;
    velocity_y_data = velocity_y_data - index;
    position_z_data = position_z_data - index;
    velocity_z_data = velocity_z_data - index;

    x_PVNT_data = x_PVNT_data - index;
    y_PVNT_data = y_PVNT_data - index;
    z_PVNT_data = z_PVNT_data - index;

    /*calculates the difference between the ro_star update value and the
    *estimated ro value at time tou*/
    delta_x_tou = *x_PVNT_data - *position_x_data;
    delta_y_tou = *y_PVNT_data - *position_y_data;
    delta_z_tou = *z_PVNT_data - *position_z_data;

    /*sets up the initial conditions based on the x,y,z input from the
    *PVNT update*/
    *new_Px_est_tou = *x_PVNT_data;
    *new_Vx_est_tou = *velocity_x_data;
    *new_Py_est_tou = *y_PVNT_data;
    *new_Vy_est_tou = *velocity_y_data;
    *new_Pz_est_tou = *z_PVNT_data;
    *new_Vz_est_tou = *velocity_z_data;

    /*sets up time output for Euler_data file*/
    delay = index;
    time_index = *clock_Ptrs[0] - (delay * dT);
for (i=0; i<index; i++) /*increments counter from 0 to the
maximum value of the index*/
{
    if (i != 0) /*allows normal integration after first iteration*/
    {
        delta_x_tou = 0.0;
        delta_y_tou = 0.0;
        delta_z_tou = 0.0;
    }
    Euler_integration(k1, k2, delta_x_tou, delta_y_tou, delta_z_tou, time_step,
new_Px_est_tou, new_Vx_est_tou, new_Py_est_tou, new_Vy_est_tou,
new_Pz_est_tou, new_Vz_est_tou, &temp_new_Px_est_tou,
&temp_new_Vx_est_tou, &temp_new_Py_est_tou, &temp_new_Vy_est_tou,
&temp_new_Pz_est_tou, &temp_new_Vz_est_tou);
    /*increments the new_ro_est_tou and new_V_sca_est_tou buffer
    *pointers*/
    new_Px_est_tou++;
    new_Vx_est_tou++;;
    new_Py_est_tou++;;
    new_Vy_est_tou++;;
    new_Pz_est_tou++;;
    new_Vz_est_tou++;;
    /*sets the now incremented buffers equal to the results from
    *the forward Euler integration*/
    *new_Px_est_tou = temp_new_Px_est_tou;
    *new_Vx_est_tou = temp_new_Vx_est_tou;
    *new_Py_est_tou = temp_new_Py_est_tou;
    *new_Vy_est_tou = temp_new_Vy_est_tou;
    *new_Pz_est_tou = temp_new_Pz_est_tou;
    *new_Vz_est_tou = temp_new_Vz_est_tou;
    /*prints Euler integration data to the output file for later
    *comparison to actual target data*/
    //          fprintf(Euler_output_data,  "%f %f %f %f %f %f \n", time_index,
(float)*new_Px_est_tou, (float)*new_Py_est_tou, (float)*new_Pz_est_tou,
(float)*new_Vx_est_tou, (float)*new_Vy_est_tou, (float)*new_Vz_est_tou);
    time_index = time_index + dT;
}
/*resets the initial velocity and position values that will go to
*the open loop filter during the next function iteration.*/
initial_x_velocity = ssSetRWorkValue(S, 0, *new_Vx_est_tou);
initial_y_position = ssSetRWorkValue(S, 1, *new_Px_est_tou);
initial_y_velocity = ssSetRWorkValue(S, 2, *new_Vy_est_tou);
initial_y_position = ssSetRWorkValue(S, 3, *new_Py_est_tou);
initial_z_velocity = ssSetRWorkValue(S, 4, *new_Vz_est_tou);
initial_z_position = ssSetRWorkValue(S, 5, *new_Pz_est_tou);
reset_index = 1; /*triggers flag to indicate that an index
*reset is needed*/
integrator_flag = ssSetIWorkValue(S, 1, 1); /*triggers open loop
if((index <= (int)MAX_INDEX) && (reset_index==0))
{
    index = ssSetWorkValue(S, 0, index+1); /*increments index value*/
    velocity_x_data++; /*increments buffer pointers*/
    position_x_data++;
    velocity_y_data++;
    position_y_data++;
    velocity_z_data++;
    position_z_data++;
    x_PVNT_data++;
    y_PVNT_data++;
    z_PVNT_data++;
}
else
{
    new_Px_est_tou=new_Px_est_tou-index;
    new_Vx_est_tou=new_Vx_est_tou-index;
    new_Py_est_tou=new_Py_est_tou-index;
    new_Vy_est_tou=new_Vy_est_tou-index;
    new_Pz_est_tou=new_Pz_est_tou-index;
    new_Vz_est_tou=new_Vz_est_tou-index;

    index = ssSetWorkValue(S, 0, 0); /*resets index value to 0*/

    ssSetPWorkValue(S, 12, (real_T *)x_PVNT_data);
    ssSetPWorkValue(S, 13, (real_T *)y_PVNT_data);
    ssSetPWorkValue(S, 14, (real_T *)z_PVNT_data);
}

/* Function: mdlUpdate ===============================
 * Abstract:
 *    Increment the state for next time around (i.e. a counter).
 */
#define MDL_UPDATE
static void mdlUpdate(SimStruct *S, int_T tid)
{
    UNUSED_ARG(tid); /* not used in single tasking mode */
}
/* Abstract: 
  * Required to have this routine. 
  */
static void mdlTerminate(SimStruct *S)
{
    real_T *velocity_x_data = ssGetPWorkValue(S, 0);
    real_T *position_x_data = ssGetPWorkValue(S, 1);
    real_T *velocity_y_data = ssGetPWorkValue(S, 2);
    real_T *position_y_data = ssGetPWorkValue(S, 3);
    real_T *velocity_z_data = ssGetPWorkValue(S, 4);
    real_T *position_z_data = ssGetPWorkValue(S, 5);
    real_T *new_Vx_est_tou = ssGetPWorkValue(S, 6);
    real_T *new_Px_est_tou = ssGetPWorkValue(S, 7);
    real_T *new_Vy_est_tou = ssGetPWorkValue(S, 8);
    real_T *new_Py_est_tou = ssGetPWorkValue(S, 9);
    real_T *new_Vz_est_tou = ssGetPWorkValue(S, 10);
    real_T *new_Pz_est_tou = ssGetPWorkValue(S, 11);
    real_T *x_PVNT_data = ssGetPWorkValue(S, 12);
    real_T *y_PVNT_data = ssGetPWorkValue(S, 13);
    real_T *z_PVNT_data = ssGetPWorkValue(S, 14);

    UNUSED_ARG(S); /* unused input argument */

    /*releases data stored in buffers*/
    free(velocity_x_data);
    free(position_x_data);
    free(velocity_y_data);
    free(position_y_data);
    free(velocity_z_data);
    free(position_z_data);
    free(new_Vx_est_tou);
    free(new_Px_est_tou);
    free(new_Vy_est_tou);
    free(new_Py_est_tou);
    free(new_Vz_est_tou);
    free(new_Pz_est_tou);
    free(x_PVNT_data);
    free(y_PVNT_data);
    free(z_PVNT_data);

    /*closes Euler integration data output file*/
    fclose(Euler_output_data);
}

#ifdef  MATLAB_MEX_FILE     /* Is this file being compiled as a MEX-file? */
#include "simulink.c"       /* MEX-file interface mechanism */
#endif
#else
#include "cg_sfun.h"    /* Code generation registration function */
#endif
B. ROAD FOLLOWING FILTER S-FUNCTION

The road following filter is similar in method to the general filter but contains two major differences. First, the S-function receives the PVNT update input just like the general filter S-function, but it utilizes an optimization routine before passing the position update on to the remainder of the program. Since the filter can use the road equations in its calculations, it is able to calculate the best position update in terms of the road parameter, $\rho$. The optimization function uses a dichotomy method to quickly and accurately find the best $\rho$ value pertaining to the PVNT $x, y, z$ input. The dichotomy method divides the area of the road it is to search in half and uses a step size to define two points on either side of the halfway mark. The function then calculates and compares the distance from these points to the inputted PVNT update. Using the results, the function will reset either the lower or upper boundary and repeat the calculations until a pre-determined tolerance is met. This results in a routine that is much faster than and just as accurate as calculating and comparing distances from each point within a given range along the road to the inputted PVNT update.

The second main difference between the filters is that all of the integration calculations are done using $\rho$ and velocity instead of the $x, y, z$ coordinates and magnitudes. Once again, this is only possible because the road equations are known before the system is simulated.

1. Manual

File: s_filter_road_following.c

Lines 26-57
Complete basic program initializations, library calls, and global variable input.

Lines 26 and 27
Designate the file name and indicate that the file is in C code, to be converted into MEX format and run in MATLAB.
Lines 29-34
Make all the necessary library calls that are required in the program.

Lines 40-42
Take in the three S-function parameters: \textit{MAX\_DELAY}, \textit{TIME\_STEP}, and \textit{ROAD\_EQUATION} from the S-function block in the Simulink model. \textit{MAX\_DELAY} is the maximum amount of expected delay in between PVNT updates while \textit{TIME\_STEP} is the time step to be used by the C code. NOTE: The time step parameter value must match the discrete time step value found on the simulation parameters menu in Simulink. The \textit{ROAD\_EQUATION} parameter is used to define which road model equations are used in the optimization function (there were two different road models used during initial simulation and testing).

Lines 46-48
Convert the parameters into “real\_T” format for use in numerical calculations later.

Line 52
Defines the global variable \textit{MAX\_INDEX}, used to ensure that buffer overflow does not occur.

Lines 53-56
Define the coefficients for the third order road model and the radius of the circle (meters) for the circular road model.

Line 57
Defines \textit{pi} as a constant for used in the phase shift of the circular road model equations found in the optimization function (see lines 103 and 107)

\textit{PVNT\_optimization} function

Line 69
Lists the inputs to the function along with buffers marked by an asterisk in front of their names.
Lines 71-76
Define and initialize the variables that are only used inside the function such as the upper and lower bounds and the desired tolerance of the final result.

Lines 79-80
Initialize the upper and lower bounds before the dichotomy loop. The ro_optimize_start real work value is taken in from the mdlOutputs function (see line 455).

Lines 85-126
Contain the dichotomy loop.

Line 85
Defines limits the amount of loop iterations to 100 and dictates that the loop should continue until the required tolerance is met.

Lines 87-88
Define the upper and lower $\rho$ limits by dividing the search area of the road in half and adding and subtracting the step size.

Lines 90-99
Compute the position points along the road based on the upper and lower $\rho$ limits if the third order road model is being used based on the inputted parameter from line 42.

Lines 101-110
Compute the position points along the road based on the upper and lower $\rho$ limits if the circular road model is being used based on the inputted parameter from line 42.

Lines 112-113
Use the distance formula to compute the distance between the PVNT position input and the calculated position points. The if/else loops in lines 115-122 compare the distance values and, based on the results, reset the right or left boundary to one of the $\rho$ limits.

Lines 124 and 125
Compute the current tolerance and increments the counter pertaining to the while loop.

**Line 127**

Sends out the new \( \rho \) update value following successful completion of the dichotomy loop.

**Euler_integration function:**

**Line 137**

Lists the inputs to the function along with buffers marked by an asterisk in front of their names.

**Lines 144-145**

Perform forward Euler integration for the \( \rho \) variable, assigning the new position and velocity values to the buffers beginning with “temp.”

**mdlInitializeSizes function**

**Line 155**

Means that there will be three parameters inputted into the S-function block in Simulink.

**Lines 156-159**

Return an error to MATLAB if the incorrect number of parameters is found.

**Line 161**

Defines zero continuous states since the model is running with a preset, fixed step time.

**Line 162**

Defines seven discrete states which must match the number of input ports found in line 164.

**Lines 165-171**

Set the size of each input port.

**Lines 172-178**

Denote each input port as a direct feed through port.
Line 180
Defines four output ports from the S-function.

Lines 181-184
Define the width of each port.

Line 186
Defines one sample time to be used

Lines 187-190
Define the number of real, integer, pointer, and mode work vectors to be used in the program. The work vectors can be thought of as a value of a certain type (real, integer, pointer, etc.) that is stored in persistent memory. This means that the value will be stored even while the program is called multiple times.

Line 191
Defines the number of zero crossings to be zero as it is not used in the filter program.

mdllInitializeSampleTimes function

Line 205
Defines the program’s sample time to be set to $dT$, which come from the second parameter input to the S-function block in line 46.

Line 206
Indicates a 0.0 second offset time and line 208 indicates that a function call is made on the first element of the first output port.

mdllStart function

Defines all of the variables that need to be initialized only once, i.e. the very first time the program is run in the simulation.

Line 222
Predefines the integer work vectors for the index counter and the integrator flag that indicates when the discrete integrator blocks in the open-loop filter subsystem need to be reset.
Line 223
Predefines the real work vectors for the initial x, y, and z positions and velocities.

Lines 228-230
Predefine the buffers that are used in the code for data storage

Lines 235-241
Initialize the buffers to a number of positions equal to MAX_INDEX (from line 52) with each position having enough memory to store a piece of data with the size real_T. The calloc command also initializes every position in the buffers to zero.

Lines 243-247
Define the first value for the index counter, integrator reset flag, position and velocity initial conditions, and the starting $\rho$ variable for the optimization function to be zero.

Lines 251-257
Set the pointer work vectors to point to the first position of each of the buffers

Lines 259-267
Set and store the initial integer and real work values.

 mdlOutputs function

Lines 278-294
Contain the input and output declarations.

Lines 278-279 and 285-286
Define the pointers and values of the position and velocity estimates coming from the open loop filter function call.

Lines 280 and 287
Define the pointer and value coming in from the PVNT update delay subsystem

Lines 281-283 and 288-290
Define the actual PVNT update (x,y,z) from the target model subsystem.

Lines 284 and 291
Designate a port for the clock input.
Define output ports for the position and velocity initial conditions to the open-loop filter function call.

Line 294
Defines the integrator reset signal, which is also fed into the open-loop filter function call.

Lines 296-303
Contain declarations for the work values and the buffers which match the declarations found in the mdlStart function.

Lines 309-319
Define and initialize the non-persistent variables that are used only in the Euler_integration, PVNT_optimization, and mdlOutputs function.

Lines 331-337
Retrieves the values that were stored in the pointer work vectors.

Lines 340-344
Retrieves the values that were stored in the integer and real work vectors.

Lines 352-389
Contained in an if loop that executes only if the index counter is less than or equal to the preset MAX_INDEX value. This ensures that no data is written to the buffers beyond their maximum preset number of storage positions, reducing the risk of buffer overflow.

Lines 354-355
Set the second and third output ports to the position and velocity initial conditions.

Line 357
Sets the integrator_reset output to the integer work value integrator_flag.

Lines 360-364
Call the open loop filter function call block in the Simulink diagram through the first output port.

Lines 366-367
Take in the estimated position and velocity values from the first six inputs (arriving from the outputs of the open-loop filter function call).

**Lines 370-371**
Set the respective buffer values to the inputted position and velocity estimates. These values are then also stored in the real work vectors designating position and velocity initial conditions.

**Line 378**
Resets the integrator flag integer work value to zero (if it was set to one following the Euler integration loop, see line 460).

**Lines 381-383**
Take in the PVNT position update \((x,y,z)\) from the true target model subsystem in the Simulink diagram.

**Lines 386-388**
Assign PVNT position update to buffers.

**Lines 391-447**
Contained in an if loop that is only triggered if the input from the PVNT delay subsystem is set high, indicating that a PVNT update is available.

**Lines 397-399**
Adjust the pointers to each PVNT buffer so that they now refer to time \(\tau\). This is controlled by the *index* integer work vector which is incremented after each iteration of the `mdlOutputs` function (see line 471).

**Line 402**
Calls the `PVNT_optimization` function and receives the new \(\rho\) value.

**Lines 407-408**
Adjust the pointers to each position and velocity buffer so that they now refer to time \(\tau\), the time to which the PVNT update refers.

**Line 412**
Calculates the difference between the estimated position data at time \(\tau\) and the PVNT position update at time \(\tau\) for \(\rho\).

**Lines 416-417**
Set up the values for the first position of the buffers that are used in the\n\textit{Euler\_integration} function and to pass on the updated position and velocity data\nto the open-loop filter function call.

\textbf{Line 423}\n
Begins the asynchronous portion of the S-function. The \textit{for} loop runs enough\ntimes to move the new estimated position and velocity values from time $\tau$ to time\ntime $t$ (current system time), which is controlled by the \textit{index} integer work vector\nvalue.

\textbf{Lines 426-429}\n
Set the \textit{delta} variable value originally set in \textbf{lines 412} to zero after the first\niteration of the \textit{for} loop, allowing for normal, dead-reckoning style integration.

\textbf{Line 431}\n
Passes the required variables to the \textit{Euler\_integration} function in \textbf{lines 131-146}.\nAdditionally, the “&” in front of the \textit{temp} buffers indicate that their changed\nvalues from the \textit{Euler\_integration} function will be saved after the function\nexecutes.

\textbf{Lines 435-436}\n
Increment the pointer values for the buffers that will contain the updated position\nand velocity estimates.

\textbf{Lines 440-441}\n
Actually set the buffers equal to the updates.\nAfter the \textit{for} loop runs the appropriate number of times to arrive at time $t$, the\nfinal value from each of the buffers containing the updated position and velocity\nestimates are passed to the initial condition real work vectors in \textbf{lines 451-452}.

\textbf{Line 455}\n
The \textit{ro\_optimize\_start} real work value (used in the optimization function) is set.\nAdditionally, two integrator reset values are set. The first is the \textit{reset\_index}\nvariable on \textbf{line 457} set equal to one and used inside the S-function program on\n\textbf{line 464}.\n
125
The second is the `integrator_flag` integer work value on line 460 that is outputted to the open-loop function call outside the S-function block. The remainder of the buffer pointer incrementation/resets take place in the `if/else` loop in lines 464-478. The `if` loop portion checks to see if the current `index` variable value is less than the preset `MAX_INDEX` value and if the `reset_index` variable value is equal to zero (indicating that a PVNT update did not arrive during the current `mdlOutputs` function iteration. If so, the `index` integer work value is incremented along with the pointers to the position and velocity data buffers.

If the criterion for the `if` loop are not met, meaning that a PVNT update has occurred, the buffer pointers are all reset back to their first position and the `index` integer work value is set to zero. This ensures that the buffers are simply overwritten with the new data until the next PVNT update and buffer overflow does not occur. Finally, the pointer work values are updated to now designate the new pointer values for the position and velocity data buffers.

**Lines 482-486**

Reset the pointer work values for the PVNT, position, and velocity buffers.

`mdlUpdate` function

This would be the function in which states would be incremented if they were used in the program. Since the filter design does not use these states, however, the `mdlUpdate` function is only left in the program as a formality.

`mdlTerminate` function

In this case, all of the data from the buffers must be cleared to avoid errors when re-running the simulation multiple times.

**Lines 510-516**

Designate each of the buffers that were originally defined in the `mdlStart` function

**Lines 523-529**
Actually release the data stored in the buffers.

2. Code

/* File : s_filter_road_following.c */
/* Abstract: */
/* This S-function is a combination of an open-loop filter using a */
/* function call subsystem and an asynchronous filter contained in the */
/* C code of the S-function. The model is used for a target tracking */
/* system, utilizing a delayed position update at different time */
/* intervals. When the position update (labeled PVNT) is not */
/* available, the S-function calls the open-loop filter and stores the */
/* results. When the delayed position update arrives, the loop */
/* containing the asynchronous filter is run to update the previous */
/* data from time tau (corresponding to the PVNT update) to time t */
/* (corresponding to the current time) using buffers to store all */
/* data. The model takes in parameters from the S-function block in */
/* the Simulink model for the maximum amount of delay (seconds) and */
/* the desired time step (seconds). The user can easily manipulate */
/* these parameters without having to change C code in the S-function */
/* */
/* For more details about S-functions, see */
/* matlabroot/simulink/src/sfuntmpl_doc.c */
/* Copyright 1990-2006 The MathWorks, Inc. */
/* $Revision: 1.15.4.3 $ */
/* */
#define S_FUNCTION_NAME s_filter_road_following
#define S_FUNCTION_LEVEL 2
#include "simstruc.h"
#include <stdlib.h>
#include <stdio.h>
#include <string.h>
#include <math.h>

/* Input Arguments */
/* takes in parameters that define a max value for the PVNT update delay and */
/* the desired time step */
#define MAX_DELAY ssGetSFcnParam(S,0)
#define TIME_STEP ssGetSFcnParam(S,1)
#define ROAD_EQUATION ssGetSFcnParam(S,2)

/* converts the above parameters from structs to allow them to be used in */
/* computations*/
#define dT ((real_T) mxGetPr(TIME_STEP)[0])
#define DELAY_MAX ((real_T) mxGetPr(MAX_DELAY)[0])
#define road_equation_selection ((real_T) mxGetPr(ROAD_EQUATION)[0])

/* defines global constant that is used to prevent buffer overflow and */
coefficients for road equation*/
define MAX_INDEX (DELAY_MAX/dT)
define coeff_3 0.0000192
define coeff_2 -0.025
define coeff_1 9.74
define radius 2865.0
define pi 3.14159

/* Function: PVNT_optimization
 * Abstract:
 * Performs distance measurement between the original PVNT update
 * coordinates and coordinates defined by the road equation. It then finds
 * the closest point on the road to the PVNT coordinates and sets that
 * point as the actual PVNT position update. The third parameter in the
 * S-function block determines which optimization equation is called
 * based on which road equation is to be used.
 */
void PVNT_optimization (real_T *x_PVNT_data, real_T *y_PVNT_data, real_T *z_PVNT_data, real_T ro_optimize_start, real_T *ro_star, int road_eq_selector)
{
    float lower_ro_limit=0.0, upper_ro_limit=0.0;
    float x_left=0.0, y_left=0.0, z_left=0.0, x_right=0.0, y_right=0.0, z_right=0.0;
    float distance_1=0.0, distance_2=0.0, step_size=0.5;
    float left_boundary = 0.0, right_boundary = 0.0, tolerance = 0.000001;
    float L = 2*tolerance;  /*sets L so it is initially higher than tolerance*/
    int j = 0;
   
    /*initializes upper and lower bounds for optimization loop*/
    left_boundary = (float)ro_optimize_start - 50.0;
    right_boundary = (float)ro_optimize_start + 50.0;
    
    /*optimization routine for PVNT update: utilizes dichotomy technique to
    * compare distance from points along the road model to PVNT update
    * point. final value is outputted as the ro_star update*/
    while (L>=tolerance && j<=100)
    {
        lower_ro_limit = (right_boundary+left_boundary-step_size)/2.0;
        upper_ro_limit = (right_boundary+left_boundary+step_size)/2.0;

        if (road_eq_selector == 0)
        {
            x_left = lower_ro_limit;
            y_left = coeff_3*pow(lower_ro_limit,3) + coeff_2*pow(lower_ro_limit,2) + coeff_1*lower_ro_limit;
            z_left = 0.0;
            
            x_right = upper_ro_limit;
            y_right = coeff_3*pow(upper_ro_limit,3) + coeff_2*pow(upper_ro_limit,2) + coeff_1*upper_ro_limit;
            z_right = 0.0;
        }
        if (road_eq_selector == 1)
x_left = radius + radius * sin(lower_ro_limit/radius + 3*pi/2);
"y_left = radius*sin(lower_ro_limit/radius);
z_left = 0.0;

x_right = radius + radius * sin(upper_ro_limit/radius + 3*pi/2);
y_right = radius*sin(upper_ro_limit/radius);
z_right = 0.0;

distance_1 = sqrt(pow(x_left-*x_PVNT_data,2) + pow(y_left-*y_PVNT_data,2) +
pow(z_left-*z_PVNT_data,2));
distance_2 = sqrt(pow(x_right-*x_PVNT_data,2) + pow(y_right-*y_PVNT_data,2) +
pow(z_right-*z_PVNT_data,2));

if(distance_1 <= distance_2)
{
   right_boundary = upper_ro_limit;
}
else
{
   left_boundary = lower_ro_limit;
}

L = fabs(distance_1 - distance_2); /*computes current error*/
j++; /*increments counter*/

*ro_star = (left_boundary+right_boundary)/2;

/* Function: Euler_integration ===============================
 * Abstract:
 * Performs asynchronous forward Euler integration once the PVNT update is
 * received in order to rewrite over the previous data from time tau to
 * time t.
 */
void Euler_integration(double k1, double k2, float delta_ro_tou, float time_step, real_T 
*new_ro_est_tou, real_T *new_V_sca_est_tou, real_T *temp_new_V_sca_est_tou, 
real_T *temp_new_ro_est_tou)
{

/*performs asynchronous double integration with a time step
 *equal to dT seconds and stores the results in a temp variable
 *to be transferred to the buffers after they have been
 *incremented*/
*temp_new_ro_est_tou = *new_ro_est_tou+ (*new_V_sca_est_tou + 
k1*delta_ro_tou)*time_step;
*temp_new_V_sca_est_tou =*new_V_sca_est_tou+ (k2*delta_ro_tou)*time_step;
}

/* Function: mdlInitializeSizes ================================
 * Abstract:
 * Setup sizes of the various vectors.
 */
static void mdlInitializeSizes(SimStruct *S)
{
    ssSetNumSFcnParams(S, 3); /* Number of expected parameters */
    if (ssGetNumSFcnParams(S) != ssGetSFcnParamsCount(S))
        return; /* Parameter mismatch will be reported by Simulink */

    ssSetNumContStates(S, 0);               /*defines 0 continuous states*/
    ssSetNumDiscStates(S, 7);               /*defines 7 discrete state*/

    if (!ssSetNumInputPorts(S, 7)) return;  /*defines 7 input ports*/
    ssSetInputPortWidth(S, 0, 1);           /*sets input 1 port size to 1*/
    ssSetInputPortWidth(S, 1, 1);           /*sets input 2 port size to 1*/
    ssSetInputPortWidth(S, 2, 1);           /*sets input 3 port size to 1*/
    ssSetInputPortWidth(S, 3, 1);           /*sets input 4 port size to 1*/
    ssSetInputPortWidth(S, 4, 1);           /*sets input 5 port size to 1*/
    ssSetInputPortWidth(S, 5, 1);           /*sets input 6 port size to 1*/
    ssSetInputPortWidth(S, 6, 1);           /*sets input 7 port size to 1*/
    ssSetInputPortDirectFeedThrough(S, 0, 1);
    ssSetInputPortDirectFeedThrough(S, 1, 1);
    ssSetInputPortDirectFeedThrough(S, 2, 1);
    ssSetInputPortDirectFeedThrough(S, 3, 1);
    ssSetInputPortDirectFeedThrough(S, 4, 1);
    ssSetInputPortDirectFeedThrough(S, 5, 1);
    ssSetInputPortDirectFeedThrough(S, 6, 1);

    if (!ssSetNumOutputPorts(S, 4)) return;
    ssSetOutputPortWidth(S, 0, 1);       /*sets output port 1 width to 1*/
    ssSetOutputPortWidth(S, 1, 1);       /*sets output port 2 width to 1*/
    ssSetOutputPortWidth(S, 2, 1);       /*sets output port 3 width to 1*/
    ssSetOutputPortWidth(S, 3, 1);       /*sets output port 4 width to 1*/

    ssSetNumSampleTimes(S, 1);
    ssSetNumRWork(S, 3);                 /*real vector*/
    ssSetNumIWork(S, 2);                 /*integer vector*/
    ssSetNumPWork(S, 7);                 /*pointer vector*/
    ssSetNumModes(S, 0);                 /*mode vector*/
    ssSetNumNonsampledZCs(S, 0);         /*number of zero crossings*/

    /* Take care when specifying exception free code - see sfuntmpl_doc.c */
    ssSetOptions(S, SS_OPTION_EXCEPTION_FREE_CODE);
}

/* Function: mdlInitializeSampleTimes =====================================
 * Abstract:
 *    Discrete sample time of dT seconds and specify that we are doing
 *    function-calls on the 1st element of the 1st output port.
 */
static void mdlInitializeSampleTimes(SimStruct *S)
{
    ssSetSampleTime(S, 0, dT); /*sets sample time to dT seconds*/
    ssSetOffsetTime(S, 0, 0.0); /*indicates 0 offset time*/
ssSetCallSystemOutput(S, 0); /* call on first element */
ssSetModelReferenceSampleTimeDefaultInheritance(S);
}

/*Function: mdlStart ==============================*/

/*Abstract:
 * This function sets up the variables passed between the function and
 * the s-function.
*/
#define MDL_START

static void mdlStart(SimStruct *S)
{
  int_T index, integrator_flag;
  real_T initial_position, initial_velocity, ro_optimize_start;

  /*The real_T variables below denote the buffers used to store
   *velocity and position data over multiple iterations of the
   *s-function*/
  real_T *velocity_data, *position_data;
  real_T *new_V_sca_est_tou, *new_ro_est_tou;
  real_T *x_PVNT_data, *y_PVNT_data, *z_PVNT_data;

  /*The buffers are allocated enough memory to store 'MAX_INDEX' data
   *with each data space being 'real_T' size. The 'calloc' command also
   *initializes the buffers*/
  velocity_data = (real_T *) calloc(MAX_INDEX, sizeof(real_T));
  position_data = (real_T *) calloc(MAX_INDEX, sizeof(real_T));
  new_V_sca_est_tou = (real_T *) calloc(MAX_INDEX, sizeof(real_T));
  new_ro_est_tou = (real_T *) calloc(MAX_INDEX, sizeof(real_T));
  x_PVNT_data = (real_T *) calloc(MAX_INDEX, sizeof(real_T));
  y_PVNT_data = (real_T *) calloc(MAX_INDEX, sizeof(real_T));
  z_PVNT_data = (real_T *) calloc(MAX_INDEX, sizeof(real_T));

  index = 0;                           /*initializes index to 0*/
  integrator_flag = 0;                //sets integration reset flag to 0
  initial_velocity = 0.0;             //initializes position and velocity
  initial_position = 0.0;              //IC's to 0
  ro_optimize_start = 0.0;

  /*Sets the pointer work variables for the buffers*/
  ssSetPWorkValue(S, 0, (real_T *)velocity_data);
  ssSetPWorkValue(S, 1, (real_T *)position_data);
  ssSetPWorkValue(S, 2, (real_T *)new_V_sca_est_tou);
  ssSetPWorkValue(S, 3, (real_T *)new_ro_est_tou);
  ssSetPWorkValue(S, 4, (real_T *)x_PVNT_data);
  ssSetPWorkValue(S, 5, (real_T *)y_PVNT_data);
  ssSetPWorkValue(S, 6, (real_T *)z_PVNT_data);

  ssSetIWorkValue(S, 0, index);       /*sets the first integer work
   *value to the index variable*/
ssSetIWorkValue(S, 1, integrator_flag); /*sets the second integer work
     *value to the integrator flag*/

ssSetRWorkValue(S, 0, initial_velocity); /*sets the real work values*/
ssSetRWorkValue(S, 1, initial_position);
ssSetRWorkValue(S, 2, ro_optimize_start);

/* Function: mdlOutputs ================================================
   * Abstract:
   *    Issue ssCallSystemWithTid on 1st output element of 1st output port.
   */
static void mdlOutputs(SimStruct *S, int_T tid)
{
    /*S-function input and output declarations*/
    real_T *ro_est          = ssGetRealDiscStates(S,0);
    real_T *V_sca_est      = ssGetRealDiscStates(S,1);
    real_T *PVNT            = ssGetRealDiscStates(S,2);
    real_T *x_PVNT          = ssGetRealDiscStates(S,3);
    real_T *y_PVNT          = ssGetRealDiscStates(S,4);
    real_T *z_PVNT          = ssGetRealDiscStates(S,5);
    real_T *clock           = ssGetRealDiscStates(S,6);
    InputRealPtrsType ro_est_Ptrs      = ssGetInputPortRealSignalPtrs(S,0);
    InputRealPtrsType V_sca_est_Ptrs    = ssGetInputPortRealSignalPtrs(S,1);
    InputRealPtrsType PVNT_Ptrs         = ssGetInputPortRealSignalPtrs(S,2);
    InputRealPtrsType x_PVNT_Ptrs       = ssGetInputPortRealSignalPtrs(S,3);
    InputRealPtrsType y_PVNT_Ptrs       = ssGetInputPortRealSignalPtrs(S,4);
    InputRealPtrsType z_PVNT_Ptrs       = ssGetInputPortRealSignalPtrs(S,5);
    InputRealPtrsType clock_Ptrs        = ssGetInputPortRealSignalPtrs(S,6);
    real_T *TgtP_IC         = ssGetOutputPortRealSignal(S,1);
    real_T *TgtV_IC         = ssGetOutputPortRealSignal(S,2);
    real_T *integrator_reset = ssGetOutputPortRealSignal(S,3);

    int_T index, integrator_flag;
    real_T initial_velocity, initial_position;
    real_T temp_new_V_sca_est_tou, temp_new_ro_est_tou;

    /*buffer declarations for mdlOutputs*/
    real_T *velocity_data, *position_data;
    real_T *new_V_sca_est_tou, *new_ro_est_tou;
    real_T *x_PVNT_data , *y_PVNT_data, *z_PVNT_data;

    /*defines pointer to output file for forward Euler integration results*/
    //FILE *Euler_output_data;

    /*defines intermediate position and velocity matrices*/
    float delta_ro_tou = 0.0, time_index = 0.0;
    int i = 0; /*counter*/
    int road_eq_selector = road_equation_selection; /*picks which road
     *equation is to be used*/
    int reset_index = 0; /*flag indicating and index reset to 0*/
    double k1=0.5, k2=0.5; /*sets integrator gains*/
    float delay = 0.0, time_step = dT;

    for (i = 0; i < num_time_steps; i++) { /*main loop*/
        /*First integrator step*/
        temp_new_V_sca_est_tou = k1 * V_sca_est + k2 * PVNT;
        temp_new_ro_est_tou    = k1 * ro_est + k2 * PVNT;

        /*Second integrator step*/
        V_sca_est += temp_new_V_sca_est_tou;
        ro_est += temp_new_ro_est_tou;

        /*Update output with new values*/
        *TgtP_IC = x_PVNT;
        *TgtV_IC = y_PVNT;
        *integrator_reset = i;

        /*Advance forward with Euler integration for current time step*/
        delay += time_step;
        time_index += time_step;

        /*Handle index reset*/
        if (reset_index == 0) {
            /*Reset index*/
            delay = 0.0;
            time_index = 0.0;
            reset_index = 1;
        }
    }
}
/*optimization variables*/
real_T ro_star, ro_optimize_start;

/* ssCallSystemWithTid is used to execute a function-call subsystem. The
* 2nd argument is the element of the 1st output port index which
* connected to the function-call subsystem. Function-call subsystems
* can be driven by the first output port of s-function blocks.
*/

UNUSED_ARG(tid); /* not used in single tasking mode */

/*Retrieves the pointer work values for the buffers*/
velocity_data        = (real_T *)ssGetPWorkValue(S, 0);
position_data        = (real_T *)ssGetPWorkValue(S, 1);
new_V_sca_est_tou   = (real_T *)ssGetPWorkValue(S, 2);
new_ro_est_tou       = (real_T *)ssGetPWorkValue(S, 3);
x_PVNT_data          = (real_T *)ssGetPWorkValue(S, 4);
y_PVNT_data          = (real_T *)ssGetPWorkValue(S, 5);
z_PVNT_data          = (real_T *)ssGetPWorkValue(S, 6);

/*Retrieves integer and real work values*/
index                = ssGetIWorkValue(S,0);
integrator_flag      = ssGetIWorkValue(S,1);
initial_velocity     = ssGetRWorkValue(S,0);
initial_position     = ssGetRWorkValue(S,1);
ro_optimize_start    = ssGetRWorkValue(S,2);

/*creates .txt file for output results*/
//Euler_output_data = fopen("Euler_data_rf.txt", "w");

/*Entire sequence is in an 'if' loop to ensure that there is no
overflow for the position and velocity arrays (defined with a maximum
of MAX_INDEX data points.)*/
if(index <= (int)MAX_INDEX)
{
    TgtP_IC[0] = initial_position;    /*sets outputs to initial V and P*/
    TgtV_IC[0] = initial_velocity;
    integrator_reset[0] = integrator_flag;    /*sets output 3 to integration
    *reset flag*/

    if(!ssCallSystemWithTid(S,0,tid))    /*calls system with task ID 1*/
        { /* Error occurred which will be reported by Simulink */
            return;
        }

    ro_est_Ptrs = ssGetInputPortRealSignalPtrs(S,0);    /*Gets inputs*/
    V_sca_est_Ptrs = ssGetInputPortRealSignalPtrs(S,1);

    /*assigns the position and velocity data to the buffers*/
    *position_data = (real_T*)ro_est_Ptrs[0];
    *velocity_data = (real_T*)V_sca_est_Ptrs[0];
/*resets the initial velocity and position values*/
initial_velocity = ssSetRWorkValue(S, 0, (real_T)*V_sca_est_Ptrs[0]);
initial_position = ssSetRWorkValue(S, 1, (real_T)*ro_est_Ptrs[0]);

/*resets the integrator reset to 0*/
integrator_flag = ssSetIWorkValue(S, 1, 0);

/*takes in x, y, and z coordinates from PVNT update*/
x_PVNT_Ptrs = ssGetInputPortRealSignalPtrs(S,3);
y_PVNT_Ptrs = ssGetInputPortRealSignalPtrs(S,4);
z_PVNT_Ptrs = ssGetInputPortRealSignalPtrs(S,5);

/*assigns coordinates to buffers*/
*x_PVNT_data = (real_T)*x_PVNT_Ptrs[0];
*y_PVNT_data = (real_T)*y_PVNT_Ptrs[0];
*z_PVNT_data = (real_T)*z_PVNT_Ptrs[0];

PVNT_optimization (x_PVNT_data, y_PVNT_data, z_PVNT_data,
ro_optimize_start, &ro_star, road_eq_selector);

/*calls the estimated ro and velocity values at time tou from the
*buffers*/
position_data = position_data - index;
velocity_data = velocity_data - index;

/*calculates the difference between the ro_star update value and the
*estimated ro value at time tou*/
delta_ro_tou = ro_star - *position_data;

/*sets up the initial conditions based on the ro input from the
*PVNT update*/
*new_ro_est_tou = ro_star;
*new_V_sca_est_tou = *velocity_data;

/*sets up time output for Euler_data file*/
delay = index;
time_index = *clock_Ptrs[0] - (delay / dT);

for (i=0; i<index; i++) /*increments counter from 0 to the
*maximum value of the index*/
{
if (i != 0)  /*allows normal integration after first iteration*/
{
    delta_ro_tou = 0.0;
}

Euler_integration(k1, k2, delta_ro_tou, time_step, new_ro_est_tou, new_V_sca_est_tou, &temp_new_V_sca_est_tou, &temp_new_ro_est_tou);

/*increments the new_ro_est_tou and new_V_sca_est_tou buffer *pointers*/
new_ro_est_tou++;
new_V_sca_est_tou++;

/*sets the now incremented buffers equal to the results from *the forward Euler integration*/
new_ro_est_tou = temp_new_ro_est_tou;
new_V_sca_est_tou = temp_new_V_sca_est_tou;

/*prints Euler integration data to the output file for later *comparison to actual target data*/
//fprintf(Euler_output_data, "%f %f %f \n", time_index, (float)*new_ro_est_tou, (float)*new_V_sca_est_tou);
time_index = time_index + dT;
}

/*resets the initial velocity and position values that will go to *the open loop filter during the next function iteration.*/
initial_velocity = ssSetRWorkValue(S, 0, *new_V_sca_est_tou);
initial_position = ssSetRWorkValue(S, 1, *new_ro_est_tou);

/*resets the initial ro value for use in the optimization loop*/
ro_optimize_start = ssSetRWorkValue(S, 2, initial_position);

reset_index = 1;  /*triggers flag to indicate that an index *reset is needed*/
integrator_flag = ssSetIWorkValue(S, 1, 1);  /*integrator reset*/

if((index <= (int)MAX_INDEX) && (reset_index==0))  /*checks to see if flag is set*/
{
    velocity_data++;  /*increments buffer pointers*/
    position_data++;
    x_PVNT_data++;
    y_PVNT_data++;
    z_PVNT_data++;
    index = ssSetIWorkValue(S, 0, index+1);  /*increments index value*/
}
else
{
    new_ro_est_tou=new_ro_est_tou-index;
    new_V_sca_est_tou=new_V_sca_est_tou-index;
    index = ssSetIWorkValue(S, 0, 0);  /*resets index value to 0*/
}
/* resets the pointer work values for the velocity_data and position_data buffers */
ssSetPWorkValue(S, 0, (real_T *)velocity_data);
ssSetPWorkValue(S, 1, (real_T *)position_data);
ssSetPWorkValue(S, 4, (real_T *)x_PVNT_data);
ssSetPWorkValue(S, 5, (real_T *)y_PVNT_data);
ssSetPWorkValue(S, 6, (real_T *)z_PVNT_data);
}

/******************************************************************************
* Function: mdlUpdate
* Abstract:
* Increment the state for next time around (i.e. a counter).
*******************************************************************************/
#define MDL_UPDATE
static void mdlUpdate(SimStruct *S, int_T tid)
{
    UNUSED_ARG(tid); /* not used in single tasking mode */
}

/******************************************************************************
* Function: mdlTerminate
* Abstract:
* Required to have this routine.
*******************************************************************************/
static void mdlTerminate(SimStruct *S)
{
    real_T *velocity_data       = ssGetPWorkValue(S, 0);
    real_T *position_data       = ssGetPWorkValue(S, 1);
    real_T *new_V_sca_est_tou   = ssGetPWorkValue(S, 2);
    real_T *new_ro_est_tou      = ssGetPWorkValue(S, 3);
    real_T *x_PVNT_data         = ssGetPWorkValue(S, 4);
    real_T *y_PVNT_data         = ssGetPWorkValue(S, 5);
    real_T *z_PVNT_data         = ssGetPWorkValue(S, 6);

    //FILE *Euler_output_data;
    UNUSED_ARG(S); /* unused input argument */

    /* releases data stored in buffers */
    free(velocity_data);
    free(position_data);
    free(new_V_sca_est_tou);
    free(new_ro_est_tou);
    free(x_PVNT_data);
    free(y_PVNT_data);
    free(z_PVNT_data);

    /* closes Euler integration data output file */
    //fclose(Euler_output_data);
}
ifdef MATLAB_MEX_FILE /* Is this file being compiled as a MEX-file? */
#include "simulink.c" /* MEX-file interface mechanism */
#else
#include "cg_sfun.h" /* Code generation registration function */
#endif
LIST OF REFERENCES


INITIAL DISTRIBUTION LIST

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   Ft. Belvoir, Virginia

2. Dudley Knox Library
   Naval Postgraduate School
   Monterey, California

3. Prof Anthony Healey
   Chairman, Department of Mechanical and Astronautical Engineering
   Naval Postgraduate School
   Monterey, California

4. Prof Isaac Kaminer
   Naval Postgraduate School
   Monterey, California

5. Dr. Vladimir Dobrokhodov
   Naval Postgraduate School
   Monterey, California

6. Dr. Eng. Ioannis Kitsios
   Naval Postgraduate School
   Monterey, California