Appendix A

Introduction to MATLAB and Simulink

This appendix provides a quick reference for using MATLAB with its toolboxes and Simulink with its blocksets for DSP applications. These tools are used extensively in the experiments and examples in this book. This appendix covers useful topics related to DSP in MATLAB, the Signal Processing Toolbox, the Filter Design Toolbox, Simulink, the DSP Blockset, and the Fixed-Point Blockset. More detailed descriptions are documented in the MATLAB, toolbox, and Simulink user's guides listed in references at the end of this appendix.

A.1 USING MATLAB

MATLAB stands for "MATrix LABoratory" and is a technical-computing language that allows the user to perform numerical computation, simulation, acquisition, and visualization of data, algorithm design, analysis, development, and implementation. Unlike other high-level programming languages such as C, MATLAB provides a comprehensive suite of mathematical, statistical, and engineering functions. The functionality is extended with interactive graphical capabilities for creating plots. Furthermore, extensive toolboxes are available for working under the MATLAB environment. Toolboxes are collections of algorithms, written by experts in their fields, that provide application-specific capabilities. These toolboxes enhance MATLAB's functionality in

signal and image processing, data analysis and statistics, mathematical modeling, control system design, etc.

In addition to MATLAB and its toolboxes, there is another software package called Simulink for modeling, simulating, and analyzing dynamic systems. Simulink is integrated closely with the MATLAB environment. Variables and results derived from Simulink can be put in the MATLAB workspace for postprocessing and visualization. Like the toolboxes that extend MATLAB functions, many blocksets that add additional blocks to the Simulink environment are available. These blocksets include the DSP Blockset, Communication Blockset, and Fixed-Point Blockset.

A.1.1 Startup

To start MATLAB, double-click on the icon on the desktop. A MATLAB window is displayed, as shown in Fig. A.1. This window provides an integrated environment for developing MATLAB code. The command window is the main window in which the user keys MATLAB commands after the prompt >>. Some MATLAB commands are listed as follows:

- The help command is used to show a list of programs installed in the MAT-LAB environment.
- 2. The help topic command is used to display the usage of a particular MAT-LAB topic or syntax.
- 3. The demo command is used to bring up a demo window.

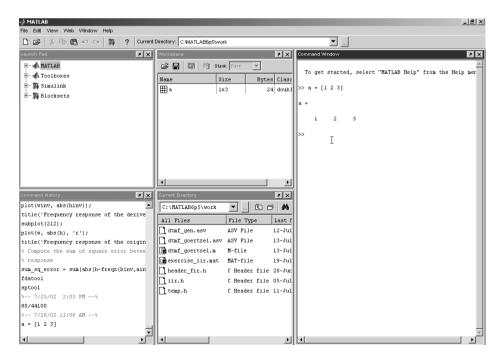
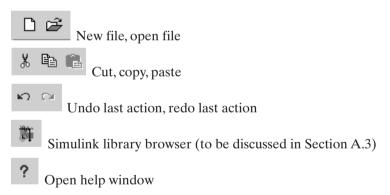


Figure A.1 MATLAB environment (version 6.5) using a five-panel desktop layout

- **4.** The whos command is used to display a list of variables currently loaded into the workspace.
- **5.** The what dir command is used to display the files in the directory dir.
- **6.** The clear all command is used to clear all of the variables.
- 7. The diary ('record.out') command is used create a file called record.out that contains all of the functions entered in the command window. The command diary('off') turns off the record mode.
- **8.** The guit command is used to exit MATLAB.

In addition to the **Command Window** shown in the right side of Fig. A.1, there are **Command History** and **Current Directory** windows located in the bottom left of the screen. The **Command History** window records all of the executed commands, as well as the date and time when these commands were executed. This feature is useful in recalling the commands that have been executed previously. The **Current Directory** window keeps track of the files inside the current directory. In the upperleft screen are two default windows: **Launch Pad** and **Workspace**. The **Launch Pad** window contains all of the installed toolboxes that can be easily launched by double-clicking on the desired toolbox. The **Workspace** window is used to organize the loaded variables and also displays the size of variables, bytes, and class.

A set of toolbars at the top of the screen performs the following functions:



The MATLAB environment also includes the following features under the **File** pull-down menu:

- **1.** Import data from a file directory into the MATLAB workspace.
- 2. Save the MATLAB workspace to a file.
- **3.** Set path, which allows commonly used files in the set directory to be searched.
- **4.** Preference, which allows the user to specify the window's color and font size, number display formats, editor used, print options, figure copy options, etc.

A.1.2 Useful Syntax

MATLAB uses very simple programming notations to execute mathematical statements. MATLAB syntax is expressed in a matrix-style operation. The user

can represent and manipulate scalars, vectors, and matrices in the MATLAB workspace, as shown by the examples summarized in Table A.1.

A.1.3 Plots

MATLAB is known for its outstanding graphics capabilities. Both 2-D and 3-D graphics can be plotted and manipulated using some commonly used commands summarized in Table A.2.

MATLAB version 6 (or higher) provides a set of powerful editing tools for editing graphics. It supports a point-and-click editing mode to modify the plots.

TABLE A.1 Some Useful MATLAB Syntaxes and Examples

Input	Comment
g = [1 2 3]	g is a row (1×3) vector
p = [1 2 3]'	p is a column (3×1) vector
G = [1 2 3; 4 5 6; 7 8 9]	G is a 3 × 3 matrix
g*g'	Performs an inner-product operation
g.*g	Performs element-by-element multiplication
g+g	Performs element-by-element addition
G*g	Performs matrix-vector multiplication
r = 0:2:10	r is a six-element vector = [0 2 4 6 8 10] is used to separate the start and end
G(1,:)	Selects the first row of G
G(:,1)	Selects the first column of G
G(1:2,1:2)	Selects the upper 2 × 2 submatrix of G
G^2	Performs G ² (square of the matrix)
G.^2	Performs an element-by-element square
[g, g]	Concatenates the row vector
[g; g]	Places the next vector in the next row
ones (1,3)	Creates a 1×3 vector of all elements = 1
zeros (2,3)	Creates a 2×3 matrix of all elements = 0
save result	Saves all variables to file result.mat
save result1 x,y,z	Saves variables x, y, and z to the file result1.mat
load result.mat	Loads all variables from the file result.mat
clear all	Clears all of the workspace variables

TABLE A.2 MATLAB Commands and Examples for 2-D and 3-D Graphics

Input	Comment
t = 0:0.001:1; x = exp(-10.*t);	Sets up the time axis Creates a vector x
plot(t,x); plot(t,x,'ro')	Line plot of the x-axis (t) vs. the y-axis (x) Line plot using a red circle
<pre>figure stem(t,x) stem(t,x, 'filled')</pre>	Creates a new figure plot Discrete sequence plot terminated with an empty circle or full circle ('filled')
<pre>subplot(1,2,1); plot(t,x); subplot(1,2,2); plot(x,t); close all</pre>	Display window divided into one row and two columns Note: (x,y,z) = (number of rows, number of columns, index) Closes all of the figures
<pre>x = 0:1:50; y1 = sin(2*pi*x/8); y2 = cos(2*pi*x/8); plot (x,y1,x,y2) plot (x,y1, 'r'); hold on; plot (x,y2, 'g'); hold off;</pre>	Time index from 0 to 50, step = 1 Generates two sinewaves Plots two sinewaves in the same plot Plots two sinewaves, one at a time in the same plot
axis([0 10 -1 1]) grid on	Sets up the x-axis and y-axis limit Shows grid lines
<pre>xlabel('time') ylabel('amplitude') title('Two sine waves')</pre>	Specifies the name of the x-label Specifies the name of the y-label Specifies the title of the plot
<pre>t = 0:pi/50:5*pi plot3(sin(t), cos(t),t) axis square; grid on</pre>	Prepares the coordinates Performs a 3-D plot whose coordinates are elements of x, y, and z Changes to a square axis plot with a grid
<pre>[X,Y] = meshgrid ([-2:0.1:2]); Z = X.*exp(-X.^2 - Y.^2); mesh (X, Y, Z); surf (X, Y, Z); meshc (X, Y, Z); meshz (X, Y, Z); pcolor (X, Y, Z); surfl (X, Y, Z);</pre>	Transforms the vector into an array x, y Generates another array z Mesh surface plot Surface plot Surface plot with a contour beneath Surface plot with a curtain Pseudo-color plot 3-D-shaded surface with lighting

Graphical objects can be selected by clicking on the object, and multiple objects can be selected by shift-clicking. Several editing features are highlighted in Fig. A.2.

Besides selecting, moving, resizing, changing color, cutting, copying, and pasting graphic objects, MATLAB also provides tools for zoom-in, zoom-out, camera

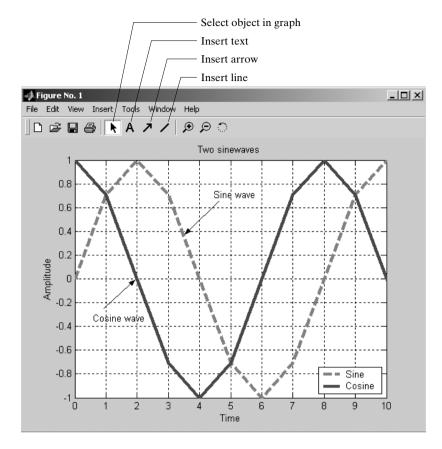


Figure A.2 Example of editing graphics

settings, basic data fitting, and displaying data statistics. These tools can be executed from the **Tools** pull-down menu. Once the diagram has been edited, the user can export the diagram into different graphical file types such as .bmp, .jpeg, etc., or save the diagram as a .fig file. A more detailed explanation can be found in the manual *Using MATLAB Graphics* [2].

A.1.4 Programming

We can enter and execute MATLAB commands directly in the command window. Since we may need to reuse these MATLAB commands and use several commands in sequence for a specific task, a better way is to use the MATLAB M-file editor (or another text editor) to write the MATLAB code (consisting of a sequence of commands) and save it into an M-file with an extension .m. The M-file editor can be activated by clicking on either the **New M-file** icon or the **Open File** icon

There are two types of M-file: scripts and functions. Comments that start with a percent sign (%) can be inserted anywhere inside a function or script file. A script

file has no input and output arguments and contains a sequence of MATLAB statements that make use of workspace data globally. To execute the M-file, simply type the name of the M-file in the command window and press **Enter** or **Return**. The user can interrupt a running program by pressing **Ctrl**–**C** or **Crtl**+**Break** at any time.

A function file must contain the word function in the first line of the program. For example,

```
function [mean, stdev] = stat(x)
% STAT - statistics of mean and standard deviation
n = length(x);
mean = sum(x)/n;
stdev = sqrt(sum((x-mean).^2)/n);
```

As shown in the example, a function can contain multiple output variables, which are enclosed in square brackets. Input arguments are enclosed in parentheses. All variables inside the function file are local and are not shared with the calling workspace.

Only the M-file function can be converted to a C (or C++) file. A script M-file cannot be compiled into a C file and thus needs to be converted to a function M-file first. Other M-files that cannot be compiled include an M-file that uses objects, an M-file that uses input and eval to manipulate a workspace, and MATLAB built-in functions. MATLAB supports C and C++ compilers such as (1) a C compiler (LCC C version 2.4), (2) Watcom C/C++ versions 10.6 & 11.0, (3) Borland C++ versions 5.0 onward, and (4) Microsoft Visual C++ versions 5.0 and 6.0. The compiler can be set up by issuing the following statement: mex -setup.

The compiler for building external interface files [MATLAB executable (MEX)] is chosen from the list of installed C compilers. The user can perform the following conversions:

• To convert an M-file to a C-file and create a C MEX-file, use

```
mcc -x <M-filename>
```

• To convert an M-file to a C-file and create a Simulink S-file, use

```
mcc -S <M-filename>
```

• To convert an M-file to a C-file and create a stand-alone C program, use

```
mcc -m <M-filename>
```

• To convert an M-file to a C-file and create a stand-alone C++ application, use

```
mcc -p <M-filename>
```

A.1.5 Data Types

It is important to understand the data types used in MATLAB since they affect the precision, the dynamic range, and quantization errors in the computation. By default, all MATLAB computations are carried out in double-precision, floating-point format

Data type	Description
single	Single precision, floating-point (32 bits)
double	Double precision, floating-point (64 bits)
int8, uint8	Signed, unsigned integer (8 bits)
int16, uint16	Signed, unsigned integer (16 bits)
int32, uint32	Signed, unsigned integer (32 bits)
char	Character array (or string)
cell	Cell array has elements that contain other arrays
structure	Structure array has field arrays and contains other arrays
user class	User-specified class

TABLE A.3 Data Types Used in MATLAB (adapted from [1])

(64 bits). However, data can be stored in single precision (32 bits) or integer (8 bits, 16 bits, and 32 bits) format to reduce memory requirements. Table A.3 lists some fundamental data types.

Using the data type as a prefix for numbers or variables can specify the precision. For example,

Note that we can find the number of bytes used in representing the array by looking at the workspace window. Besides representing numbers in the required format, we can also display the numeric format without converting the data by specifying the display format in File \rightarrow Preference \rightarrow Command Window \rightarrow Text display \rightarrow Numeric format.

MATLAB allows data to be grouped under a single variable known as the cell array. For example, we can group three different variables as an array x and specify the cell array using curly braces $\{\}$ as follows:

An alternate form for specifying the preceding cell array by name is to use the structure data type as follows:

```
X.num = [1 2 3]
X.char = 'hello'
X.matrix = eye(3)
```

A.1.6 Useful Commands

Many useful MATLAB functions are commonly used for matrix and linear-algebra operations, as summarized in Table A.4. MATLAB also provides several Fourier analysis commands for signal processing and data analysis. These functions are listed in Table A.5.

More signal-processing functions are discussed in the next section on MATLAB toolboxes.

TABLE A.4 MATLAB Functions for Matrix and Linear Algebra (adapted from [1])

Function	Comment
norm	Matrix or vector norm
rank	Matrix rank
det	Determinant of matrix
trace	Sum of diagonal elements
null	Null space
orth	Orthogonalization

TABLE A.5 Fourier Analysis Function (adapted from [1])

Function	Comment
fft	FFT
fft2	Two-dimensional FFT
fftn	N-dimensional FFT
ifft	IFFT
ifft2	Two-dimensional IFFT
ifftn	N-dimensional IFFT
abs	Magnitude
angle	Phase angle
unwrap	Unwraps a phase angle in radian
fftshift	Moves a zeroth lag to the center of the spectrum
cplxpair	Sorts numbers into complex-conjugate pairs
nextpow2	Next higher power-of-two number

A.2 USING DIGITAL SIGNAL PROCESSING TOOLBOXES AND INTERACTIVE TOOLS

As mentioned earlier, a MATLAB toolbox contains many application-specific M-files to solve a particular problem. Commonly used DSP toolboxes include the Signal Processing Toolbox [3], Filter Design Toolbox [4], Communication Toolbox [5], Image Processing Toolbox [6], Wavelet Toolbox [7], etc. This section summarizes the signal processing functions and tools in the Signal Processing Toolbox and Filter Design Toolbox. The reader can refer to the user guides for more details.

A.2.1 Signal Processing Toolbox

The Signal Processing Toolbox contains many functions that perform signal-processing algorithms, such as filter design and implementation, spectral analysis, windowing, statistical signal processing, transforms, multirate signal processing, waveform generation, and other operations. In addition to these DSP functions, the toolbox also contains two useful interactive tools: (1) the Signal Processing Tool, which provides interactive tools for analyzing and filtering signals, and (2) the Filter Design and Analysis Tool, which provides advanced filter-design tools for designing digital filters, quantizing filter coefficients, and analyzing quantization effects. These tools are explained in the following sections.

Table A.6 lists some important signal-processing functions in the Signal Processing Toolbox. This list is grouped under different categories. To learn about the detailed usage of these functions, simply type

```
help function name
```

to display the help menu for that particular function.

The signal-processing functions summarized in Table A.6, together with the powerful graphical tools, provide a comprehensive tool for signal processing. The Signal Processing Toolbox further provides an interactive tool that integrates the functions with the GUI, a topic which is discussed in the next section.

A.2.2 Signal Processing Tool

SPTool provides several tools for use in analyzing signals, designing and analyzing filters, filtering signals, and analyzing the spectrum of signals. The user can open this tool by typing

sptool

in the MATLAB command window. The **SPTool** main window appears, as shown in Fig. A.3.

Four windows can be accessed within SPTool:

The Signal Browser is used to view input signals. Signals from the workspace
or file can be loaded into SPTool by clicking on File → Import. An Import to SPTool window appears, which allows the user to select data from the file or

 TABLE A.6
 Signal-Processing Functions (adapted from [3])

Filter analysis	Description
filternorm freqs freqspace freqz freqzplot grpdelay impz unwrap zplane	Two-norm or inf-norm of a digital filter Frequency response of an analog filter Frequency spacing for a frequency response Frequency response of a digital filter Plots frequency-response data Group delay of a filter Impulse response of a digital filter Unwraps a phase angle Zero-pole plot
Filter implementation	Description
conv conv2 deconv fftfilt filter filter2 filtfilt filtic latcfilt medfilt1 sgolayfilt sosfilt upfirdn	Convolution and polynomial multiplication 2-D convolution Deconvolution and polynomial division FFT-based FIR filtering using overlap-add FIR or IIR filtering 2-D digital filtering Zero-phase digital filtering Initial condition of a transposed form-II IIR filter Lattice and lattice-ladder filtering 1-D median filtering Savitzky-Golay filtering Second-order (biquad) IIR filtering Upsample, FIR filtering, and downsample
FIR filter design	Description
convmtx cremez fir1 fir2 fircls fircls1 firls firrcos intfilt kaiserord remez remezord sgolay	Convolution matrix Complex and nonlinear-phase equiripple FIR Window-based FIR-filter design Frequency sampling-based FIR-filter design Constrained least-square FIR-filter design (multiband), (lowpass and highpass) Least-square linear-phase FIR-filter design Raised-cosine FIR-filter design Interpolation FIR-filter design FIR-filter design using a Kaiser window Computes a Parks–McClellan optimal FIR-filter design and filter-order estimation Savitzky–Golay filter design
IIR filter design	Description
bilinear butter buttord	Bilinear transformation Butterworth filter design Order and cutoff frequency for a Butterworth filter

(Continued)

TABLE A.6 (Continued)

IIR filter design	Description
cheby1 cheblord cheby2 cheb2ord ellip ellipord impinvar maxflat prony stmcb yulewalk	Chebyshev Type I filter design Order for a Chebyshev Type I filter Chebyshev Type II filter design Order for a Chebyshev Type II filter Elliptic filter design Minimum order for an elliptic filter Impulse-invariance method Generalized-digital Butterworth filter design Prony's method for IIR-filter design Linear model using a Steiglitz-McBride iteration Recursive digital-filter design (least-square)
Linear system transform	Description
latc2tf polystab polyscale residuez sos2ss sos2tf sos2zp ss2sos ss2tf ss2zp tf2latc tf2sos tf2ss tf2zp zp2sos zp2ss zp2tf	Converts lattice parameters to a transfer function Stabilizes a polynomial Scales the roots of a polynomial z-transform partial-fraction expansion Converts a second-order section to a state-space form Converts a second-order section to a transfer function Converts a second-order section to a zero-pole-gain form Convert state-space parameters to a second-order form Converts a state-space to a transfer function Convert state-space parameters to a zero-pole-gain Converts a transfer function to a lattice-filter form Converts a transfer function to a second-order section Converts a transfer function to a state-space Converts a transfer function to a zero-pole-gain Converts a zero-pole-gain to a second-order form Converts a zero-pole-gain to a state-space form Converts a zero-pole-gain to a transfer function
Windows	Description
bartlett barthannwin blackman blackmanharris bohmanwin boxcar chebwin gausswin hamming hann Kaiser	Bartlett window Modified Bartlett–Hanning window Blackman window Minimum four-term Blackman–Harris window Bohman window Rectangular window Chebyshev window Gaussian window Hamming window Hann (Hanning) window Kaiser window
nuttallwin	Nuttall-defined four-term Blackman–Harris

(Continued)

TABLE A.6 (Continued)

Windows	Description
triang	Triangular window
tukeywin	Tukey window
window	Window-function gateway
Transforms	Description
bitrevorder	Permutes input into a bit-reversed order
czt	Chirp z-transform
dct	Discrete cosine transform
dftmtx	DFT matrix
fft	1-D FFT
fft2	2-D FFT
fftshift	Moves zero-th lag to the center of the spectrum
goertzel	Second-order Goertzel algorithm
hilbert	Computes an analytic signal using a Hilbert transform
idct	Inverse-discrete cosine transform
ifft	1-D IFFT
ifft2	2-D IFFT
Cepstral analysis	Description
cceps	Complex cepstral analysis
icceps	Inverse-complex cepstrum
rceps	Real cepstrum, minimum phase reconstruction
0	
Statistical and	
Statistical and spectrum analysis	Description
	Description Estimates the magnitude-square coherence function
spectrum analysis	-
spectrum analysis	Estimates the magnitude-square coherence function
spectrum analysis cohere corrcoef	Estimates the magnitude-square coherence function Correlation coefficients
spectrum analysis cohere corrcoef corrmtx	Estimates the magnitude-square coherence function Correlation coefficients Autocorrelation matrix Covariance matrix Cross-spectral density
spectrum analysis cohere corrcoef corrmtx cov	Estimates the magnitude-square coherence function Correlation coefficients Autocorrelation matrix Covariance matrix Cross-spectral density Power-spectrum density (PSD) estimate using Burg's method
cohere corrcoef corrmtx cov	Estimates the magnitude-square coherence function Correlation coefficients Autocorrelation matrix Covariance matrix Cross-spectral density Power-spectrum density (PSD) estimate using Burg's method PSD estimate using a covariance method
cohere corrcoef corrmtx cov csd pburg pcov peig	Estimates the magnitude-square coherence function Correlation coefficients Autocorrelation matrix Covariance matrix Cross-spectral density Power-spectrum density (PSD) estimate using Burg's method PSD estimate using a covariance method PSD estimate using an eigenvector method
cohere corrcoef corrmtx cov csd pburg pcov	Estimates the magnitude-square coherence function Correlation coefficients Autocorrelation matrix Covariance matrix Cross-spectral density Power-spectrum density (PSD) estimate using Burg's method PSD estimate using a covariance method PSD estimate using an eigenvector method PSD estimate using a periodogram method
cohere corrcoef corrmtx cov csd pburg pcov peig	Estimates the magnitude-square coherence function Correlation coefficients Autocorrelation matrix Covariance matrix Cross-spectral density Power-spectrum density (PSD) estimate using Burg's method PSD estimate using a covariance method PSD estimate using an eigenvector method PSD estimate using a periodogram method PSD estimate using a modified covariance method
cohere corrcoef corrmtx cov csd pburg pcov peig periodogram	Estimates the magnitude-square coherence function Correlation coefficients Autocorrelation matrix Covariance matrix Cross-spectral density Power-spectrum density (PSD) estimate using Burg's method PSD estimate using a covariance method PSD estimate using an eigenvector method PSD estimate using a periodogram method PSD estimate using a modified covariance method PSD estimate using a Thomson multitaper method
spectrum analysis cohere corrcoef corrmtx cov csd pburg pcov peig periodogram pmcov pmtm pmusic	Estimates the magnitude-square coherence function Correlation coefficients Autocorrelation matrix Covariance matrix Cross-spectral density Power-spectrum density (PSD) estimate using Burg's method PSD estimate using a covariance method PSD estimate using an eigenvector method PSD estimate using a periodogram method PSD estimate using a modified covariance method PSD estimate using a Thomson multitaper method PSD estimate using the MUSIC method
spectrum analysis cohere corrcoef corrmtx cov csd pburg pcov peig periodogram pmcov pmtm pmusic psdplot	Estimates the magnitude-square coherence function Correlation coefficients Autocorrelation matrix Covariance matrix Cross-spectral density Power-spectrum density (PSD) estimate using Burg's method PSD estimate using a covariance method PSD estimate using an eigenvector method PSD estimate using a periodogram method PSD estimate using a modified covariance method PSD estimate using a Thomson multitaper method PSD estimate using the MUSIC method PIots PSD data
cohere corrcoef corrmtx cov csd pburg pcov peig periodogram pmcov pmtm pmusic psdplot pwelch	Estimates the magnitude-square coherence function Correlation coefficients Autocorrelation matrix Covariance matrix Cross-spectral density Power-spectrum density (PSD) estimate using Burg's method PSD estimate using a covariance method PSD estimate using an eigenvector method PSD estimate using a periodogram method PSD estimate using a modified covariance method PSD estimate using a Thomson multitaper method PSD estimate using the MUSIC method Plots PSD data PSD estimate using Welch's method
cohere corrcoef corrmtx cov csd pburg pcov peig periodogram pmcov pmtm pmusic psdplot pwelch pyulear	Estimates the magnitude-square coherence function Correlation coefficients Autocorrelation matrix Covariance matrix Cross-spectral density Power-spectrum density (PSD) estimate using Burg's method PSD estimate using a covariance method PSD estimate using an eigenvector method PSD estimate using a periodogram method PSD estimate using a modified covariance method PSD estimate using a Thomson multitaper method PSD estimate using the MUSIC method PSD estimate using Welch's method PSD estimate using Welch's method PSD estimate using the Yule–Walker autoregressive method
cohere corrcoef corrmtx cov csd pburg pcov peig periodogram pmcov pmtm pmusic psdplot pwelch pyulear rooteig	Estimates the magnitude-square coherence function Correlation coefficients Autocorrelation matrix Covariance matrix Cross-spectral density Power-spectrum density (PSD) estimate using Burg's method PSD estimate using a covariance method PSD estimate using an eigenvector method PSD estimate using a periodogram method PSD estimate using a modified covariance method PSD estimate using a Thomson multitaper method PSD estimate using the MUSIC method PSD estimate using Welch's method PSD estimate using Welch's method PSD estimate using the Yule—Walker autoregressive method Frequency and power estimation using the eigenvector algorithm
cohere corrcoef corrmtx cov csd pburg pcov peig periodogram pmcov pmtm pmusic psdplot pwelch pyulear rooteig rootmusic	Estimates the magnitude-square coherence function Correlation coefficients Autocorrelation matrix Covariance matrix Cross-spectral density Power-spectrum density (PSD) estimate using Burg's method PSD estimate using a covariance method PSD estimate using an eigenvector method PSD estimate using a periodogram method PSD estimate using a modified covariance method PSD estimate using a Thomson multitaper method PSD estimate using the MUSIC method Plots PSD data PSD estimate using Welch's method PSD estimate using the Yule—Walker autoregressive method Frequency and power estimation using the eigenvector algorithm Frequency and power estimation using the MUSIC algorithm
cohere corrcoef corrmtx cov csd pburg pcov peig periodogram pmcov pmtm pmusic psdplot pwelch pyulear rooteig rootmusic tfe	Estimates the magnitude-square coherence function Correlation coefficients Autocorrelation matrix Covariance matrix Cross-spectral density Power-spectrum density (PSD) estimate using Burg's method PSD estimate using a covariance method PSD estimate using an eigenvector method PSD estimate using a periodogram method PSD estimate using a modified covariance method PSD estimate using a Thomson multitaper method PSD estimate using the MUSIC method Plots PSD data PSD estimate using Welch's method PSD estimate using the Yule–Walker autoregressive method Frequency and power estimation using the eigenvector algorithm Frequency and power estimation using the MUSIC algorithm Transfer function estimate
cohere corrcoef corrmtx cov csd pburg pcov peig periodogram pmcov pmtm pmusic psdplot pwelch pyulear rooteig rootmusic tfe xcorr	Estimates the magnitude-square coherence function Correlation coefficients Autocorrelation matrix Covariance matrix Cross-spectral density Power-spectrum density (PSD) estimate using Burg's method PSD estimate using a covariance method PSD estimate using an eigenvector method PSD estimate using a periodogram method PSD estimate using a modified covariance method PSD estimate using a Thomson multitaper method PSD estimate using the MUSIC method Plots PSD data PSD estimate using Welch's method PSD estimate using the Yule–Walker autoregressive method Frequency and power estimation using the eigenvector algorithm Frequency and power estimation using the MUSIC algorithm Transfer function estimate Crosscorrelation function
cohere corrcoef corrmtx cov csd pburg pcov peig periodogram pmcov pmtm pmusic psdplot pwelch pyulear rooteig rootmusic tfe	Estimates the magnitude-square coherence function Correlation coefficients Autocorrelation matrix Covariance matrix Cross-spectral density Power-spectrum density (PSD) estimate using Burg's method PSD estimate using a covariance method PSD estimate using an eigenvector method PSD estimate using a periodogram method PSD estimate using a modified covariance method PSD estimate using a Thomson multitaper method PSD estimate using the MUSIC method Plots PSD data PSD estimate using Welch's method PSD estimate using the Yule–Walker autoregressive method Frequency and power estimation using the eigenvector algorithm Frequency and power estimation using the MUSIC algorithm Transfer function estimate

(Continued)

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TABLE A.6 (Continued)

Linear prediction	Description
ac2rc ac2poly is2rc lar2rc levinson lpc lsf2poly poly2ac poly21sf poly2rc rc2ac rc2is rc2lar rc2poly rlevinson schurrc	Autocorrelation sequence to reflection coefficients Autocorrelation sequence to a prediction polynomial Inverse sine parameters to reflection coefficients Log area ratios to reflection coefficients conversion Levinson–Durbin recursion Linear-predictive coefficients using autocorrelation Line-spectral frequencies to prediction polynomial Prediction polynomial to an autocorrelation sequence Prediction polynomial to line-spectral frequencies Prediction polynomial to reflection coefficients Reflection coefficients to an autocorrelation sequence Reflection coefficients to inverse-sine parameters Reflection coefficients to log-area ratios Reflection coefficients to a prediction polynomial Reverse Levinson–Durbin recursion Schur algorithm
Multirate signal processing	Description
decimate downsample interp interp1 resample spline upfirdn upsample	Resamples at a lower sampling rate (decimation) Downsamples an input signal Resamples data at a higher sample rate (interpolation) General 1-D interpolation Changes the sampling rate by any rational factor Cubic spline interpolation Upsample, FIR filtering, down sample Upsample input signal
Waveform generation	Description
chirp diric gauspuls gmonopuls pulstran rectpuls sawtooth sinc square tripuls vco	Swept-frequency cosine Dirichlet or periodic-sinc function Gaussian-modulated sinusoidal pulse Gaussian monopulse Pulse train Sampled aperiodic rectangle Sawtooth (triangle) wave Since function Square wave Sampled aperiodic triangle Voltage-controlled oscillator

workspace. To view the signal, simply highlight it and click on **View**. The **Signal Browser** window, shown in Fig. A.4., allows the user to zoom-in and zoom-out from the signal, read the data values via markers, display the format, and even play the selected signal using the computer's speakers.

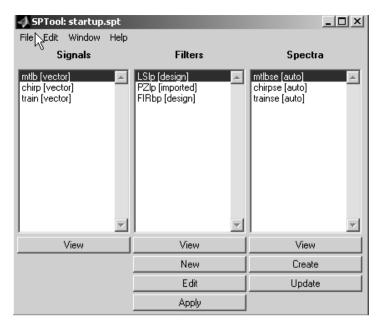


Figure A.3 SPTool window

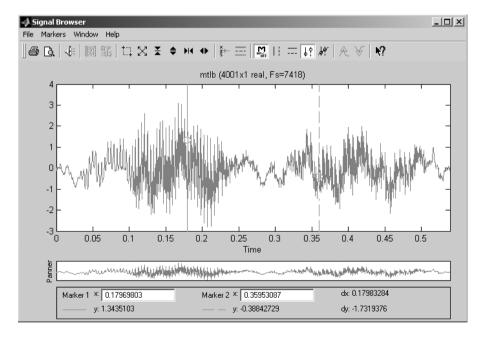


Figure A.4 Signal Browser window

2. The Filter Designer is used for designing digital FIR and IIR filters. The user simply clicks on the New icon for a new filter or the Edit icon for an existing filter under the Filter column in SPTool to open the Filter Designer window, as

shown in Fig. A.5. The user can design lowpass, highpass, bandpass, and bandstop filters using different filter-design algorithms. In addition, the user can also design a filter using the **Pole/Zero Editor** to graphically place poles and zeros in the z-plane. A useful feature is the ability to overlay the input spectrum onto the frequency response of the filter by clicking on the **Frequency Magnitude/Phase** icon

3. Once the filter has been designed, frequency specification and other filter characteristics can be verified by using the Filter Viewer. Select the name of the designed filter and click on the View icon under the Filter column in SPTool to open the Filter Viewer window, as shown in Fig. A.6. The user can analyze the filter in terms of its magnitude response, phase response, group delay, zeropole plot, impulse response, and step response.

When the filter characteristics have been confirmed, the user can then select the input signal and the designed filter. Click on the **Apply** button to perform filtering and generate the output signal. The **Apply Filter** window appears, as shown in Fig. A.7, and allows the user to specify the file name of the output signal. The **Algorithm** list provides a choice of several filter structures.

4. The final GUI window is the **Spectrum Viewer**, as shown in Fig. A.8. The user can view existing spectra by clicking on file names and then on the **View** button. Select the signal and click on the **Create** button to view the **Spectrum Viewer** window. The user can select one of the many spectral-estimation methods, such as Burg, covariance, FFT, modified covariance, MUSIC, Welch, Yule-Walker autoregressive (AR), etc., to implement the spectrum estimation. In

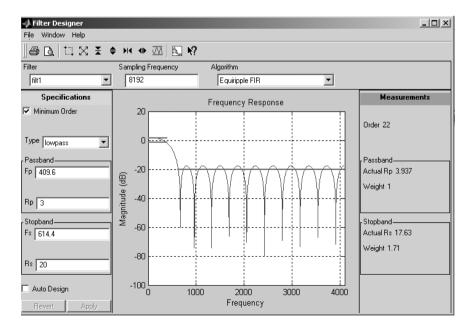


Figure A.5 Filter Designer window

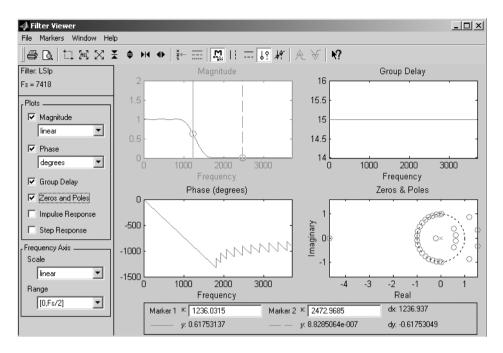


Figure A.6 Filter Viewer window

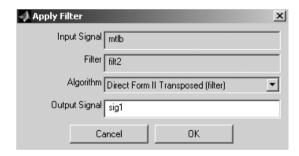


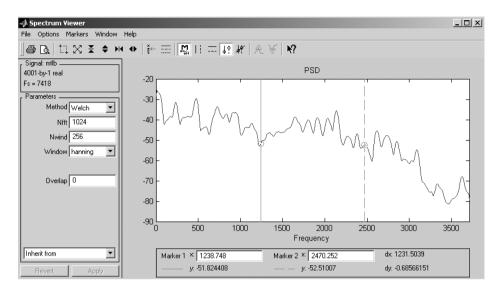
Figure A.7 Apply Filter window

addition, the size of the FFT, window functions, and overlapping samples can be selected to complete the power-spectrum density (PSD) estimation.

SPTool also provides a useful tool for exporting signals, filter parameters, and spectra to the MATLAB workspace or files. These saved parameters are represented in MATLAB as a structure of signals, filters, and spectra. More information can be found in the *Signal Processing Toolbox User's Guide* [3]. A step-by-step example of using SPTool in designing an IIR filter is given in Chapter 2.

A.2.3 Filter Design Toolbox

The Filter Design Toolbox is a collection of tools that provides advanced techniques for designing, simulating, and analyzing digital filters. It extends the capabilities of



Appendix A

Figure A.8 Spectrum Viewer window

the Signal Processing Toolbox by adding filter structures and design methods for complex real-time DSP applications. This toolbox also provides functions that simplify the design of fixed-point filters and the analysis of quantization effects.

The toolbox provides the following key features for digital-filter designs:

- **1.** Advanced FIR filter-design methods: The advanced equiripple FIR design automatically determines the minimum filter order required. It also provides constrained-ripple, minimum-phase, extra-ripple, and maximal-ripple designs. In addition, the least *P*-th norm FIR design allows the user to adjust the tradeoff between minimum-stopband energy and minimum order equiripple characteristics.
- **2.** Advanced IIR filter-design methods: Allpass IIR filter design with arbitrary group delay enables the equalization of nonlinear group delays of other IIR filters to obtain an overall approximate linear-phase passband response. Least-Pth-norm IIR design creates optimal IIR filters with arbitrary magnitude, and constrained least-P-th-norm IIR design constrains the maximum radius of the filter poles to improve the robustness of the quantization.
- **3.** *Quantization*: The toolbox provides quantization functions for signals, filters, and FFTs. It also supports quantization of filter coefficients, including coefficients created using the Signal Processing Toolbox.
- **4.** Analysis of quantized filters: The toolbox supports analysis of the frequency response, zero-pole plot, impulse response, group delay, step response, and phase response of quantized filters. In addition, it supports limit-cycle analysis for fixed-point IIR filters.

It is important to emphasize that the Filter Design Toolbox supports designing, simulating, and analyzing fixed-point filters for a wide precision range. It also

allows the user to compute quantized FFTs and IFFTs. These functions ease the task of determining the effects of quantization on real-world designs. Quantization tools allow the user to model the behavior of fixed-point filters and FFT algorithms precisely. The quantized algorithms in the toolbox exactly match the output of the algorithms implemented on a fixed-point processor because the simulation is bit-true.

The Filter Design Toolbox includes a new GUI tool called FDATool. This tool allows the user to design optimal FIR and IIR filters from scratch, import previously designed filters, quantize floating-point filters, and analyze quantization effects. This tool is introduced in the next section.

A.2.4 Filter Design and Analysis Tool

This interactive tool provides several advanced techniques that support designing, simulating, and analyzing fixed-point and floating-point filters for a wide range of precision. This tool performs the following functions:

- 1. Designs filters
- 2. Converts filters between different structures
- **3.** Ouantizes filters
- 4. Quantizes data
- 5. Quantizes FFT and IFFT
- 6. Designs adaptive filters

In addition to the existing functions listed in Table A.6, FDATool contains the additional filter-design functions listed in Table A.7.

FDATool can be activated by typing

fdatool

in the MATLAB command window. The **Filter Design & Analysis Tool** window is shown in Fig. A.9. This window includes tools similar to those shown in the **Filter**

TABLE A.7 Additional Filter Design Functions in FDATool

Filter design function	Description
firlpnorm	Designs minimax FIR filters using the
	least-Pth algorithm
gremez	Designs optimal FIR filters (Remez
	exchange)
iirgrpdelay	Designs IIR filters (specifies group delay
	in the passband)
iirlpnorm	Designs minimax IIR filters using the
	least-Pth algorithm
iirlpnormc	Designs minimax IIR filters using the
	least-Pth algorithm, which restricts filter
	poles and zeros within a fixed radius

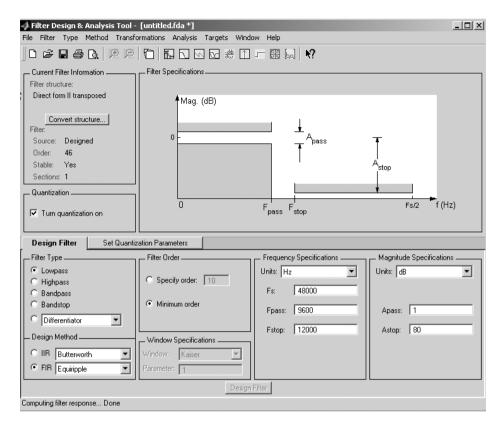


Figure A.9 Filter Design & Analysis Tool window

Designer window in SPTool (Fig. A.5). The design steps and available features that can be used to view filter characteristics are also similar. However, FDATool is a more advanced filter-design tool that includes additional filter types such as differentiator, Hilbert transform, multiband, arbitrary magnitude and group delay, Nyquist, and raised-cosine. FDATool also has an additional option that allows the default filter structure (direct-form II transposed) to be converted to different structures, such as direct-form I, direct-form II, direct-form I transposed, state-space, and its lattice equivalents, as shown in Fig. A.10.

In addition, FDATool is a very powerful tool for investigating quantized filters. Once a filter has been designed and verified, we can turn on the quantization mode by clicking on the **Set Quantization Parameters** icon located at the bottom-left side of window and then check the box **Turn quantization on**. The bottom panel of the **Filter Design & Analysis Tool** window changes, as shown in Fig. A.11. We can select (1) **Mode**, (2) **Round mode**, (3) **Overflow mode**, and (4) **Format** for representing coefficients, input, output, multiplicands, products, and sums of the filter. The options and explanations for setting these quantizer properties are listed in Table A.8. This new panel also provides a useful option for limiting filter coefficients to less than 1 by scaling the input of the filter, which can be accomplished by clicking on the

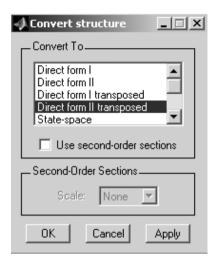


Figure A.10 Convert Structure window

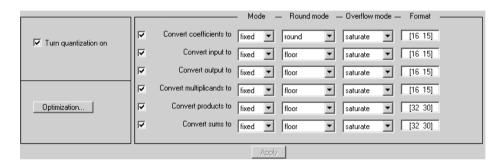


Figure A.11 Setting quantization parameters in FDATool

Optimization button (the meanings of these options are explained in Chapter 7). Furthermore, the direct-form IIR filter can also be converted to a cascade of second-order sections for a more stable implementation. This option can be activated by clicking on **Edit** → **Convert to Second-Order** sections. This option is only applicable to IIR filters and is explained in more detail in Chapter 7.

In addition to these features, the new version of FDATool also provides an option for frequency transformation, which can be accessed by clicking on the **Transform Filter** icon . This option transforms the original designed filter to another filter with different characteristics. It also contains a **Filter Realization Wizard** (Realize Model) that allows the user to create a fixed-point or floating-point filter blockset, which can be used in the Simulink environment. Furthermore, filters can also be imported from variables or discrete-time filter objects in the workspace by clicking on the **Import Filter** icon .

Besides using FDATool in specifying the quantized filter, we can also construct a quantized filter object using the function <code>qfilt</code> in MATLAB. For example,

```
Hq=qfilt('quantizer',{'float',[32 8],'round'})
```

TABLE A.8 Quantizer Properties

Quantizer property	Description
Mode:	
fixed	Specifies fixed-point arithmetic (default)
ufixed	Unsigned fixed-point calculations
float	Specifies floating-point arithmetic
double	Specifies double-precision, floating-point arithmetic
single	Specifies single-precision, floating-point arithmetic
Round mode:	
ceil	Rounds a value to the nearest integer towards $+\infty$
convergent	As in ceil (If tie, round down if the next-to-last bit is even; up if odd)
fix	Rounds a value to the nearest integer toward 0
floor	Rounds a value to the nearest integer toward $-\infty$
round	Rounds a value to the nearest integer (default): rounds a negative number toward $-\infty$, a positive number toward $+\infty$, and ties toward $+\infty$
Overflow mode:	
saturate	Sets the overflowed values to the maximum or minimum values (default)
wrap	Maps overflow values to the number range using modular arithmetic
Format:	
[wl,fl] (fixed) [wl,exp] (float) [32,8] (single) [64,11] (double)	Default value is [16 15] for Q.15 Maximum wordlength wl = 53 bits, fl: fractional length exp up to 11 bits; 64-bit >wl>exp, exp: exponent length IEEE-754 single precision (exp = 8, fl = 23, sign = 1) IEEE-754 double precision (exp = 11, fl = 52, sign = 1)

creates a quantized-filter object that is quantized to single-precision, floating-point format and that uses round mode for rounding coefficients and arithmetic results. The object is defined as a structure data type, and direct referencing can modify its properties after the object is constructed. For example, to change the value of the scaling factor at the input of the filter to 0.5, we can use the following command:

```
Hg.scalevalues = 0.5
```

FDATool also allows the user to construct the other quantized objects listed in Table A.9. A detailed explanation can be found in the *Filter Design Toolbox User's Guide* [4].

A useful report showing the minimum, maximum, number of overflows, number of underflows, and number of operations of the most recent application of F (quantized filter object, quantized FFT object, or quantizer object) can be produced by issuing the following command:

greport (F)

Quantized function	Description
qfft	Constructor for quantized FFT objects
qfilt	Constructor for quantized filter objects
quantizer	Constructor for quantizer objects
unitquantizer	Constructor for unit-quantizer objects

TABLE A.9 Quantizer, Quantized Filter, and Quantized FFT

Finally, the latest FDATool also provides a set of adaptive-filter functions using different adaptive algorithms for updating the filter coefficients listed in Table 9.1. Examples of using adaptive-filter functions are given in Chapter 9.

A.3 USING SIMULINK

Simulink provides a useful interactive interface for use in designing dynamic systems. This section introduces some basic functional blocks and their operations. Detailed information can be found in the *Simulink: User's Guide* [8]. To start Simulink from the MATLAB environment, type the following command:

simulink

or click on the Simulink icon at the top of the MATLAB window. A **Simulink Library Browser** appears, which contains the main Simulink blocks and all of the available Simulink blocksets, as shown in Fig. A.12. In this section, we examine the main Simulink Block Library. The DSP Blockset [9] and the Fixed-Point Blockset [10] are explained in Section A.4.

Right-click on **Simulink**, shown in Fig. A.12, and click on the menu **Open the 'Simulink' Library** to open the Simulink **Library** window, as shown in Fig. A.13. The functional blocks (libraries) include the following:

- **1. Continuous** contains a set of blocks that perform numeric derivative, integration, delay, etc.
- **2. Discrete** contains a set of blocks that perform discrete-dynamic systems, zero-order hold, delay, etc.
- **3. Look-up Tables** contains a set of blocks that perform table lookups.
- **4.** User-Defined Functions contains a set of MATLAB functions, S-functions, etc.
- **5. Math operations** performs math functions such as scaling, dot product, product, trigonometric operations, and logical operations, etc.
- **6. Discontinuities** contains a set of blocks for quantizer, saturation, relay, switch, nonlinear functions, etc.
- 7. Signals Routing manipulates the signal flow, storage and reading of data, etc.
- **8. Signal Attributes** determines the data-type conversion line probing for data width and sample time, specify attribute of signal line, etc.

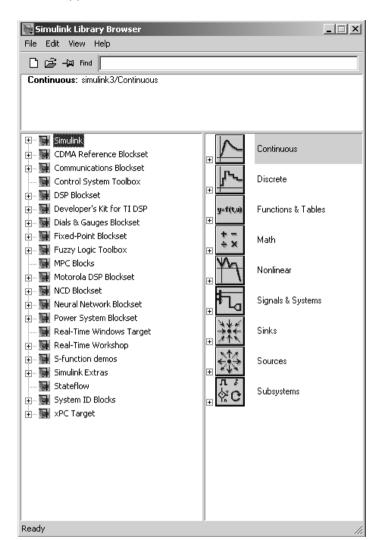


Figure A.12 Simulink Library Browser window

- **9. Sinks** contains a set of destination blocks such as scope, files, workspace, graphs, etc.
- **10. Sources** contains all source elements such as signal generator, random number, noise, clock, etc.
- **11. Port & Subsystems** contains useful blocks that allow external trigger, conditional switching, iterative system, subsystem clock, I/O port, etc.
- **12. Model Verification** checks the dynamic range of signals, input resolution, and static upper and lower bounds, etc.
- **13. Model-Wide Utilities** provides documentation of model and linearization of running model.

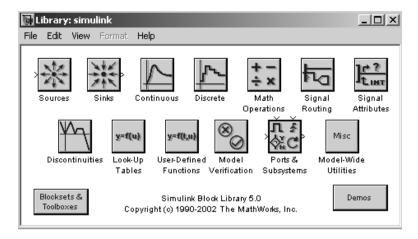


Figure A.13 Simulink Library window

To begin the block-building process, simply click on the **Create a new model** icon to open a new worksheet. In order to use these functional blocks, click on the selected block icon and drag it into the opened worksheet. Most blocks (except the sink and source blocks) have at least one input and one output node. We can connect two blocks by clicking on the output node of the first block and dragging the mouse to the input node of the second block. A quicker way of connecting two blocks is to click on the first (source) block and hold down the Ctrl key while also clicking on the second (destination) block. These two blocks are connected automatically.

In this section, we use a simple example to show the processes of designing and building a simple dynamic digital system in Simulink from scratch. The complete system is shown in Fig. A.14. The step-by-step procedure of building this complete design involves the following steps:

- Step 1: Click on File → New → Model in the Simulink Library Browser, or simply click on the Create a new model icon to open a new worksheet.
- **Step 2:** Select a sinusoidal source by double-clicking on **DSP Blockset**, and then click on **DSP Sources**. Select **Sine Wave** and drag it into the worksheet. Double-click on this block and set the following parameters: Amplitude = 1, Frequency = 1,000, and Sample time = 1/10,000.
- Step 3: Select a white-noise source by clicking on Random Source (in DSP Blockset → DSP Sources) and drag it into the worksheet. Double-click on this block, and set the following parameters: Source Type = Gaussian, Mean = 0, Variance = 1, and Sample time = 1/10,000.
- Step 4: Attach a gain block to the output of each source. The gain block is located in Simulink → Math Operations → Gain. Drag two gain blocks into the worksheet, set the gain for the sinusoidal source to 1, and set the gain for the noise source to 0.1. Link the output of the source to the input of the gain block by clicking on the output node of the

- source block. Then, complete the connection by dragging the mouse to the input node of the gain block, as shown in Fig. A.14.
- Step 5: Combine the two signal sources using a summing block. Click on Sum (in Simulink → Math Operations) to select the summing block, and drag it into the worksheet. Connect it to the outputs of the Gain blocks using the method described in Step 4.
- Step 6: Apply a digital bandpass filter to the output of the summing junction to enhance the sinewave and attenuate the noise. The digital filter is located at DSP Blockset. Click on Filtering, and double-click on Filter Designs. Drag Digital Filter Design (FDATool block) to the worksheet and double-click on the block to open a window similar to that shown in Fig. A.9. We can specify the following parameters for designing the bandpass filter: Filter Type = Bandpass, Design Method = FIR → Equiripple, Filter order = minimum order, Fs = 10,000, Fstop1 = 900, Fpass1 = 950, Fpass2 = 1,050, Fstop2 = 1,100, Astop1 = 40 dB, Apass = 1 dB, and Astop2 = 40 dB. After entering all of the parameters, click on the Design Filter button to start the filter design and implementation.
- Step 7: Examine the signals before and after the bandpass filter using a Spectrum Scope. This scope can be found in DSP Blockset → DSP Sinks. Drag it to the worksheet. We can also view the output from the summing junction (noisy sinewave) by using another Spectrum Scope. Drag the second Spectrum Scope into the worksheet, click on the input node, and drag the mouse to the line between the summing junction and the bandpass filter. Enable the buffer input option of the Spectrum Scope, and use the default FFT length. This configuration allows the user to compare the difference between the original signal and the filtered signal.
- Step 8: Set the Simulink parameters before running the simulation once the Simulink model has been completed, as shown in Fig. A.14. Click on Simulation and select Simulation parameters from the menu to open the window shown in Fig. A.15. Set the parameters as shown in Fig. A.15 for a 10-second simulation at a sampling frequency of 10,000 Hz. Close the window and start the simulation by clicking on the button . Two spectrum plots are displayed that show that the processed signal has a 40 dB reduction in noise power. To stop the simulation, simply click on the button .

We can save the simulation model shown in Fig. A.14 to a file bpf.mdl and use the file later in the MATLAB environment by typing

bpf

In the second example, we open a dynamic system (from Simulink demo [8]) by typing

combfilter

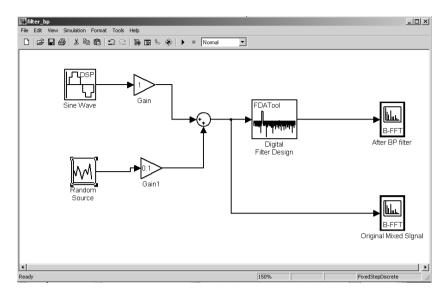


Figure A.14 A simple Simulink model for digital filtering

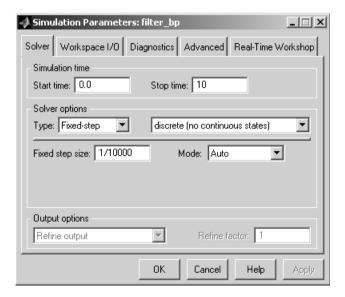


Figure A.15 Simulation parameters to run the filter simulation

This Simulink model is shown in Fig. A.16. This example illustrates the differences between digital filtering using single (upper path) and double (lower path) precisions. We can click on the block and drag it around. Double-click on any block to show either the internal block's detail or open up the block's parameters window. For example, double-click on the **Signal Waveform** block to open the **Block Parameters** menu, which allows the user to adjust the waveform type, amplitude, and frequency of the signal. The user can change the default 1 Hz sinewave into a 2 Hz squarewave with an amplitude of 50.

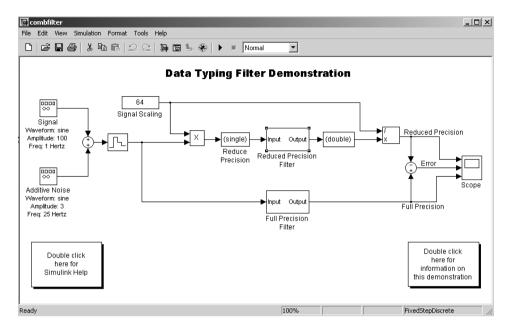


Figure A.16 A Simulink demo showing single- and double-precision filtering

Several blocks can be grouped together to form a subsystem. The subsystem can be created by selecting multiple blocks using Shift-click or by using the mouse cursor to form a bounding box that includes these blocks. We then select **Create subsystem** under the **Edit** menu to form a subsystem. The details of the subsystem can be viewed by double-clicking on it. For example, in the preceding Simulink demo, we can select the **Signal Scaling** block and the **Multiply** block to form a subsystem. Double-click on this new subsystem to view the internal details.

The subsystem can be masked to allow the user to customize the dialog box and icon for this subsystem. Click on the subsystem, and select Mask Subsystem under the Edit menu. A Mask Editor window is opened, as shown in Fig. A.17. Click on the Parameters pane, and then click on the Add icon to specify the attributes of mask parameters including prompt, variables, and control type. In this example, the parameters constant and c are specified in the Prompt and Variable fields, respectively. Make sure that the Type entry is selected as edit and that both Evaluate and Tunable are turned on. The block description is defined in the Documentation pane, as shown in Fig. A.18. The Documentation menu specifies the dialog box once the subsystem block is double-clicked.

Before starting Simulink, the simulation parameters must be specified by selecting **Simulation Parameters** under the **Simulation** menu. The **Simulation Parameters** window appears, as shown in Fig. A.19. The start and stop time for the simulation can be specified in the **Start time** (0 sec) and **Stop time** (4 sec) fields. Simulink provides a number of solvers for the simulations. In DSP, the solver option is set to **Fixed-step** (take the same step size during simulation) and **discrete** (no continuous states). The **Fixed step size** field can be set to **auto** or to the sampling period of the

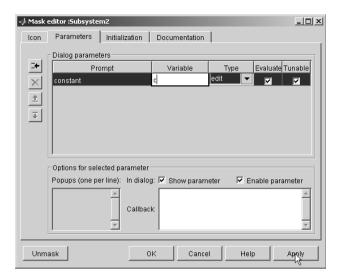


Figure A.17 Mask editor window

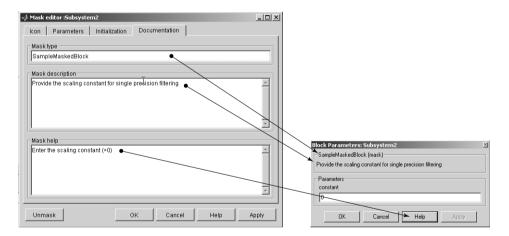


Figure A.18 Block documentation

DSP system. Finally, the mode is set to **SingleTasking** (for a single-tasking system with a single-sampling rate). Other available modes are **MultiTasking** (a multisampling rate is used in the system) and **Auto** (automatically adjust between rates). Click on **OK** and start the simulation by clicking on the play icon .

The **Scope** block displays the outputs from the reduced-precision and full-precision filters, and the error (middle plot) between these precisions is shown in Fig. A.20. The user can also double-click on the **Reduce Precision** block to change the **data type** to **int8**, **int16**, or **int32**, and observe the differences in comparison with the double-precision comb filter. More information on the usage of various blocks and settings can be found in the *Simulink: User's Guide* [8].

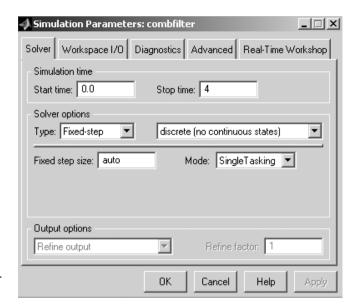


Figure A.19 Simulation Parameters window

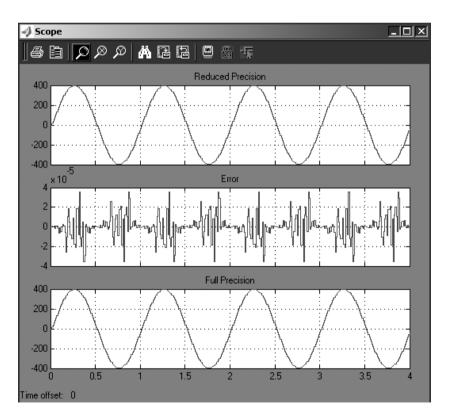


Figure A.20 Scope display

Additional Simulink tools can be used for simulating a dynamic system. These tools include:

- 1. Simulink Accelerator
- 2. Model Differencing tool
- 3. Profiler
- 4. Model Coverage tool

The **Simulink Accelerator** speeds up the execution of Simulink simulations. This tool uses part of the Real-Time Workshop to create and compile C code to replace the code that Simulink uses in normal mode. This mode can be selected by clicking on the **Simulink** menu and selecting **Accelerator**. Alternatively, the user can select **Accelerator** from the menu located in the middle of the toolbar

The **Model Differencing** tool is available under the **Tools** menu of Simulink. It finds and displays the differences between two Simulink models, and the differences are reported in the Model Differences Report.

To activate the profiler, simply click on **Profiler** under Simulink's **Tools** menu. The Simulink Profiler collects and profiles performance data while the model is being simulated. When the simulation finishes, Simulink generates a report that describes the time taken to execute each functional block. This feature allows the user to identify the blocks or subsystems that require more time for execution and thus are targets for optimization.

The **Model Coverage** tool validates the user's models. It analyzes the execution of blocks that serve as decision points in the model. These blocks include switch, triggered subsystem, enabled subsystem, absolute value, saturation, stateflow charts, etc. The user can run the coverage tool by selecting **Coverage Settings** from Simulink's **Tools** menu. When the **Coverage Settings** dialog box is displayed, simply check **Enable Coverage Reporting**. Note that both model-coverage reporting and acceleration mode cannot be enabled at the same time because Simulink disables coverage reporting if the accelerator option is enabled.

Several Simulink blocksets, such as DSP Blockset, Fixed-Point Blockset, Communication Blockset, Code Division Multiple Access (CDMA) Reference Blockset, and many others, are available to run in the Simulink environment. The following section describes two important blocksets that are related to DSP applications.

A.4 USING BLOCKSETS

A blockset is a set of special blocks that are included in an application library for handling the application tasks in Simulink. The most commonly used blocksets for DSP applications are DSP and Fixed-Point blocksets [9, 10].

A.4.1 DSP Blockset

The DSP Blockset is a collection of signal-processing blocks for use with Simulink. To access the DSP Blockset, type the following in the MATLAB command window:

dsplib

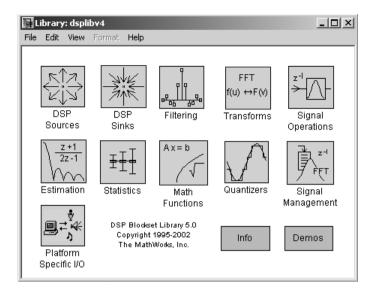


Figure A.21 Functions available in the DSP Blockset

Many important DSP algorithm blocks are included in the DSP Blockset Library. As shown in Fig. A.21, this library includes the following blocks:

- **1. DSP Sources** contains special sources for generating signals such as chirp, discrete impulse, sinewave, clock, etc. It also contains blocks that capture signals from workspaces, wave devices, and wave files.
- **2. DSP Sinks** contains functional blocks for counter, matrix, time, vector, and spectrum scope. It also contains a block for writing signals into workspaces, wave devices, and wave files.
- **3. Filtering** contains three main sections: (1) filter design (contains many FIR-and IIR-filter designs and structures), (2) multirate filters (contains decimation, interpolation, analysis and synthesis, and wavelet analysis and synthesis filters), and (3) adaptive filters (contains LMS, Kalman, and RLS adaptive filters).
- **4. Transforms** contains a set of transforms that include FFT, IFFT, discrete cosine transform (DCT), inverse discrete cosine transform (IDCT), and real and complex cepstrum.
- **5. Signal Operations** provides a set of commonly used DSP operations such as convolution, upsample, downsample, integer and fractional delay, zero padding, sample-and-hold, etc.
- **6. Estimation** contains three sections: (1) linear prediction (contains autocorrelation), (2) power-spectrum estimation (contains magnitude-square FFT, Burg, covariance, Yule-Walker, and STFT), and (3) parametric estimation (Burg, covariance, and Yule-Walker AR estimators).
- **7. Statistics** contains a set of blocks for computing autocorrelation, correlation, histogram, variance, median, mean, etc.

- **8. Math Functions** contains three functional groups: (1) math operations (dB gain, conversion, normalization, complex exponential, etc.), (2) matrix and linear algebra (matrix factorization, matrix inversion, linear-system solvers, pseudo inversion, matrix multiplication, matrix square, Toeplitz, etc.), and (3) polynomial-function evaluation.
- Quantizers contains several blocks for quantizer and uniform encoder and decoder.
- **10. Signal Management** contains four groups: (1) switches and counters, (2) buffer for buffering and unbuffering signals, (3) matrix and signal indexing, and (4) signal attributes that convert the dimension of the signal, check signal and frame status, etc.
- 11. Platform Specific I/O contains a set of blocks that take in sound signals from a standard audio device and output to a standard audio device in real time. It also contains a set of blocks that read from and write to a file with a standard .way format.

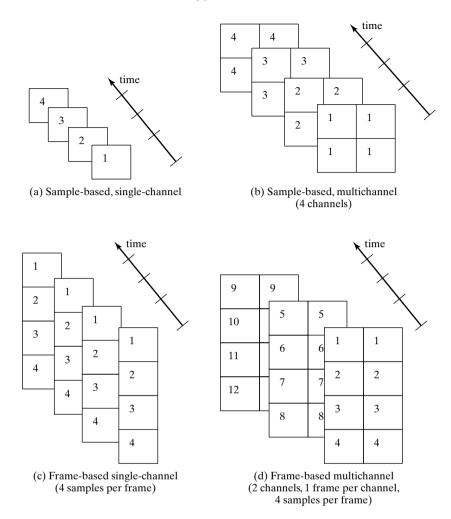
The DSP Blockset has the flexibility to handle single and multichannel signals, as will as sample and frame-based signals. Therefore, the following four possible types of input signal can be implemented in Simulink, as shown in Fig. A.22: (1) sample-based single-channel signals, (2) sample-based multichannel signals, (3) frame-based single channel signals and, (4) frame-based multichannel signals.

Frame-based (block) processing can accelerate real-time systems, as explained in Chapter 3. In real-time DSP systems, data acquisition can be carried out at a higher rate when a block of samples is transferred to the processor and when these block samples are processed at once. In contrast, sample processing interrupts the processor at every sample. Therefore, frame-based processing minimizes the overhead incurred in interrupting the processor. However, we have to consider the latency introduced by frame-based processing in some applications, as explained in Chapter 3.

In Simulink models, frame-based and sample-based signals are indicated by a double line (\Rightarrow) and single line (\rightarrow), respectively. A sample-based signal can be converted to a frame-based signal by using the **Buffer** block, and an **Unbuffer** block converts a frame-based signal back to a sample-based signal. The DSP Sources Library provides a set of blocks for creating sample-based and frame-based signals.

Two different delays affect Simulink models: (1) computational delay and (2) algorithmic delay. Computational delay depends on how fast the computer hardware and software can execute a block of data. There are several methods to reduce computational delay, such as using frame-based processing or using the Real-Time Workshop to generate generic real-time code for specific hardware. Algorithmic delay is intrinsic to the algorithm and is independent of the processor. Algorithmic delay is commonly implemented in Simulink using an integer-delay block. It can also occur in certain conditions known as tasking latency, which arises from the synchronization requirements of Simulink's tasking mode. For example, the multirate blocks operating in multitasking mode are subject to tasking latency.

More information on and examples of using the DSP Blockset can be found in the DSP Blockset User's Guide [9].



Appendix A

Figure A.22 Different types of signal processing available in Simulink

A.4.2 Fixed-Point Blockset

The Fixed-Point Blockset [10] simulates effects commonly encountered in fixed-point systems such as digital filtering. The Fixed-Point Blockset extends the functional blocksets of the standard Simulink. This blockset allows the user to develop, simulate, analyze, and implement DSP systems using fixed-point arithmetic. A fixed-point simulation is vital for resolving finite-wordlength problems associated with different algorithms before entering the coding and implementation stages of the development cycle. Such a simulation greatly enhances the accuracy of the code and speeds up system-development time. Some important features of this blockset include the following:

1. Integer, fractional, and generalized fixed-point data types with a wordlength from 1 to 128 bits

- **2.** IEEE-754 format with single and double precisions, as well as non-IEEE standards
- 3. Methods to control the overflow, scaling, and rounding of fixed-point data
- **4.** An interface tool to profile the statistics of the simulation
- **5.** Generation of C code (integer type only) for execution on a given embedded processor

Figure A.23 shows the window that contains the blocks in the Fixed-Point Blockset Library. This window can be opened by typing

fixpt

in the MATLAB command window. The window shows that the library contains the following different blocks: (1) Math, (2) Data Type Conversion & Propagation, (3) Look-Up tables, (4) Logic & Comparison, (5) Filters (for performing filtering), (6) Delays & Holds, (7) Select (for selecting one from many inputs), (8) Nonlinear (for performing different nonlinear threshold functions), (9) Calculus, (10) Bits, (11) Sources, (12) FixPt GUI, (13) Demos, and (14) Edge Detect.

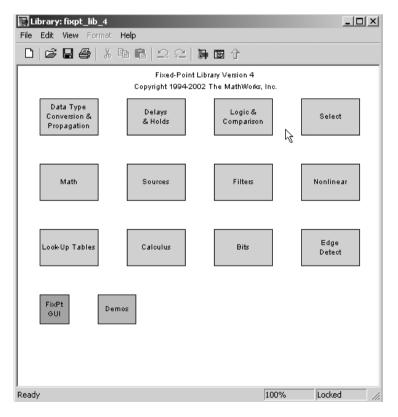


Figure A.23 Main window for the Fixed-Point Blockset Library

Configure Fixed-Point Blocks

The Fixed-Point Blockset provides an option for specifying the **output data type** via the block dialog box. Possible options are listed in Table A.10. In the signed and unsigned integer representation (sint, uint), the default binary point is to the right of the LSB. In the unsigned fractional-data type (ufrac), the default binary point is to the left of the MSB. In the signed fractional-data type (sfrac), the binary point is just to the right of the MSB (sign bit). In order to provide flexibility in defining different Q formats (or generalized fixed-point numbers) with a user-specified binary point, the Fixed-Point Blockset provides generalized signed and unsigned fixed-point numbers (sfix, ufix), and the binary point is determined from the value in the output scaling. For example, if the scaling factor is 2^E , the binary point is located E bits left of the LSB. The final data type is the floating-point data format, which can be represented by IEEE-754 single precision float('single'), IEEE-754 double precision float('double'), and a non-IEEE format expressed as float(w,e) for a total of w bits with e exponent bits, (w-e-1) fractional bits, and 1 sign bit.

The option of selecting output scaling is applicable to fixed-point data types. There are two general scaling modes: binary point-only scaling and slope/bias scaling. In binary point-only scaling mode, power-of-two scaling (e.g., 2^{-E} , where E is unrestricted) is used. This scaling has the effect of moving the binary point to the left by E bits. In slope/bias scaling mode, a slope of $S = F2^E$ is realized, where a bias of E bits is used to scale the integer number E as E bits is used to scale the integer number E as E bits is normalized to E specifies the binary point, and E is the fractional slope, which is normalized to E bits considered under slope/bias scaling with E and E bits considered under slope/bias scaling with E bits considered under slope/bias scaling and maximizes usage of the finite number of bits.

Fixed-point numbers can be rounded using the following rounding modes:

- 1. Zero, which rounds toward zero and is similar to the fix function in MATLAB.
- 2. Nearest, which rounds toward the nearest representable number and is similar to the round function.

TABLE A.10 Definition of	f Output Data Types
--------------------------	---------------------

Option	Description
uint(n)	n-bit unsigned integer
sint(n)	n-bit signed integer
ufrac(n,g)	n-bit unsigned fractional number
sfrac(n,g)	n-bit signed fractional number
	Note: the number of guard bits, g (optional), lies at the left
	of the default binary point
ufix(n)	n-bit unsigned fixed-point number
sfix(n)	n-bit signed fixed-point number
float('single')	IEEE-754 single-precision, floating-point number
float('double')	IEEE-754 double-precision, floating-point number
float(w,e)	Total bits w and exponent bits e

- **3.** Ceiling, which rounds toward positive infinity and is similar to the ceil function.
- **4.** Floor, which rounds toward negative infinity and is similar to the floor function.

Overflow handling in the Fixed-Point Blockset can be set by checking the **Saturate to max or min when overflow occurs** check box to specify the use of saturated mode. If unchecked, overflow wraps to the other end of the number range.

An example of an analog-to-digital model (shown in Fig. A.24) is used to illustrate the effects of using different wordlengths in representing digitized samples with the Fixed-Point Blockset. To start the simulation, type

fxpadc

in the MATLAB command line. The user can double-click on any block and adjust the default parameters. The **Signal Generator** block is configured to generate a sinewave with double-precision amplitude on the interval [-2, +2]. The **Zero-Order Hold** block simulates the sampling of the continuous sinewave, and the **Dbl To FixPt1** block converts the double-precision, floating-point number to a fixed-point representation. Once the desired parameters have been specified, the user can click on the play button \blacktriangleright to start the simulation.

In the example, different fixed-point implementations using (1) Q.15 sfrac(16), (2) Q1.14 sfix(16) with scaling of 2^{-14} , and (3) 8-bit wordlength sfix(8) with scaling of 2^{-4} can be implemented individually inside the **Dbl to FixPt1** block. The result for a Q.15 representation is shown in Fig. A.25(a), which is unable to cover the entire range of the input signal due to the fact that the dynamic range of a Q.15 number is [-1, +0.99]. By introducing an additional integer bit as in the Q1.14 format, the dynamic range increases to [-2, +1.99], but with reduced precision. The result is illustrated in Fig. A.25(b), which shows very little difference between the double-precision and Q1.14 representation. A coarser quantization

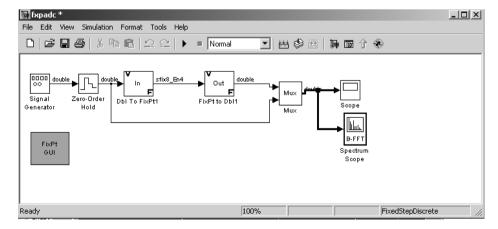
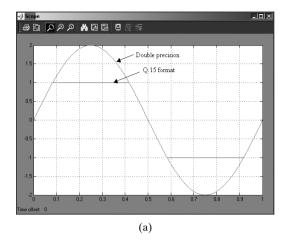
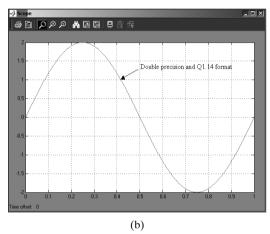


Figure A.24 Simulink block diagram for a fixed-point implementation of ADC





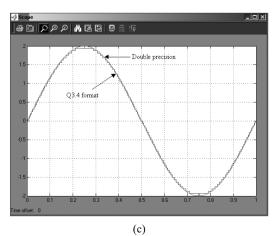


Figure A.25 Different data representations: (a) Q.15 and double precision, (b) Q1.14 and double precision, (c) Q3.4 and double precision

step is observed in Fig. A.25(c) for the 8-bit wordlength, where the binary point is the fourth bit to the left of the rightmost bit.

C-Code Generation in Fixed-Point Blockset

After the system has been designed and analyzed using the Fixed-Point Blockset, C code can be generated directly using the Real-Time Workshop [11]. The C code generated from the fixed-point block uses only integer types and can be used directly on embedded fixed-point processors, which can be either a fixed-point or floating-point architecture. The code can also be generated for testing on a rapid prototyping system such as xPC [12] and a real-time window target [13]. It can also generate code for non-real-time testing on a computer running on any supported operating system. Please refer to reference [10] for detailed information on the step-by-step generation of fixed-point C code.

A.5 MATLAB LINK FOR CODE COMPOSER STUDIO

MATLAB has introduced a new toolbox called MATLAB Link for CCS, which establishes bidirectional links between MATLAB, CCS, and TMS320 processors [14]. As shown in Fig. A.26, the MATLAB Link for CCS connects MATLAB to Texas Instruments software and hardware. It allows the MATLAB user to create an object that links to CCS and the real-time data exchange (RTDX) so that the user can transfer data to and from the processor without halting it. In other words, the user can open the CCS environment, download the program, communicate with the DSP processor, access the processor's registers and memories, and perform data logging while the DSP processor is running. This capability allows the user to change a parameter or variable in MATLAB and transfer the value into the running DSP processor in order to tune and alter the algorithm in real time.

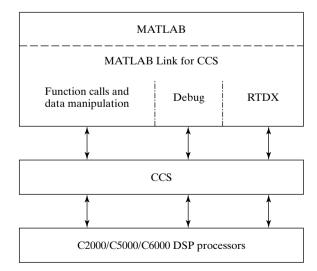


Figure A.26 Block diagram of MATLAB Link for CCS

The current version of MATLAB Link for CCS can link the TMS320C6701 Evaluation Module (EVM) TMS320C6711 DSK, C6000 and C5000 simulators, and other DSP boards that are supported in the setup utility. In this section, we give a brief overview of some important features of MATLAB Link for CCS and show how to link it to the C5000 simulator.

Appendix A

First, we can check whether the link for CCS is installed in MATLAB by typing

```
help ccslink
```

in the MATLAB command line. If the software is installed, it returns a list of commands for analysis and debugging with CCS. Next, we can check the boards and processors that are installed on the computer by typing

```
ccsboardinfo
```

which returns a list of boards and processors that are installed and recognized by CCS.

In order to select the C5000 simulator, type

```
[boardnum, procnum] = boardprocsel
```

and select the C5400 (or C5500) simulator. Note that if only a single board (or simulator) is installed in the system, this step can be skipped. After successful selection of the board or simulator, we can create the link between MATLAB and CCS by typing the following MATLAB command:

```
cc = ccsdsp('boardnum', boardnum, 'procnum', procnum);
```

We notice that CCS is placed in the background, and we can view the status by typing

```
disp(cc)
```

In addition, we can set the visibility for the CCS window by typing

```
visible(cc,1)
```

The next step is to load a project file (e.g., ccstut_54xx.pjt) into CCS using the following MATLAB commands:

However, we cannot build the CCS project in the MATLAB window. The user must build the project in CCS by clicking on the **Build-All** icon in the CCS window. After

the project is built, we can load the ccstut_54xx.out file using the command

After loading, we can set the breakpoint, reset the program counter, and run the program using the following commands:

Data located in memory on the target processor can be transferred to the MATLAB environment. For example, to read the data ddtav and idtav from CCS to the MATLAB workspace, type the following commands:

```
ddtav = read(cc,address(cc,'ddat'),'single',4)
idtav = read(cc,address(cc,'idat'),'int16',4)
```

MATLAB supports several data types. In the preceding example, the data type single is used instead of double, which is not supported by the C5000 simulator.

Besides reading from the target processor, the user can also write data to memory on the target processor. For example, we can modify ddtav and idtav as follows:

In addition, DSP registers can be viewed and modified. A simple example that changes the value of ACC A is shown as follows:

```
regal=cc.regread('AL','binary')
dec2hex(regal)
cc.regwrite('AL',hex2dec('683'),'binary')
dec2hex(cc.regread('AL','binary'))
```

More detailed information on MATLAB Link for CCS and how to use links for RTDX, which requires actual hardware (EVM or DSK) to be hooked up to the host computer, are documented in [14].

SUGGESTED READINGS

- 1 The MathWorks. *Using MATLAB*. Version 6, 2002.
- 2 The MathWorks. Using MATLAB Graphics. Version 6, 2002.
- 3 The MathWorks. Signal Processing Toolbox User's Guide. Version 6, 2002.

- 4 The MathWorks. Filter Design Toolbox User's Guide. Version 2, 2002.
- **5** The MathWorks. *Communication Toolbox User's Guide*. Version 2, 2002.
- 6 The MathWorks. Image Processing Toolbox User's Guide. Version 3, 2002.
- 7 The MathWorks. Wavelet Toolbox User's Guide. Version 2, 2002.
- 8 The MathWorks. Simulink: User's Guide. Version 5, 2002.
- **9** The MathWorks. DSP Blockset User's Guide: For Use with Simulink. Version 5, 2002.
- 10 The MathWorks. Fixed-Point Blockset User's Guide: For Use with Simulink. Version 4, 2002.
- 11 The MathWorks. Real-Time Workshop User's Guide: For Use with Simulink. Version 5, 2002.
- 12 The MathWorks. xPC Target User's Guide: For Use with Simulink. Version 2, 2002.
- **13** The MathWorks. *Real-Time Window Target User's Guide: For Use with Simulink*. Version 2.1, 2002.
- 14 The MathWorks. MATLAB Link for Code Composer Studio. Version 1, 2002