# **Python Control Library Documentation**

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The Python Control Systems Library (*python-control*) is a Python package that implements basic operations for analysis and design of feedback control systems.

### **Features**

- Linear input/output systems in state-space and frequency domain
- Block diagram algebra: serial, parallel, and feedback interconnections
- Time response: initial, step, impulse
- Frequency response: Bode and Nyquist plots
- Control analysis: stability, reachability, observability, stability margins
- Control design: eigenvalue placement, LQR, H2, Hinf
- Model reduction: balanced realizations, Hankel singular values
- Estimator design: linear quadratic estimator (Kalman filter)

### **Documentation**

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**CHAPTER** 

ONE

### INTRODUCTION

Welcome to the Python Control Systems Toolbox (python-control) User's Manual. This manual contains information on using the python-control package, including documentation for all functions in the package and examples illustrating their use.

### 1.1 Overview of the toolbox

The python-control package is a set of python classes and functions that implement common operations for the analysis and design of feedback control systems. The initial goal is to implement all of the functionality required to work through the examples in the textbook Feedback Systems by Astrom and Murray. A *MATLAB compatibility module* is available that provides many of the common functions corresponding to commands available in the MATLAB Control Systems Toolbox.

### 1.2 Some differences from MATLAB

The python-control package makes use of NumPy and SciPy. A list of general differences between NumPy and MATLAB can be found here.

In terms of the python-control package more specifically, here are some thing to keep in mind:

- You must include commas in vectors. So [1 2 3] must be [1, 2, 3].
- Functions that return multiple arguments use tuples.
- You cannot use braces for collections; use tuples instead.

### 1.3 Installation

The *python-control* package can be installed using pip, conda or the standard distutils/setuptools mechanisms. The package requires numpy and scipy, and the plotting routines require matplotlib. In addition, some routines require the slycot library in order to implement more advanced features (including some MIMO functionality).

To install using pip:

```
pip install slycot # optional
pip install control
```

Many parts of *python-control* will work without *slycot*, but some functionality is limited or absent, and installation of *slycot* is recommended.

*Note*: the *slycot* library only works on some platforms, mostly linux-based. Users should check to insure that slycot is installed correctly by running the command:

```
python -c "import slycot"
```

and verifying that no error message appears. It may be necessary to install *slycot* from source, which requires a working FORTRAN compiler and either the *lapack* or *openplas* library. More information on the slycot package can be obtained from the slycot project page.

For users with the Anaconda distribution of Python, the following commands can be used:

```
conda install numpy scipy matplotlib # if not yet installed conda install -c conda-forge control
```

This installs *slycot* and *python-control* from conda-forge, including the *openblas* package.

Alternatively, to use setuptools, first download the source and unpack it. To install in your home directory, use:

```
python setup.py install --user
```

or to install for all users (on Linux or Mac OS):

```
python setup.py build
sudo python setup.py install
```

## 1.4 Getting started

There are two different ways to use the package. For the default interface described in *Function reference*, simply import the control package as follows:

```
>>> import control
```

If you want to have a MATLAB-like environment, use the MATLAB compatibility module:

```
>>> from control.matlab import *
```

### LIBRARY CONVENTIONS

The python-control library uses a set of standard conventions for the way that different types of standard information used by the library.

# 2.1 LTI system representation

Linear time invariant (LTI) systems are represented in python-control in state space, transfer function, or frequency response data (FRD) form. Most functions in the toolbox will operate on any of these data types and functions for converting between compatible types is provided.

### 2.1.1 State space systems

The StateSpace class is used to represent state-space realizations of linear time-invariant (LTI) systems:

$$\frac{dx}{dt} = Ax + Bu$$
$$y = Cx + Du$$

where u is the input, y is the output, and x is the state.

To create a state space system, use the StateSpace constructor:

$$sys = StateSpace(A, B, C, D)$$

State space systems can be manipulated using standard arithmetic operations as well as the feedback(), parallel(), and series() function. A full list of functions can be found in Function reference.

### 2.1.2 Transfer functions

The TransferFunction class is used to represent input/output transfer functions

$$G(s) = \frac{\text{num}(s)}{\text{den}(s)} = \frac{a_0 s^m + a_1 s^{m-1} + \dots + a_m}{b_0 s^n + b_1 s^{n-1} + \dots + b_n},$$

where n is generally greater than or equal to m (for a proper transfer function).

To create a transfer function, use the *TransferFunction* constructor:

sys = TransferFunction(num, den)

Transfer functions can be manipulated using standard arithmetic operations as well as the feedback(), parallel(), and series() function. A full list of functions can be found in Function reference.

### 2.1.3 FRD (frequency response data) systems

The FrequencyResponseData (FRD) class is used to represent systems in frequency response data form.

The main data members are *omega* and *fresp*, where *omega* is a 1D array with the frequency points of the response, and *fresp* is a 3D array, with the first dimension corresponding to the output index of the FRD, the second dimension corresponding to the input index, and the 3rd dimension corresponding to the frequency points in omega.

FRD systems have a somewhat more limited set of functions that are available, although all of the standard algebraic manipulations can be performed.

### 2.1.4 Discrete time systems

A discrete time system is created by specifying a nonzero 'timebase', dt. The timebase argument can be given when a system is constructed:

- dt = None: no timebase specified (default)
- dt = 0: continuous time system
- dt > 0: discrete time system with sampling period 'dt'
- dt = True: discrete time with unspecified sampling period

Only the StateSpace, TransferFunction, and InputOutputSystem classes allow explicit representation of discrete time systems.

Systems must have compatible timebases in order to be combined. A system with timebase *None* can be combined with a system having a specified timebase; the result will have the timebase of the latter system. Similarly, a discrete time system with unspecified sampling time (dt = True) can be combined with a system having a specified sampling time; the result will be a discrete time system with the sample time of the latter system. For continuous time systems, the  $sample\_system()$  function or the StateSpace.sample() and TransferFunction.sample() methods can be used to create a discrete time system from a continuous time system. See Utility functions and conversions. The default value of 'dt' can be changed by changing the values of control.config.defaults['statesp.default\_dt'] and control.config.defaults['xferfcn.default\_dt'].

### 2.1.5 Conversion between representations

LTI systems can be converted between representations either by calling the constructor for the desired data type using the original system as the sole argument or using the explicit conversion functions ss2tf() and tf2ss().

### 2.2 Time series data

A variety of functions in the library return time series data: sequences of values that change over time. A common set of conventions is used for returning such data: columns represent different points in time, rows are different components (e.g., inputs, outputs or states). For return arguments, an array of times is given as the first returned argument, followed by one or more arrays of variable values. This convention is used throughout the library, for example in the functions forced\_response(), step\_response(), impulse\_response(), and initial\_response().

Note:	The convention used by python-control is different from the convention used in the scipy.signal library.	In
Scipy's	convention the meaning of rows and columns is interchanged. Thus, all 2D values must be transposed when the convention the meaning of rows and columns is interchanged.	hen
they are	e used with functions from scipy.signal.	

Types:

- Arguments can be arrays, matrices, or nested lists.
- Return values are arrays (not matrices).

The time vector is either 1D, or 2D with shape (1, n):

```
T = [[t1, t2, t3, ..., tn]]
```

Input, state, and output all follow the same convention. Columns are different points in time, rows are different components. When there is only one row, a 1D object is accepted or returned, which adds convenience for SISO systems:

```
U = [[u1(t1), u1(t2), u1(t3), ..., u1(tn)]
       [u2(t1), u2(t2), u2(t3), ..., u2(tn)]
       ...
       [ui(t1), ui(t2), ui(t3), ..., ui(tn)]]
Same for X, Y
```

So, U[:,2] is the system's input at the third point in time; and U[1] or U[1,:] is the sequence of values for the system's second input.

The initial conditions are either 1D, or 2D with shape (j, 1):

```
X0 = [[x1]
      [x2]
      ...
      [xj]]
```

As all simulation functions return arrays, plotting is convenient:

```
t, y = step_response(sys)
plot(t, y)
```

The output of a MIMO system can be plotted like this:

```
t, y, x = forced_response(sys, u, t)
plot(t, y[0], label='y_0')
plot(t, y[1], label='y_1')
```

The convention also works well with the state space form of linear systems. If D is the feedthrough *matrix* of a linear system, and U is its input (*matrix* or *array*), then the feedthrough part of the system's response, can be computed like this:

```
ft = D * U
```

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# 2.3 Package configuration parameters

The python-control library can be customized to allow for different default values for selected parameters. This includes the ability to set the style for various types of plots and establishing the underlying representation for state space matrices.

To set the default value of a configuration variable, set the appropriate element of the *control.config.defaults* dictionary:

```
control.config.defaults['module.parameter'] = value
```

The ~control.config.set\_defaults function can also be used to set multiple configuration parameters at the same time:

```
control.config.set_defaults('module', param1=val1, param2=val2, ...]
```

Finally, there are also functions available set collections of variables based on standard configurations.

Selected variables that can be configured, along with their default values:

- bode.dB (False): Bode plot magnitude plotted in dB (otherwise powers of 10)
- bode.deg (True): Bode plot phase plotted in degrees (otherwise radians)
- bode.Hz (False): Bode plot frequency plotted in Hertz (otherwise rad/sec)
- bode.grid (True): Include grids for magnitude and phase plots
- freqplot.number\_of\_samples (None): Number of frequency points in Bode plots
- freqplot.feature\_periphery\_decade (1.0): How many decades to include in the frequency range on both sides of features (poles, zeros).
- statesp.use\_numpy\_matrix (True): set the return type for state space matrices to *numpy.matrix* (verus numpy.ndarray)
- statesp.default\_dt and xferfcn.default\_dt (None): set the default value of dt when constructing new LTI systems
- statesp.remove\_useless\_states (True): remove states that have no effect on the input-output dynamics of the system

Additional parameter variables are documented in individual functions

Functions that can be used to set standard configurations:

reset_defaults()	Reset configuration values to their default (initial) val-
	ues.
use_fbs_defaults()	Use Feedback Systems (FBS) compatible settings.
use_matlab_defaults()	Use MATLAB compatible configuration settings.
use_numpy_matrix([flag, warn])	Turn on/off use of Numpy matrix class for state space
	operations.
use_legacy_defaults(version)	Sets the defaults to whatever they were in a given re-
	lease.

### 2.3.1 control.reset defaults

```
control.reset defaults()
```

Reset configuration values to their default (initial) values.

### 2.3.2 control.use fbs defaults

```
control.use_fbs_defaults()
```

Use Feedback Systems (FBS) compatible settings.

#### The following conventions are used:

• Bode plots plot gain in powers of ten, phase in degrees, frequency in rad/sec, no grid

### 2.3.3 control.use matlab defaults

```
control.use_matlab_defaults()
```

Use MATLAB compatible configuration settings.

### The following conventions are used:

- Bode plots plot gain in dB, phase in degrees, frequency in rad/sec, with grids
- State space class and functions use Numpy matrix objects

### 2.3.4 control.use numpy matrix

```
control.use numpy matrix (flag=True, warn=True)
```

Turn on/off use of Numpy *matrix* class for state space operations.

#### **Parameters**

- **flag** (bool) If flag is *True* (default), use the Numpy (soon to be deprecated) *matrix* class to represent matrices in the ~*control.StateSpace* class and functions. If flat is *False*, then matrices are represented by a 2D *ndarray* object.
- warn (bool) If flag is *True* (default), issue a warning when turning on the use of the Numpy *matrix* class. Set *warn* to false to omit display of the warning message.

#### **Notes**

Prior to release 0.9.x, the default type for 2D arrays is the Numpy *matrix* class. Starting in release 0.9.0, the default type for state space operations is a 2D array.

### 2.3.5 control.use legacy defaults

```
control.use_legacy_defaults(version)
```

Sets the defaults to whatever they were in a given release.

**Parameters version** (*string*) – version number of the defaults desired. ranges from '0.1' to '0.8.4'.

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### **FUNCTION REFERENCE**

The Python Control Systems Library control provides common functions for analyzing and designing feedback control systems.

# 3.1 System creation

ss(A, B, C, D[, dt])	Create a state space system.
tf(num, den[, dt])	Create a transfer function system.
frd(d, w)	Construct a frequency response data model
rss([states, outputs, inputs])	Create a stable <i>continuous</i> random state space object.
drss([states, outputs, inputs])	Create a stable <i>discrete</i> random state space object.

### 3.1.1 control.ss

control.ss
$$(A, B, C, D[, dt])$$

Create a state space system.

The function accepts either 1, 4 or 5 parameters:

- **ss (sys)** Convert a linear system into space system form. Always creates a new system, even if sys is already a StateSpace object.
- ss (A, B, C, D) Create a state space system from the matrices of its state and output equations:

$$\dot{x} = A \cdot x + B \cdot u$$

$$y = C \cdot x + D \cdot u$$

ss(A, B, C, D, dt) Create a discrete-time state space system from the matrices of its state and output equations:

$$x[k+1] = A \cdot x[k] + B \cdot u[k]$$
 
$$y[k] = C \cdot x[k] + D \cdot u[ki]$$

The matrices can be given as *array like* data types or strings. Everything that the constructor of numpy. matrix accepts is permissible here too.

- sys (StateSpace or TransferFunction) A linear system
- A (array\_like or string) System matrix

- B(array\_like or string) Control matrix
- C(array\_like or string) Output matrix
- D (array\_like or string) Feed forward matrix
- dt (If present, specifies the sampling period and a discrete time) system is created

**Returns out** – The new linear system

Return type StateSpace

Raises ValueError – if matrix sizes are not self-consistent

#### See also:

```
StateSpace, tf, ss2tf, tf2ss
```

### **Examples**

```
>>> # Create a StateSpace object from four "matrices".
>>> sys1 = ss("1. -2; 3. -4", "5.; 7", "6. 8", "9.")
```

```
>>> # Convert a TransferFunction to a StateSpace object.
>>> sys_tf = tf([2.], [1., 3])
>>> sys2 = ss(sys_tf)
```

### 3.1.2 control.tf

```
control.tf(num, den[, dt])
```

Create a transfer function system. Can create MIMO systems.

The function accepts either 1, 2, or 3 parameters:

- **tf(sys)** Convert a linear system into transfer function form. Always creates a new system, even if sys is already a TransferFunction object.
- **tf(num, den)** Create a transfer function system from its numerator and denominator polynomial coefficients.

If num and den are 1D array\_like objects, the function creates a SISO system.

To create a MIMO system, *num* and *den* need to be 2D nested lists of array\_like objects. (A 3 dimensional data structure in total.) (For details see note below.)

- **tf(num, den, dt)** Create a discrete time transfer function system; dt can either be a positive number indicating the sampling time or 'True' if no specific timebase is given.
- tf('s') or tf('z') Create a transfer function representing the differential operator ('s') or delay operator ('z').

- sys(LTI (StateSpace or TransferFunction)) A linear system
- num (array\_like, or list of list of array\_like) Polynomial coefficients of the numerator

den (array\_like, or list of list of array\_like) - Polynomial coefficients of the denominator

**Returns out** – The new linear system

Return type TransferFunction

Raises

- **ValueError** if *num* and *den* have invalid or unequal dimensions
- **TypeError** if *num* or *den* are of incorrect type

#### See also:

TransferFunction, ss, ss2tf, tf2ss

#### **Notes**

num[i][j] contains the polynomial coefficients of the numerator for the transfer function from the (j+1)st input to the (i+1)st output. den[i][j] works the same way.

The list [2, 3, 4] denotes the polynomial  $2s^2 + 3s + 4$ .

The special forms tf('s') and tf('z') can be used to create transfer functions for differentiation and unit delays.

### **Examples**

```
>>> # Create a MIMO transfer function object

>>> # The transfer function from the 2nd input to the 1st output is

>>> # (3s + 4) / (6s^2 + 5s + 4).

>>> num = [[[1., 2.], [3., 4.]], [[5., 6.], [7., 8.]]]

>>> den = [[[9., 8., 7.], [6., 5., 4.]], [[3., 2., 1.], [-1., -2., -3.]]]

>>> sys1 = tf(num, den)
```

```
>>> # Create a variable 's' to allow algebra operations for SISO systems
>>> s = tf('s')
>>> G = (s + 1)/(s**2 + 2*s + 1)
```

```
>>> # Convert a StateSpace to a TransferFunction object.
>>> sys_ss = ss("1. -2; 3. -4", "5.; 7", "6. 8", "9.")
>>> sys2 = tf(sys1)
```

### 3.1.3 control.frd

```
control. frd(d, w)
```

Construct a frequency response data model

frd models store the (measured) frequency response of a system.

This function can be called in different ways:

frd (response, freqs) Create an frd model with the given response data, in the form of complex response vector, at matching frequency freqs [in rad/s]

frd(sys, freqs) Convert an LTI system into an frd model with data at frequencies freqs.

#### **Parameters**

- response (array\_like, or list) complex vector with the system response
- freq(array\_lik or lis) vector with frequencies
- **sys**(*LTI* (StateSpace or TransferFunction)) A linear system

**Returns** sys – New frequency response system

Return type FRD

#### See also:

FRD, ss, tf

### 3.1.4 control.rss

```
control.rss (states=1, outputs=1, inputs=1)
```

Create a stable *continuous* random state space object.

#### **Parameters**

- **states** (*integer*) Number of state variables
- inputs (integer) Number of system inputs
- outputs (integer) Number of system outputs

**Returns** sys – The randomly created linear system

**Return type** *StateSpace* 

**Raises** ValueError – if any input is not a positive integer

### See also:

drss

### **Notes**

If the number of states, inputs, or outputs is not specified, then the missing numbers are assumed to be 1. The poles of the returned system will always have a negative real part.

### 3.1.5 control.drss

```
control.drss(states=1, outputs=1, inputs=1)
```

Create a stable *discrete* random state space object.

#### **Parameters**

- states (integer) Number of state variables
- inputs (integer) Number of system inputs
- outputs (integer) Number of system outputs

**Returns** sys – The randomly created linear system

Return type StateSpace

**Raises** ValueError – if any input is not a positive integer

#### See also:

rss

### **Notes**

If the number of states, inputs, or outputs is not specified, then the missing numbers are assumed to be 1. The poles of the returned system will always have a magnitude less than 1.

# 3.2 System interconnections

append(sys1, sys2,, sysn)	Group models by appending their inputs and outputs
connect(sys, Q, inputv, outputv)	Index-based interconnection of an LTI system.
feedback(sys1[, sys2, sign])	Feedback interconnection between two I/O systems.
negate(sys)	Return the negative of a system.
parallel(sys1, *sysn)	Return the parallel connection sys1 + sys2 (+ .
series(sys1, *sysn)	Return the series connection (sysn * .

### 3.2.1 control.append

```
control.append(sys1, sys2, ..., sysn)
```

Group models by appending their inputs and outputs

Forms an augmented system model, and appends the inputs and outputs together. The system type will be the type of the first system given; if you mix state-space systems and gain matrices, make sure the gain matrices are not first.

### **Parameters**

- sys1 (StateSpace or Transferfunction) LTI systems to combine
- sys2 (StateSpace or Transferfunction) LTI systems to combine
- .. (StateSpace or Transferfunction) LTI systems to combine
- sysn (StateSpace or Transferfunction) LTI systems to combine

**Returns** sys – Combined LTI system, with input/output vectors consisting of all input/output vectors appended

Return type LTI system

### **Examples**

```
>>> sys1 = ss([[1., -2], [3., -4]], [[5.], [7]]", [[6., 8]], [[9.]])
>>> sys2 = ss([[-1.]], [[1.]], [[0.]])
>>> sys = append(sys1, sys2)
```

### 3.2.2 control.connect

```
control.connect(sys, Q, inputv, outputv)
```

Index-based interconnection of an LTI system.

The system sys is a system typically constructed with append, with multiple inputs and outputs. The inputs and outputs are connected according to the interconnection matrix Q, and then the final inputs and outputs are trimmed according to the inputs and outputs listed in inputv and outputv.

NOTE: Inputs and outputs are indexed starting at 1 and negative values correspond to a negative feedback interconnection.

### **Parameters**

- **sys** (StateSpace Transferfunction) System to be connected
- Q (2D array) Interconnection matrix. First column gives the input to be connected. The second column gives the index of an output that is to be fed into that input. Each additional column gives the index of an additional input that may be optionally added to that input. Negative values mean the feedback is negative. A zero value is ignored. Inputs and outputs are indexed starting at 1 to communicate sign information.
- inputv (1D array) list of final external inputs, indexed starting at 1
- outputv (1D array) list of final external outputs, indexed starting at 1

Returns sys – Connected and trimmed LTI system

Return type LTI system

### **Examples**

```
>>> sys1 = ss([[1., -2], [3., -4]], [[5.], [7]], [[6, 8]], [[9.]])
>>> sys2 = ss([[-1.]], [[1.]], [[0.]])
>>> sys = append(sys1, sys2)
>>> Q = [[1, 2], [2, -1]] # negative feedback interconnection
>>> sysc = connect(sys, Q, [2], [1, 2])
```

#### 3.2.3 control.feedback

```
control.feedback(sys1, sys2=1, sign=-1)
```

Feedback interconnection between two I/O systems.

#### **Parameters**

- **sys1** (*scalar*, StateSpace, TransferFunction, *FRD*) The primary process.
- **sys2** (*scalar*, StateSpace, TransferFunction, *FRD*) The feedback process (often a feedback controller).
- **sign** (*scalar*) The sign of feedback. *sign* = -1 indicates negative feedback, and *sign* = 1 indicates positive feedback. *sign* is an optional argument; it assumes a value of -1 if not specified.

### Returns out

Return type StateSpace or TransferFunction

Raises

- **ValueError** if *sys1* does not have as many inputs as *sys2* has outputs, or if *sys2* does not have as many inputs as *sys1* has outputs
- NotImplementedError if an attempt is made to perform a feedback on a MIMO TransferFunction object

#### See also:

series, parallel

#### **Notes**

This function is a wrapper for the feedback function in the StateSpace and TransferFunction classes. It calls TransferFunction.feedback if *sys1* is a TransferFunction object, and StateSpace.feedback if *sys1* is a StateSpace object. If *sys1* is a scalar, then it is converted to *sys2*'s type, and the corresponding feedback function is used. If *sys1* and *sys2* are both scalars, then TransferFunction.feedback is used.

### 3.2.4 control.negate

```
control.negate(sys)
```

Return the negative of a system.

```
Parameters sys (StateSpace, TransferFunction or FRD) -
```

Returns out

**Return type** *StateSpace* or *TransferFunction* 

#### **Notes**

This function is a wrapper for the <u>\_\_neg\_\_</u> function in the StateSpace and TransferFunction classes. The output type is the same as the input type.

### **Examples**

```
>>> sys2 = negate(sys1) \# Same as sys2 = -sys1.
```

### 3.2.5 control.parallel

```
control.parallel (sys1, *sysn)
Return the parallel connection sys1 + sys2 + ... + sysn)
```

#### **Parameters**

- sys1(scalar, StateSpace, TransferFunction, or FRD) -
- \*sysn (other scalars, StateSpaces, TransferFunctions, or FRDs)

#### Returns out

Return type scalar, StateSpace, or TransferFunction

Raises ValueError – if sys1 and sys2 do not have the same numbers of inputs and outputs

#### See also:

series, feedback

### **Notes**

This function is a wrapper for the \_\_add\_\_ function in the StateSpace and TransferFunction classes. The output type is usually the type of *sys1*. If *sys1* is a scalar, then the output type is the type of *sys2*.

If both systems have a defined timebase (dt = 0 for continuous time, dt > 0 for discrete time), then the timebase for both systems must match. If only one of the system has a timebase, the return timebase will be set to match it.

### **Examples**

```
>>> sys3 = parallel(sys1, sys2) # Same as sys3 = sys1 + sys2
```

```
>>> sys5 = parallel(sys1, sys2, sys3, sys4) # More systems
```

### 3.2.6 control.series

```
control.series (sys1, *sysn)
Return the series connection (sysn * ... *) sys2 * sys1
```

#### **Parameters**

- sys1(scalar, StateSpace, TransferFunction, or FRD) -
- \*sysn (other scalars, StateSpaces, TransferFunctions, or FRDs)

#### Returns out

**Return type** scalar, *StateSpace*, or *TransferFunction* 

**Raises ValueError** – if *sys2.inputs* does not equal *sys1.outputs* if *sys1.dt* is not compatible with *sys2.dt* 

#### See also:

parallel, feedback

### **Notes**

This function is a wrapper for the \_\_mul\_\_ function in the StateSpace and TransferFunction classes. The output type is usually the type of *sys2*. If *sys2* is a scalar, then the output type is the type of *sys1*.

If both systems have a defined timebase (dt = 0 for continuous time, dt > 0 for discrete time), then the timebase for both systems must match. If only one of the system has a timebase, the return timebase will be set to match it.

### **Examples**

```
>>> sys3 = series(sys1, sys2) # Same as sys3 = sys2 * sys1
>>> sys5 = series(sys1, sys2, sys3, sys4) # More systems
```

See also the *Input/output systems* module, which can be used to create and interconnect nonlinear input/output systems.

# 3.3 Frequency domain plotting

bode_plot(syslist[, omega, plot,])	Bode plot for a system
<pre>nyquist_plot(syslist[, omega, plot,])</pre>	Nyquist plot for a system
<pre>gangof4_plot(P, C[, omega])</pre>	Plot the "Gang of 4" transfer functions for a system
nichols_plot(sys_list[, omega, grid])	Nichols plot for a system
nichols_grid([cl_mags, cl_phases, line_style])	Nichols chart grid

### 3.3.1 control.bode\_plot

control.bode\_plot(syslist, omega=None, plot=True, omega\_limits=None, omega\_num=None, margins=None, \*args, \*\*kwargs)

Bode plot for a system

Plots a Bode plot for the system over a (optional) frequency range.

### **Parameters**

- **syslist** (linsys) List of linear input/output systems (single system is OK)
- omega (list) List of frequencies in rad/sec to be used for frequency response
- **dB** (bool) If True, plot result in dB. Default is false.
- **Hz** (bool) If True, plot frequency in Hz (omega must be provided in rad/sec). Default value (False) set by config.defaults['bode.Hz']
- **deg** (bool) If True, plot phase in degrees (else radians). Default value (True) config.defaults['bode.deg']
- plot (bool) If True (default), plot magnitude and phase
- omega\_limits (tuple, list, .. of two values) Limits of the to generate frequency vector. If Hz=True the limits are in Hz otherwise in rad/s.
- omega\_num (int) Number of samples to plot. Defaults to config.defaults['freqplot.number\_of\_samples'].
- margins (bool) If True, plot gain and phase margin.
- \*args (matplotlib.pyplot.plot() positional properties, optional) Additional arguments for *matplotlib* plots (color, linestyle, etc)
- \*\*kwargs (matplotlib.pyplot.plot() keyword properties, optional) Additional keywords (passed to *matplotlib*)

### Returns

• mag (array (list if len(syslist) > 1)) – magnitude

- **phase** (array (list if len(syslist) > 1)) phase in radians
- omega (array (list if len(syslist) > 1)) frequency in rad/sec

Other Parameters grid (bool) – If True, plot grid lines on gain and phase plots. Default is set by config.defaults['bode.grid'].

The default values for Bode plot configuration parameters can be reset using the *config.defaults* dictionary, with module name 'bode'.

#### **Notes**

- 1. Alternatively, you may use the lower-level method (mag, phase, freq) = sys. freqresp (freq) to generate the frequency response for a system, but it returns a MIMO response.
- 2. If a discrete time model is given, the frequency response is plotted along the upper branch of the unit circle, using the mapping z = exp(j omega dt) where omega ranges from 0 to pi/dt and dt is the discrete timebase. If not timebase is specified (dt = True), dt is set to 1.

### **Examples**

```
>>> sys = ss("1. -2; 3. -4", "5.; 7", "6. 8", "9.")
>>> mag, phase, omega = bode(sys)
```

### 3.3.2 control.nyquist plot

control.nyquist\_plot (syslist, omega=None, plot=True, label\_freq=0, arrowhead\_length=0.1, arrowhead\_width=0.1, color=None, \*args, \*\*kwargs)

Nyquist plot for a system

Plots a Nyquist plot for the system over a (optional) frequency range.

### **Parameters**

- syslist (list of LTI) List of linear input/output systems (single system is OK)
- omega (freq range) Range of frequencies (list or bounds) in rad/sec
- Plot (boolean) If True, plot magnitude
- **color** (*string*) Used to specify the color of the plot
- label\_freq (int) Label every nth frequency on the plot
- arrowhead\_width (arrow head width) -
- arrowhead length (arrow head length) -
- \*args (matplotlib.pyplot.plot() positional properties, optional) Additional arguments for *matplotlib* plots (color, linestyle, etc)
- \*\*kwargs (matplotlib.pyplot.plot() keyword properties, optional) Additional keywords (passed to *matplotlib*)

### Returns

- real (array) real part of the frequency response array
- imag (array) imaginary part of the frequency response array
- **freq** (array) frequencies

### **Examples**

```
>>> sys = ss("1. -2; 3. -4", "5.; 7", "6. 8", "9.")
>>> real, imag, freq = nyquist_plot(sys)
```

### 3.3.3 control.gangof4\_plot

```
control.gangof4_plot (P, C, omega=None, **kwargs)

Plot the "Gang of 4" transfer functions for a system
```

Generates a 2x2 plot showing the "Gang of 4" sensitivity functions [T, PS; CS, S]

#### **Parameters**

- **P** (LTI) Linear input/output systems (process and control)
- C (LTI) Linear input/output systems (process and control)
- omega (array) Range of frequencies (list or bounds) in rad/sec
- \*\*kwargs (matplotlib.pyplot.plot() keyword properties, optional) Additional keywords (passed to *matplotlib*)

#### Returns

Return type None

### 3.3.4 control.nichols\_plot

```
control.nichols_plot (sys_list, omega=None, grid=None)
Nichols plot for a system
```

Plots a Nichols plot for the system over a (optional) frequency range.

### **Parameters**

- **sys\_list** (list of LTI, or LTI) List of linear input/output systems (single system is OK)
- omega (array like) Range of frequencies (list or bounds) in rad/sec
- grid (boolean, optional) True if the plot should include a Nichols-chart grid. Default is True.

### Returns

Return type None

### 3.3.5 control.nichols grid

```
control.nichols_grid (cl_mags=None, cl_phases=None, line_style='dotted')
Nichols chart grid
```

Plots a Nichols chart grid on the current axis, or creates a new chart if no plot already exists.

### **Parameters**

• **cl\_mags** (array-like (dB), optional) - Array of closed-loop magnitudes defining the iso-gain lines on a custom Nichols chart.

- cl\_phases (array-like (degrees), optional) Array of closed-loop phases defining the iso-phase lines on a custom Nichols chart. Must be in the range -360 < cl\_phases < 0</li>
- line\_style (string, optional) Matplotlib linestyle

Note: For plotting commands that create multiple axes on the same plot, the individual axes can be retrieved using the axes label (retrieved using the *get\_label* method for the matplotliib axes object). The following labels are currently defined:

- Bode plots: control-bode-magnitude, control-bode-phase
- Gang of 4 plots: control-gangof4-s, control-gangof4-cs, control-gangof4-ps, control-gangof4-t

### 3.4 Time domain simulation

forced_response(sys[, T, U, X0, transpose,])	Simulate the output of a linear system.
<pre>impulse_response(sys[, T, X0, input,])</pre>	Impulse response of a linear system
<pre>initial_response(sys[, T, X0, input,])</pre>	Initial condition response of a linear system
input_output_response(sys, T[, U, X0,])	Compute the output response of a system to a given in-
	put.
step_response(sys[, T, X0, input, output,])	Step response of a linear system
<pre>phase_plot(odefun[, X, Y, scale, X0, T,])</pre>	Phase plot for 2D dynamical systems

### 3.4.1 control.forced\_response

 $\begin{tabular}{ll} ${\tt control.forced\_response}(sys, & T=None, & U=0.0, & X0=0.0, & transpose=False, & interpolate=False, \\ & squeeze=True) \\ & {\tt Simulate} \ \ the \ output \ \ of \ a \ linear \ system. \\ \end{tabular}$ 

As a convenience for parameters U, X0: Numbers (scalars) are converted to constant arrays with the correct shape. The correct shape is inferred from arguments sys and T.

For information on the **shape** of parameters U, T, X0 and return values T, yout, xout, see Time series data.

#### **Parameters**

- **sys** (LTI (StateSpace or TransferFunction)) LTI system to simulate
- **T** (array\_like, optional for discrete LTI *sys*) Time steps at which the input is defined; values must be evenly spaced.
- U (array\_like or float, optional) Input array giving input at each time T (default = 0).

If U is None or 0, a special algorithm is used. This special algorithm is faster than the general algorithm, which is used otherwise.

- **X0** (array\_like or float, optional) Initial condition (default = 0).
- transpose (bool, optional (default=False)) If True, transpose all input and output arrays (for backward compatibility with MATLAB and scipy.signal. lsim())
- interpolate (bool, optional (default=False)) If True and system is a discrete time system, the input will be interpolated between the given time steps and the

output will be given at system sampling rate. Otherwise, only return the output at the times given in *T*. No effect on continuous time simulations (default = False).

• **squeeze** (bool, optional (default=True)) - If True, remove single-dimensional entries from the shape of the output. For single output systems, this converts the output response to a 1D array.

#### Returns

- T (array) Time values of the output.
- yout (array) Response of the system.
- **xout** (*array*) Time evolution of the state vector.

#### See also:

step\_response, initial\_response, impulse\_response

#### **Notes**

For discrete time systems, the input/output response is computed using the scipy.signal.dlsim() function.

For continuous time systems, the output is computed using the matrix exponential  $exp(A\ t)$  and assuming linear interpolation of the inputs between time points.

### **Examples**

```
>>> T, yout, xout = forced_response(sys, T, u, X0)
```

See Time series data.

### 3.4.2 control.impulse\_response

```
control.impulse_response (sys, T=None, X0=0.0, input=0, output=None, T_num=None, transpose=False, return_x=False, squeeze=True)

Impulse response of a linear system
```

If the system has multiple inputs or outputs (MIMO), one input has to be selected for the simulation. Optionally, one output may be selected. The parameters *input* and *output* do this. All other inputs are set to 0, all other outputs are ignored.

For information on the **shape** of parameters T, X0 and return values T, yout, see Time series data.

- sys (StateSpace, TransferFunction) LTI system to simulate
- **T** (array\_like or float, optional) Time vector, or simulation time duration if a scalar (time vector is autocomputed if not given; see step\_response() for more detail)
- **x0** (array\_like or float, optional) Initial condition (default = 0) Numbers are converted to constant arrays with the correct shape.
- input (int) Index of the input that will be used in this simulation.

- **output** (*int*) Index of the output that will be used in this simulation. Set to None to not trim outputs
- **T\_num** (*int*, *optional*) Number of time steps to use in simulation if T is not provided as an array (autocomputed if not given); ignored if sys is discrete-time.
- **transpose** (*bool*) If True, transpose all input and output arrays (for backward compatibility with MATLAB and scipy.signal.lsim())
- **return**  $\mathbf{x}$  (bool) If True, return the state vector (default = False).
- **squeeze** (bool, optional (default=True)) If True, remove single-dimensional entries from the shape of the output. For single output systems, this converts the output response to a 1D array.

#### Returns

- T (array) Time values of the output
- yout (array) Response of the system
- **xout** (*array*) Individual response of each x variable

#### See also:

forced\_response, initial\_response, step\_response

#### **Notes**

This function uses the *forced\_response* function to compute the time response. For continuous time systems, the initial condition is altered to account for the initial impulse.

### **Examples**

```
>>> T, yout = impulse_response(sys, T, X0)
```

### 3.4.3 control.initial\_response

```
control.initial_response (sys, T=None, X0=0.0, input=0, output=None, T_num=None, trans-
pose=False, return_x=False, squeeze=True)
Initial condition response of a linear system
```

If the system has multiple outputs (MIMO), optionally, one output may be selected. If no selection is made for the output, all outputs are given.

For information on the **shape** of parameters T, X0 and return values T, yout, see Time series data.

- sys (StateSpace or TransferFunction) LTI system to simulate
- **T** (array\_like or float, optional) Time vector, or simulation time duration if a number (time vector is autocomputed if not given; see step\_response() for more detail)
- **x0** (array\_like or float, optional) Initial condition (default = 0) Numbers are converted to constant arrays with the correct shape.

- **input** (*int*) Ignored, has no meaning in initial condition calculation. Parameter ensures compatibility with step\_response and impulse\_response
- **output** (*int*) Index of the output that will be used in this simulation. Set to None to not trim outputs
- **T\_num** (*int*, *optional*) Number of time steps to use in simulation if T is not provided as an array (autocomputed if not given); ignored if sys is discrete-time.
- **transpose** (bool) If True, transpose all input and output arrays (for backward compatibility with MATLAB and scipy.signal.lsim())
- **return\_x** (bool) If True, return the state vector (default = False).
- **squeeze** (bool, optional (default=True)) If True, remove single-dimensional entries from the shape of the output. For single output systems, this converts the output response to a 1D array.

#### Returns

- T (array) Time values of the output
- yout (array) Response of the system
- xout (array) Individual response of each x variable

#### See also:

forced\_response, impulse\_response, step\_response

### **Notes**

This function uses the *forced response* function with the input set to zero.

### **Examples**

```
>>> T, yout = initial_response(sys, T, X0)
```

### 3.4.4 control.input output response

```
control.input_output_response(sys, T, U=0.0, X0=0, params=\{\}, method='RK45', return\_x=False, squeeze=True)
```

Compute the output response of a system to a given input.

Simulate a dynamical system with a given input and return its output and state values.

- **sys** (InputOutputSystem) Input/output system to simulate.
- **T** (array-like) Time steps at which the input is defined; values must be evenly spaced.
- U (array-like or number, optional) Input array giving input at each time T (default = 0).
- **X0** (array-like or number, optional) Initial condition (default = 0).
- return\_x (bool, optional) If True, return the values of the state at each time (default = False).

• **squeeze** (bool, optional) – If True (default), squeeze unused dimensions out of the output response. In particular, for a single output system, return a vector of shape (nsteps) instead of (nsteps, 1).

#### Returns

- T (array) Time values of the output.
- yout (array) Response of the system.
- **xout** (*array*) Time evolution of the state vector (if return x=True)

#### Raises

- **TypeError** If the system is not an input/output system.
- ValueError If time step does not match sampling time (for discrete time systems)

### 3.4.5 control.step\_response

```
control.step_response (sys, T=None, X0=0.0, input=None, output=None, T_num=None, transpose=False, return_x=False, squeeze=True)

Step response of a linear system
```

If the system has multiple inputs or outputs (MIMO), one input has to be selected for the simulation. Optionally, one output may be selected. The parameters *input* and *output* do this. All other inputs are set to 0, all other outputs are ignored.

For information on the **shape** of parameters *T*, *X0* and return values *T*, *yout*, see *Time series data*.

#### **Parameters**

- sys (StateSpace or TransferFunction) LTI system to simulate
- **T** (array\_like or float, optional) Time vector, or simulation time duration if a number. If T is not provided, an attempt is made to create it automatically from the dynamics of sys. If sys is continuous-time, the time increment dt is chosen small enough to show the fastest mode, and the simulation time period tfinal long enough to show the slowest mode, excluding poles at the origin and pole-zero cancellations. If this results in too many time steps (>5000), dt is reduced. If sys is discrete-time, only tfinal is computed, and final is reduced if it requires too many simulation steps.
- **x0** (array\_like or float, optional) Initial condition (default = 0)

  Numbers are converted to constant arrays with the correct shape.
- **input** (*int*) Index of the input that will be used in this simulation.
- **output** (*int*) Index of the output that will be used in this simulation. Set to None to not trim outputs
- **T\_num** (*int*, *optional*) Number of time steps to use in simulation if T is not provided as an array (autocomputed if not given); ignored if sys is discrete-time.
- **transpose** (bool) If True, transpose all input and output arrays (for backward compatibility with MATLAB and scipy.signal.lsim())
- **return\_x** (bool) If True, return the state vector (default = False).
- **squeeze** (bool, optional (default=True)) If True, remove single-dimensional entries from the shape of the output. For single output systems, this converts the output response to a 1D array.

#### Returns

- T (array) Time values of the output
- yout (array) Response of the system
- xout (array) Individual response of each x variable

#### See also:

forced\_response, initial\_response, impulse\_response

#### **Notes**

This function uses the *forced\_response* function with the input set to a unit step.

### **Examples**

```
>>> T, yout = step_response(sys, T, X0)
```

### 3.4.6 control.phase\_plot

 $\label{eq:control.phase_plot} $$ (odefun, X=None, Y=None, scale=1, X0=None, T=None, lingrid=None, linetime=None, logtime=None, timepts=None, parms=(), verbose=True) $$ Phase plot for 2D dynamical systems$ 

Produces a vector field or stream line plot for a planar system.

**Call signatures:** phase\_plot(func, X, Y, ...) - display vector field on meshgrid phase\_plot(func, X, Y, scale, ...) - scale arrows phase\_plot(func. X0=(...), T=Tmax, ...) - display stream lines phase\_plot(func, X, Y, X0=[...], T=Tmax, ...) - plot both phase\_plot(func, X0=[...], T=Tmax, lingrid=N, ...) - plot both phase\_plot(func, X0=[...], lintime=N, ...) - stream lines with arrows

- **func** (callable(x, t, ...)) Computes the time derivative of y (compatible with odeint). The function should be the same for as used for scipy.integrate. Namely, it should be a function of the form dxdt = F(x, t) that accepts a state x of dimension 2 and returns a derivative dx/dt of dimension 2.
- **X** (3-element sequences, optional, as [start, stop, npts]) Two 3-element sequences specifying x and y coordinates of a grid. These arguments are passed to linspace and meshgrid to generate the points at which the vector field is plotted. If absent (or None), the vector field is not plotted.
- Y (3-element sequences, optional, as [start, stop, npts]) Two 3-element sequences specifying x and y coordinates of a grid. These arguments are passed to linspace and meshgrid to generate the points at which the vector field is plotted. If absent (or None), the vector field is not plotted.
- scale (float, optional) Scale size of arrows; default = 1
- **X0** (*ndarray* of *initial* conditions, optional) List of initial conditions from which streamlines are plotted. Each initial condition should be a pair of numbers.
- **T** (array-like or number, optional) Length of time to run simulations that generate streamlines. If a single number, the same simulation time is used for all initial conditions. Otherwise, should be a list of length len(X0) that gives the simulation time for each initial condition. Default value = 50.

- lingrid (integer or 2-tuple of integers, optional) Argument is either N or (N, M). If X0 is given and X, Y are missing, a grid of arrows is produced using the limits of the initial conditions, with N grid points in each dimension or N grid points in x and M grid points in y.
- lintime (integer or tuple (integer, float), optional) If a single integer N is given, draw N arrows using equally space time points. If a tuple (N, lambda) is given, draw N arrows using exponential time constant lambda
- timepts (array-like, optional) Draw arrows at the given list times [t1, t2,...]
- parms (tuple, optional) List of parameters to pass to vector field: func(x, t, \*parms)

#### See also:

box\_grid construct box-shaped grid of initial conditions

### **Examples**

# 3.5 Block diagram algebra

series(sys1, *sysn)	Return the series connection (sysn * .
parallel(sys1, *sysn)	Return the parallel connection sys1 + sys2 (+ .
feedback(sys1[, sys2, sign])	Feedback interconnection between two I/O systems.
negate(sys)	Return the negative of a system.

# 3.6 Control system analysis

dcgain(sys)	Return the zero-frequency (or DC) gain of the given sys-
	tem
evalfr(sys, x)	Evaluate the transfer function of an LTI system for a
	single complex number x.
freqresp(sys, omega)	Frequency response of an LTI system at multiple angu-
	lar frequencies.
margin(sysdata)	Calculate gain and phase margins and associated
	crossover frequencies
stability_margins(sysdata[, returnall, epsw])	Calculate stability margins and associated crossover fre-
	quencies.
phase_crossover_frequencies(sys)	Compute frequencies and gains at intersections with real
	axis in Nyquist plot.
pole(sys)	Compute system poles.
zero(sys)	Compute system zeros.
pzmap(sys[, plot, grid, title])	Plot a pole/zero map for a linear system.
root_locus(sys[, kvect, xlim, ylim,])	Root locus plot
sisotool(sys[, kvect, xlim_rlocus,])	Sisotool style collection of plots inspired by MATLAB's
	sisotool.

### 3.6.1 control.dcgain

```
control.dcgain(sys)
```

Return the zero-frequency (or DC) gain of the given system

Returns gain – The zero-frequency gain, or np.nan if the system has a pole at the origin

Return type ndarray

### 3.6.2 control.evalfr

```
control.evalfr(sys, x)
```

Evaluate the transfer function of an LTI system for a single complex number x.

To evaluate at a frequency, enter x = omega\*j, where omega is the frequency in radians

#### **Parameters**

- sys (StateSpace or TransferFunction) Linear system
- **x** (scalar) Complex number

#### Returns fresp

Return type ndarray

#### See also:

fregresp, bode

#### **Notes**

This function is a wrapper for StateSpace.evalfr and TransferFunction.evalfr.

### **Examples**

```
>>> sys = ss("1. -2; 3. -4", "5.; 7", "6. 8", "9.")
>>> evalfr(sys, 1j)
array([[ 44.8-21.4j]])
>>> # This is the transfer function matrix evaluated at s = i.
```

**Todo:** Add example with MIMO system

### 3.6.3 control.freqresp

```
control.freqresp(sys, omega)
```

Frequency response of an LTI system at multiple angular frequencies.

- **sys** (StateSpace or TransferFunction) Linear system
- omega (array\_like) A list of frequencies in radians/sec at which the system should be evaluated. The list can be either a python list or a numpy array and will be sorted before evaluation.

#### Returns

- mag ((self.outputs, self.inputs, len(omega)) ndarray) The magnitude (absolute value, not dB or log10) of the system frequency response.
- **phase** ((self.outputs, self.inputs, len(omega)) ndarray) The wrapped phase in radians of the system frequency response.
- **omega** (*ndarray or list or tuple*) The list of sorted frequencies at which the response was evaluated.

#### See also:

evalfr, bode

#### **Notes**

This function is a wrapper for StateSpace.freqresp and TransferFunction.freqresp. The output omega is a sorted version of the input omega.

### **Examples**

```
>>> sys = ss("1. -2; 3. -4", "5.; 7", "6. 8", "9.")
>>> mag, phase, omega = freqresp(sys, [0.1, 1., 10.])
>>> mag
array([[[ 58.8576682 , 49.64876635, 13.40825927]]])
>>> phase
array([[[-0.05408304, -0.44563154, -0.66837155]]])
```

Todo: Add example with MIMO system

#>>> sys = rss(3, 2, 2) #>>> mag, phase, omega = freqresp(sys, [0.1, 1., 10.]) #>>> mag[0, 1, :] #array([55.43747231, 42.47766549, 1.97225895]) #>>> phase[1, 0, :] #array([-0.12611087, -1.14294316, 2.5764547]) #>>> # This is the magnitude of the frequency response from the 2nd #>>> # input to the 1st output, and the phase (in radians) of the #>>> # frequency response from the 1st input to the 2nd output, for #>>> # s = 0.1i, i, 10i.

### 3.6.4 control.margin

```
control.margin(sysdata)
```

Calculate gain and phase margins and associated crossover frequencies

```
Parameters sysdata (LTI system or (mag, phase, omega) sequence) -
```

sys [StateSpace or TransferFunction] Linear SISO system

mag, phase, omega [sequence of array\_like] Input magnitude, phase (in deg.), and frequencies (rad/sec) from bode frequency response data

#### Returns

- gm (float) Gain margin
- **pm** (*float*) Phase margin (in degrees)
- wg (float) Frequency for gain margin (at phase crossover, phase = -180 degrees)

- wp (*float*) Frequency for phase margin (at gain crossover, gain = 1)
- Margins are calculated for a SISO open-loop system.
- If there is more than one gain crossover, the one at the smallest
- margin (deviation from gain = 1), in absolute sense, is
- returned. Likewise the smallest phase margin (in absolute sense)
- is returned.

### **Examples**

```
>>> sys = tf(1, [1, 2, 1, 0])
>>> gm, pm, wg, wp = margin(sys)
```

### 3.6.5 control.stability\_margins

control.stability\_margins(sysdata, returnall=False, epsw=0.0)

Calculate stability margins and associated crossover frequencies.

#### **Parameters**

• **sysdata**(LTI system or (mag, phase, omega) sequence) **sys** [LTI system] Linear SISO system

**mag, phase, omega** [sequence of array\_like] Arrays of magnitudes (absolute values, not dB), phases (degrees), and corresponding frequencies. Crossover frequencies returned are in the same units as those in *omega* (e.g., rad/sec or Hz).

- **returnall** (bool, optional) If true, return all margins found. If False (default), return only the minimum stability margins. For frequency data or FRD systems, only margins in the given frequency region can be found and returned.
- **epsw** (*float*, *optional*) Frequencies below this value (default 0.0) are considered static gain, and not returned as margin.

#### Returns

- gm (float or array\_like) Gain margin
- **pm** (*float or array\_loke*) Phase margin
- sm (float or array\_like) Stability margin, the minimum distance from the Nyquist plot to
   -1
- wg (*float or array\_like*) Frequency for gain margin (at phase crossover, phase = -180 degrees)
- wp (float or array\_like) Frequency for phase margin (at gain crossover, gain = 1)
- ws (float or array\_like) Frequency for stability margin (complex gain closest to -1)

### 3.6.6 control.phase crossover frequencies

```
control.phase_crossover_frequencies (sys)
```

Compute frequencies and gains at intersections with real axis in Nyquist plot.

```
Parameters sys (SISO LTI system) -
```

#### Returns

- **omega** (*ndarray*) 1d array of (non-negative) frequencies where Nyquist plot intersects the real axis
- gain (ndarray) 1d array of corresponding gains

### **Examples**

### 3.6.7 control.pole

```
control.pole(sys)
```

Compute system poles.

Parameters sys (StateSpace or TransferFunction) - Linear system

**Returns** poles – Array that contains the system's poles.

Return type ndarray

Raises NotImplementedError - when called on a TransferFunction object

### See also:

```
zero, TransferFunction.pole, StateSpace.pole
```

### 3.6.8 control.zero

```
control.zero(sys)
```

Compute system zeros.

Parameters sys (StateSpace or TransferFunction) - Linear system

**Returns** zeros – Array that contains the system's zeros.

Return type ndarray

Raises NotImplementedError - when called on a MIMO system

#### See also:

```
pole, StateSpace.zero, TransferFunction.zero
```

## 3.6.9 control.pzmap

control.pzmap (sys, plot=None, grid=None, title='Pole Zero Map', \*\*kwargs)
Plot a pole/zero map for a linear system.

### **Parameters**

- **sys** (*LTI* (StateSpace or TransferFunction)) Linear system for which poles and zeros are computed.
- plot (bool, optional) If True a graph is generated with Matplotlib, otherwise the poles and zeros are only computed and returned.
- grid (boolean (default = False)) If True plot omega-damping grid.

#### Returns

- **pole** (*array*) The systems poles
- **zeros** (*array*) The system's zeros.

## 3.6.10 control.root locus

Root locus plot

Calculate the root locus by finding the roots of 1+k\*TF(s) where TF is self.num(s)/self.den(s) and each k is an element of kvect.

#### **Parameters**

- **sys** (LTI object) Linear input/output systems (SISO only, for now).
- kvect (list or ndarray, optional)—List of gains to use in computing diagram.
- **xlim** (tuple or list, optional) Set limits of x axis, normally with tuple (see matplotlib.axes).
- ylim (tuple or list, optional) Set limits of y axis, normally with tuple (see matplotlib.axes).
- plotstr (matplotlib.pyplot.plot() format string, optional) plotting style specification
- plot (boolean, optional) If True (default), plot root locus diagram.
- **print\_gain** (bool) If True (default), report mouse clicks when close to the root locus branches, calculate gain, damping and print.
- **grid** (bool) If True plot omega-damping grid. Default is False.
- ax (matplotlib.axes.Axes) Axes on which to create root locus plot

#### Returns

- **rlist** (*ndarray*) Computed root locations, given as a 2D array
- **klist** (*ndarray or list*) Gains used. Same as klist keyword argument if provided.

## 3.6.11 control.sisotool

control.sisotool (sys, kvect=None, xlim\_rlocus=None, ylim\_rlocus=None, plotstr\_rlocus='C0', rlocus\_grid=False, omega=None, dB=None, Hz=None, deg=None, omega\_limits=None, omega\_num=None, margins\_bode=True, tvect=None)

Sisotool style collection of plots inspired by MATLAB's sisotool. The left two plots contain the bode magnitude and phase diagrams. The top right plot is a clickable root locus plot, clicking on the root locus will change the gain of the system. The bottom left plot shows a closed loop time response.

#### **Parameters**

- **sys** (LTI object) Linear input/output systems (SISO only)
- kvect (list or ndarray, optional) List of gains to use for plotting root locus
- xlim\_rlocus (tuple or list, optional) control of x-axis range, normally with tuple (see matplotlib.axes).
- ylim\_rlocus (tuple or list, optional) control of y-axis range
- plotstr\_rlocus (matplotlib.pyplot.plot() format string, optional) plotting style for the root locus plot(color, linestyle, etc)
- rlocus\_grid (boolean (default = False)) If True plot s-plane grid.
- omega (freq\_range) Range of frequencies in rad/sec for the bode plot
- dB (boolean) If True, plot result in dB for the bode plot
- **Hz** (boolean) If True, plot frequency in Hz for the bode plot (omega must be provided in rad/sec)
- **deg** (boolean) If True, plot phase in degrees for the bode plot (else radians)
- omega\_limits (tuple, list, .. of two values) Limits of the to generate frequency vector. If Hz=True the limits are in Hz otherwise in rad/s.
- omega\_num (int) number of samples
- margins bode (boolean) If True, plot gain and phase margin in the bode plot
- **tvect** (*list or ndarray*, *optional*) List of timesteps to use for closed loop step response

## **Examples**

```
>>> sys = tf([1000], [1,25,100,0])
>>> sisotool(sys)
```

# 3.7 Matrix computations

care(A, B, Q[, R, S, E, stabilizing])	(X, L, G) = care(A, B, Q, R=None) solves the
	continuous-time algebraic Riccati equation
dare(A, B, Q, R[, S, E, stabilizing])	(X, L, G) = dare(A, B, Q, R) solves the discrete-time
	algebraic Riccati equation
lyap(A, Q[, C, E])	X = lyap(A, Q) solves the continuous-time Lyapunov
	equation
	continues on next page

Table 7 – continued from previous page

dlyap(A, Q[, C, E])	dlyap(A,Q) solves the discrete-time Lyapunov equation
ctrb(A, B)	Controllabilty matrix
obsv(A, C)	Observability matrix
gram(sys, type)	Gramian (controllability or observability)

### 3.7.1 control.care

control.care(A, B, Q, R=None, S=None, E=None, stabilizing=True)

(X, L, G) = care(A, B, Q, R=None) solves the continuous-time algebraic Riccati equation

$$A^TX + XA - XBR^{-1}B^TX + Q = 0$$

where A and Q are square matrices of the same dimension. Further, Q and R are a symmetric matrices. If R is None, it is set to the identity matrix. The function returns the solution X, the gain matrix  $G = B^T X$  and the closed loop eigenvalues L, i.e., the eigenvalues of A - B G.

(X, L, G) = care(A, B, Q, R, S, E) solves the generalized continuous-time algebraic Riccati equation

$$A^{T}XE + E^{T}XA - (E^{T}XB + S)R^{-1}(B^{T}XE + S^{T}) + Q = 0$$

where A, Q and E are square matrices of the same dimension. Further, Q and R are symmetric matrices. If R is None, it is set to the identity matrix. The function returns the solution X, the gain matrix  $G = R^{-1}$  (B^T X E + S^T) and the closed loop eigenvalues L, i.e., the eigenvalues of A - B G, E.

#### **Parameters**

- A (2D arrays) Input matrices for the Riccati equation
- **B** (2D arrays) Input matrices for the Riccati equation
- Q (2D arrays) Input matrices for the Riccati equation
- R (2D arrays, optional) Input matrices for generalized Riccati equation
- **S** (2D arrays, optional) Input matrices for generalized Riccati equation
- E (2D arrays, optional) Input matrices for generalized Riccati equation

### Returns

- X (2D array (or matrix)) Solution to the Ricatti equation
- L (1D array) Closed loop eigenvalues
- **G** (2D array (or matrix)) Gain matrix

### **Notes**

The return type for 2D arrays depends on the default class set for state space operations. See  $use\_numpy\_matrix()$ .

## 3.7.2 control.dare

control.dare(A, B, Q, R, S=None, E=None, stabilizing=True)

(X, L, G) = dare(A, B, Q, R) solves the discrete-time algebraic Riccati equation

$$A^{T}XA - X - A^{T}XB(B^{T}XB + R)^{-1}B^{T}XA + Q = 0$$

where A and Q are square matrices of the same dimension. Further, Q is a symmetric matrix. The function returns the solution X, the gain matrix  $G = (B^T X B + R)^{-1} B^T X A$  and the closed loop eigenvalues L, i.e., the eigenvalues of A - B G.

(X, L, G) = dare(A, B, Q, R, S, E) solves the generalized discrete-time algebraic Riccati equation

$$A^{T}XA - E^{T}XE - (A^{T}XB + S)(B^{T}XB + R)^{-1}(B^{T}XA + S^{T}) + Q = 0$$

where A, Q and E are square matrices of the same dimension. Further, Q and R are symmetric matrices. The function returns the solution X, the gain matrix  $G = (B^T X B + R)^{-1} (B^T X A + S^T)$  and the closed loop eigenvalues L, i.e., the eigenvalues of A - B G, E.

#### **Parameters**

- A (2D arrays) Input matrices for the Riccati equation
- **B** (2D arrays) Input matrices for the Riccati equation
- Q (2D arrays) Input matrices for the Riccati equation
- R (2D arrays, optional) Input matrices for generalized Riccati equation
- **S** (2D arrays, optional) Input matrices for generalized Riccati equation
- E (2D arrays, optional) Input matrices for generalized Riccati equation

### Returns

- X (2D array (or matrix)) Solution to the Ricatti equation
- L (1D array) Closed loop eigenvalues
- **G** (2D array (or matrix)) Gain matrix

### **Notes**

The return type for 2D arrays depends on the default class set for state space operations. See  $use\_numpy\_matrix()$ .

## 3.7.3 control.lyap

control.lyap (A, Q, C=None, E=None)

X = lyap(A, Q) solves the continuous-time Lyapunov equation

$$AX + XA^T + Q = 0$$

where A and Q are square matrices of the same dimension. Further, Q must be symmetric.

X = lyap(A, Q, C) solves the Sylvester equation

$$AX + XQ + C = 0$$

where A and Q are square matrices.

X = lyap(A, Q, None, E) solves the generalized continuous-time Lyapunov equation

$$AXE^T + EXA^T + Q = 0$$

where Q is a symmetric matrix and A, Q and E are square matrices of the same dimension.

### **Parameters**

- A (2D array) Dynamics matrix
- C (2D array, optional) If present, solve the Slyvester equation
- **E** (2D array, optional) If present, solve the generalized Laypunov equation

**Returns Q** – Solution to the Lyapunov or Sylvester equation

**Return type** 2D array (or matrix)

### **Notes**

The return type for 2D arrays depends on the default class set for state space operations. See  $use\_numpy\_matrix()$ .

## 3.7.4 control.dlyap

control.dlyap (A, Q, C=None, E=None)

dlyap(A,Q) solves the discrete-time Lyapunov equation

$$AXA^T - X + Q = 0$$

where A and Q are square matrices of the same dimension. Further Q must be symmetric.

dlyap(A,Q,C) solves the Sylvester equation

$$AXQ^T - X + C = 0$$

where A and Q are square matrices.

dlyap(A,Q,None,E) solves the generalized discrete-time Lyapunov equation

$$AXA^T - EXE^T + Q = 0$$

where Q is a symmetric matrix and A, Q and E are square matrices of the same dimension.

## 3.7.5 control.ctrb

control.ctrb (A, B)

Controllabilty matrix

## **Parameters**

- A (array\_like or string) Dynamics and input matrix of the system
- B (array\_like or string) Dynamics and input matrix of the system

**Returns** C – Controllability matrix

**Return type** 2D array (or matrix)

The return type for 2D arrays depends on the default class set for state space operations. See  $use\_numpy\_matrix()$ .

## **Examples**

```
>>> C = ctrb(A, B)
```

## 3.7.6 control.obsv

```
control.obsv(A, C)
Observability matrix
```

### **Parameters**

- A (array\_like or string) Dynamics and output matrix of the system
- C (array\_like or string) Dynamics and output matrix of the system

**Returns** O – Observability matrix

Return type 2D array (or matrix)

### **Notes**

The return type for 2D arrays depends on the default class set for state space operations. See  $use\_numpy\_matrix()$ .

## **Examples**

```
>>> O = obsv(A, C)
```

## 3.7.7 control.gram

```
control.gram(sys, type)
```

Gramian (controllability or observability)

### **Parameters**

- $\operatorname{\textbf{sys}}$  (StateSpace) System description
- **type** (*String*) Type of desired computation. *type* is either 'c' (controllability) or 'o' (observability). To compute the Cholesky factors of Gramians use 'cf' (controllability) or 'of' (observability)

**Returns** gram – Gramian of system

**Return type** 2D array (or matrix)

Raises

• ValueError -

- if system is not instance of StateSpace class \* if *type* is not 'c', 'o', 'cf' or 'of' \* if system is unstable (sys.A has eigenvalues not in left half plane)
- ImportError if slycot routine sb03md cannot be found if slycot routine sb03od cannot be found

The return type for 2D arrays depends on the default class set for state space operations. See  $use\_numpy\_matrix()$ .

## **Examples**

```
>>> Wc = gram(sys, 'c')
>>> Wo = gram(sys, 'o')
>>> Rc = gram(sys, 'cf'), where Wc = Rc' * Rc
>>> Ro = gram(sys, 'of'), where Wo = Ro' * Ro
```

# 3.8 Control system synthesis

acker(A, B, poles)	Pole placement using Ackermann method
h2syn(P, nmeas, ncon)	H_2 control synthesis for plant P.
hinfsyn(P, nmeas, ncon)	H_{inf} control synthesis for plant P.
Iqr(A, B, Q, R[, N])	Linear quadratic regulator design
1qe(A, G, C, QN, RN, [, N])	Linear quadratic estimator design (Kalman filter) for
	continuous-time systems.
mixsyn(g[, w1, w2, w3])	Mixed-sensitivity H-infinity synthesis.
place(A, B, p)	Place closed loop eigenvalues

## 3.8.1 control.acker

control.acker(A, B, poles)

Pole placement using Ackermann method

Call: K = acker(A, B, poles)

### **Parameters**

- A (2D arrays) State and input matrix of the system
- **B** (2D arrays) State and input matrix of the system
- poles (1D list) Desired eigenvalue locations

**Returns** K – Gains such that A - B K has given eigenvalues

**Return type** 2D array (or matrix)

The return type for 2D arrays depends on the default class set for state space operations. See  $use\_numpy\_matrix()$ .

## 3.8.2 control.h2syn

```
control.h2syn (P, nmeas, ncon) H_2 control synthesis for plant P.
```

### **Parameters**

- P(partitioned lti plant (State-space sys))-
- nmeas (number of measurements (input to controller)) -
- ncon(number of control inputs (output from controller))-

### Returns K

**Return type** controller to stabilize P (State-space sys)

**Raises** ImportError – if slycot routine sb10hd is not loaded

### See also:

StateSpace

## **Examples**

```
>>> K = h2syn(P,nmeas,ncon)
```

## 3.8.3 control.hinfsyn

```
control.hinfsyn(P, nmeas, ncon)
H_{inf} control synthesis for plant P.
```

### **Parameters**

- P(partitioned lti plant) -
- nmeas(number of measurements (input to controller))-
- ncon (number of control inputs (output from controller)) -

## Returns

- **K** (controller to stabilize P (State-space sys))
- CL (closed loop system (State-space sys))
- gam (infinity norm of closed loop system)
- **rcond** (*4-vector, reciprocal condition estimates of:*) 1: control transformation matrix 2: measurement transformation matrix 3: X-Riccati equation 4: Y-Riccati equation
- **TODO** (document significance of rcond)

**Raises** ImportError – if slycot routine sb10ad is not loaded

### See also:

StateSpace

## **Examples**

```
>>> K, CL, gam, rcond = hinfsyn(P,nmeas,ncon)
```

## 3.8.4 control.lqr

control.lqr
$$(A, B, Q, R[, N])$$

Linear quadratic regulator design

The lqr() function computes the optimal state feedback controller that minimizes the quadratic cost

$$J = \int_0^\infty (x'Qx + u'Ru + 2x'Nu)dt$$

The function can be called with either 3, 4, or 5 arguments:

- lqr(sys, Q, R)
- lqr(sys, Q, R, N)
- lqr(A, B, Q, R)
- lqr(A, B, Q, R, N)

where sys is an LTI object, and A, B, Q, R, and N are 2d arrays or matrices of appropriate dimension.

### **Parameters**

- A (2D array) Dynamics and input matrices
- B (2D array) Dynamics and input matrices
- sys(LTI (StateSpace or TransferFunction)) Linear I/O system
- Q (2D array) State and input weight matrices
- **R** (2D array) State and input weight matrices
- N (2D array, optional) Cross weight matrix

### Returns

- **K** (2D array (or matrix)) State feedback gains
- S (2D array (or matrix)) Solution to Riccati equation
- E (1D array) Eigenvalues of the closed loop system

## See also:

1ge

The return type for 2D arrays depends on the default class set for state space operations. See  $use\_numpy\_matrix()$ .

### **Examples**

```
>>> K, S, E = lqr(sys, Q, R, [N])
>>> K, S, E = lqr(A, B, Q, R, [N])
```

## 3.8.5 control.lge

control.lqe(A, G, C, QN, RN[, N])

Linear quadratic estimator design (Kalman filter) for continuous-time systems. Given the system

$$x = Ax + Bu + Gw$$
$$y = Cx + Du + v$$

with unbiased process noise w and measurement noise v with covariances

$$Eww' = QN, Evv' = RN, Ewv' = NN$$

The lqe() function computes the observer gain matrix L such that the stationary (non-time-varying) Kalman filter

$$x_e = Ax_e + Bu + L(y - Cx_e - Du)$$

produces a state estimate that x\_e that minimizes the expected squared error using the sensor measurements y. The noise cross-correlation NN is set to zero when omitted.

### **Parameters**

- A (2D array) Dynamics and noise input matrices
- **G** (2D array) Dynamics and noise input matrices
- QN (2D array) Process and sensor noise covariance matrices
- RN (2D array) Process and sensor noise covariance matrices
- NN (2D array, optional) Cross covariance matrix

### Returns

- L (2D array (or matrix)) Kalman estimator gain
- P (2D array (or matrix)) Solution to Riccati equation

$$AP + PA^{T} - (PC^{T} + GN)R^{-1}(CP + N^{T}G^{T}) + GQG^{T} = 0$$

• E (1D array) – Eigenvalues of estimator poles eig(A - L C)

The return type for 2D arrays depends on the default class set for state space operations. See  $use\_numpy\_matrix()$ .

### **Examples**

```
>>> K, P, E = lqe(A, G, C, QN, RN)
>>> K, P, E = lqe(A, G, C, QN, RN, NN)
```

#### See also:

lgr

## 3.8.6 control.mixsyn

```
control.mixsyn (g, w1=None, w2=None, w3=None)
Mixed-sensitivity H-infinity synthesis.
mixsyn(g,w1,w2,w3) -> k,cl,info
```

#### **Parameters**

- g(LTI; the plant for which controller must be synthesized)-
- w1 (weighting on s = (1+g\*k)\*\*-1; None, or scalar or k1-by-ny LTI)-
- w2 (weighting on k\*s; None, or scalar or k2-by-nu LTI) -
- w3 (weighting on t = g\*k\*(1+g\*k)\*\*-1; None, or scalar or k3-by-ny LTI)-
- least one of w1 (At) -
- w2 -
- w3 must not be None. (and) -

## Returns

- k (synthesized controller; StateSpace object)
- cl (closed system mapping evaluation inputs to evaluation outputs; if)
- p is the augmented plant, with  $-[z] = [p11 \ p12] [w], [y] [p21 \ g] [u]$
- then cl is the system from w > z with u = -k\*y. StateSpace object.
- info (tuple with entries, in order,)
  - gamma: scalar; H-infinity norm of cl
  - rcond: array; estimates of reciprocal condition numbers computed during synthesis. See hinfsyn for details
- If a weighting w is scalar, it will be replaced by  $I^*w$ , where I is
- ny-by-ny for w1 and w3, and nu-by-nu for w2.

## See also:

hinfsyn, augw

## 3.8.7 control.place

```
control.place (A, B, p)
Place closed loop eigenvalues
K = place(A, B, p)
```

#### **Parameters**

- A (2D array) Dynamics matrix
- B (2D array) Input matrix
- **p** (1D list) Desired eigenvalue locations

**Returns** K – Gain such that A - B K has eigenvalues given in p

**Return type** 2D array (or matrix)

#### **Notes**

**Algorithm** This is a wrapper function for <code>scipy.signal.place\_poles()</code>, which implements the Tits and Yang algorithm<sup>1</sup>. It will handle SISO, MISO, and MIMO systems. If you want more control over the algorithm, use <code>scipy.signal.place\_poles()</code> directly.

**Limitations** The algorithm will not place poles at the same location more than rank(B) times.

The return type for 2D arrays depends on the default class set for state space operations. See  $use\_numpy\_matrix()$ .

### References

### **Examples**

```
>>> A = [[-1, -1], [0, 1]]
>>> B = [[0], [1]]
>>> K = place(A, B, [-2, -5])
```

## See also:

```
place_varga, acker
```

### **Notes**

The return type for 2D arrays depends on the default class set for state space operations. See <code>use\_numpy\_matrix()</code>.

<sup>&</sup>lt;sup>1</sup> A.L. Tits and Y. Yang, "Globally convergent algorithms for robust pole assignment by state feedback, IEEE Transactions on Automatic Control, Vol. 41, pp. 1432-1452, 1996.

# 3.9 Model simplification tools

minreal(sys[, tol, verbose])	Eliminates uncontrollable or unobservable states in state-space models or cancelling pole-zero pairs in transfer functions.
balred(sys, orders[, method, alpha])	Balanced reduced order model of sys of a given order.
hsvd(sys)	Calculate the Hankel singular values.
modred(sys, ELIM[, method])	Model reduction of <i>sys</i> by eliminating the states in <i>ELIM</i> using a given method.
era(YY, m, n, nin, nout, r)	Calculate an ERA model of order <i>r</i> based on the impulse-response data <i>YY</i> .
markov(Y, U[, m, transpose])	Calculate the first $m$ Markov parameters [D CB CAB] from input $U$ , output $Y$ .

## 3.9.1 control.minreal

control.minreal (sys, tol=None, verbose=True)

Eliminates uncontrollable or unobservable states in state-space models or cancelling pole-zero pairs in transfer functions. The output sysr has minimal order and the same response characteristics as the original model sys.

#### **Parameters**

- sys (StateSpace or TransferFunction) Original system
- tol (real) Tolerance
- verbose (bool) Print results if True

**Returns** rsys – Cleaned model

**Return type** *StateSpace* or *TransferFunction* 

## 3.9.2 control.balred

control.balred(sys, orders, method='truncate', alpha=None)

Balanced reduced order model of sys of a given order. States are eliminated based on Hankel singular value. If sys has unstable modes, they are removed, the balanced realization is done on the stable part, then reinserted in accordance with the reference below.

Reference: Hsu,C.S., and Hou,D., 1991, Reducing unstable linear control systems via real Schur transformation. Electronics Letters, 27, 984-986.

### **Parameters**

- **sys** (StateSpace) Original system to reduce
- orders (integer or array of integer) Desired order of reduced order model (if a vector, returns a vector of systems)
- method (string) Method of removing states, either 'truncate' or 'matchdc'.
- alpha (float) Redefines the stability boundary for eigenvalues of the system matrix A. By default for continuous-time systems, alpha <= 0 defines the stability boundary for the real part of A's eigenvalues and for discrete-time systems, 0 <= alpha <= 1 defines the stability boundary for the modulus of A's eigenvalues. See SLICOT routines AB09MD and AB09ND for more information.

**Returns** rsys – A reduced order model or a list of reduced order models if orders is a list.

Return type StateSpace

Raises

- ValueError If method is not 'truncate' or 'matchdc'
- ImportError if slycot routine ab09ad, ab09md, or ab09nd is not found
- **ValueError** if there are more unstable modes than any value in orders

## **Examples**

```
>>> rsys = balred(sys, orders, method='truncate')
```

## 3.9.3 control.hsvd

```
control.hsvd(sys)
```

Calculate the Hankel singular values.

Parameters sys (StateSpace) – A state space system

**Returns** H – A list of Hankel singular values

Return type array

See also:

gram

### **Notes**

The Hankel singular values are the singular values of the Hankel operator. In practice, we compute the square root of the eigenvalues of the matrix formed by taking the product of the observability and controllability gramians. There are other (more efficient) methods based on solving the Lyapunov equation in a particular way (more details soon).

## **Examples**

```
>>> H = hsvd(sys)
```

## 3.9.4 control.modred

```
control.modred(sys, ELIM, method='matchdc')
```

Model reduction of sys by eliminating the states in ELIM using a given method.

### **Parameters**

- **sys** (StateSpace) Original system to reduce
- **ELIM** (array) Vector of states to eliminate
- method (string) Method of removing states in ELIM: either 'truncate' or 'matchdc'.

**Returns** rsys – A reduced order model

Return type StateSpace

**Raises** ValueError – Raised under the following conditions:

- \* if method is not either 'matchdc' or 'truncate'
- \* if eigenvalues of sys.A are not all in left half plane (sys must be stable)

### **Examples**

```
>>> rsys = modred(sys, ELIM, method='truncate')
```

## 3.9.5 control.era

control.era (YY, m, n, nin, nout, r)

Calculate an ERA model of order r based on the impulse-response data YY.

**Note:** This function is not implemented yet.

### **Parameters**

- YY (array) nout x nin dimensional impulse-response data
- m (integer) Number of rows in Hankel matrix
- **n** (integer) Number of columns in Hankel matrix
- nin (integer) Number of input variables
- nout (integer) Number of output variables
- r (integer) Order of model

**Returns** sys – A reduced order model sys=ss(Ar,Br,Cr,Dr)

Return type StateSpace

### **Examples**

```
>>> rsys = era(YY, m, n, nin, nout, r)
```

## 3.9.6 control.markov

```
control.markov(Y, U, m=None, transpose=None)
```

Calculate the first m Markov parameters [D CB CAB ...] from input U, output Y.

This function computes the Markov parameters for a discrete time system

$$x[k+1] = Ax[k] + Bu[k]$$
$$y[k] = Cx[k] + Du[k]$$

given data for u and y. The algorithm assumes that that C A $^k$ B = 0 for k > m-2 (see $^1$ ). Note that the problem is ill-posed if the length of the input data is less than the desired number of Markov parameters (a warning message is generated in this case).

### **Parameters**

- Y (array\_like) Output data. If the array is 1D, the system is assumed to be single input. If the array is 2D and transpose=False, the columns of Y are taken as time points, otherwise the rows of Y are taken as time points.
- **U** (array\_like) Input data, arranged in the same way as Y.
- m (int, optional) Number of Markov parameters to output. Defaults to len(U).
- **transpose** (bool, optional) Assume that input data is transposed relative to the standard *Time series data*. The default value is true for backward compatibility with legacy code.

**Returns** H – First m Markov parameters, [D CB CAB ...]

**Return type** ndarray

### References

#### **Notes**

Currently only works for SISO systems.

This function does not currently comply with the Python Control Library *Time series data* for representation of time series data. Use *transpose=False* to make use of the standard convention (this will be updated in a future release).

### **Examples**

```
>>> T = numpy.linspace(0, 10, 100)

>>> U = numpy.ones((1, 100))

>>> T, Y, _ = forced_response(tf([1], [1, 0.5], True), T, U)

>>> H = markov(Y, U, 3, transpose=False)
```

# 3.10 Nonlinear system support

find_eqpt(sys, x0[, u0, y0, t, params, iu,])	Find the equilibrium point for an input/output system.
<pre>linearize(sys, xeq[, ueq, t, params])</pre>	Linearize an input/output system at a given state and in-
	put.
$input\_output\_response(sys, T[, U, X0,])$	Compute the output response of a system to a given in-
	put.
ss2io(*args, **kw)	Create an I/O system from a state space linear system.
tf2io(*args, **kw)	Convert a transfer function into an I/O system
flatsys.point_to_point(sys, x0, u0, xf, uf, Tf)	Compute trajectory between an initial and final condi-
	tions.

<sup>&</sup>lt;sup>1</sup> J.-N. Juang, M. Phan, L. G. Horta, and R. W. Longman, Identification of observer/Kalman filter Markov parameters - Theory and experiments. Journal of Guidance Control and Dynamics, 16(2), 320-329, 2012. http://doi.org/10.2514/3.21006

## 3.10.1 control.iosys.find\_eqpt

control.iosys.**find\_eqpt** (sys, x0, u0=[], y0=None, t=0,  $params=\{\}$ , iu=None, iy=None, ix=None, idx=None, dx0=None,  $return\_y=False$ ,  $return\_result=False$ , \*\*kw) Find the equilibrium point for an input/output system.

Returns the value of an equlibrium point given the initial state and either input value or desired output value for the equilibrium point.

#### **Parameters**

- **x0** (list of initial state values) Initial guess for the value of the state near the equilibrium point.
- **u0** (*list of input values*, *optional*) If  $y\theta$  is not specified, sets the equilibrium value of the input. If  $y\theta$  is given, provides an initial guess for the value of the input. Can be omitted if the system does not have any inputs.
- y0 (list of output values, optional) If specified, sets the desired values of the outputs at the equilibrium point.
- t (float, optional) Evaluation time, for time-varying systems
- params (dict, optional) Parameter values for the system. Passed to the evaluation functions for the system as default values, overriding internal defaults.
- iu (list of input indices, optional) If specified, only the inputs with the given indices will be fixed at the specified values in solving for an equilibrium point. All other inputs will be varied. Input indices can be listed in any order.
- iy (list of output indices, optional) If specified, only the outputs with the given indices will be fixed at the specified values in solving for an equilibrium point. All other outputs will be varied. Output indices can be listed in any order.
- ix (list of state indices, optional) If specified, states with the given indices will be fixed at the specified values in solving for an equilibrium point. All other states will be varied. State indices can be listed in any order.
- dx0 (list of update values, optional)—If specified, the value of update map must match the listed value instead of the default value of 0.
- idx (list of state indices, optional) If specified, state updates with the given indices will have their update maps fixed at the values given in dx0. All other update values will be ignored in solving for an equilibrium point. State indices can be listed in any order. By default, all updates will be fixed at dx0 in searching for an equilibrium point.
- return\_y (bool, optional) If True, return the value of output at the equilibrium point.
- return\_result (bool, optional) If True, return the result option from the scipy.optimize.root() function used to compute the equilibrium point.

### Returns

- **xeq** (*array of states*) Value of the states at the equilibrium point, or *None* if no equilibrium point was found and *return\_result* was False.
- **ueq** (*array of input values*) Value of the inputs at the equilibrium point, or *None* if no equilibrium point was found and *return result* was False.
- **yeq** (*array of output values, optional*) If *return\_y* is True, returns the value of the outputs at the equilibrium point, or *None* if no equilibrium point was found and *return\_result* was False.

• **result** (scipy.optimize.OptimizeResult, optional) – If *return\_result* is True, returns the *result* from the scipy.optimize.root() function.

## 3.10.2 control.iosys.linearize

```
control.iosys.linearize (sys, xeq, ueq=[], t=0, params={}, **kw) Linearize an input/output system at a given state and input.
```

This function computes the linearization of an input/output system at a given state and input value and returns a control.StateSpace object. The eavaluation point need not be an equilibrium point.

#### **Parameters**

- **sys** (InputOutputSystem) The system to be linearized
- **xeq** (array) The state at which the linearization will be evaluated (does not need to be an equlibrium state).
- **ueq** (array) The input at which the linearization will be evaluated (does not need to correspond to an equlibrium state).
- t (float, optional) The time at which the linearization will be computed (for time-varying systems).
- params (dict, optional) Parameter values for the systems. Passed to the evaluation functions for the system as default values, overriding internal defaults.

**Returns** ss\_sys - The linearization of the system, as a LinearIOSystem object (which is also a StateSpace object.

Return type LinearIOSystem

### 3.10.3 control.iosys.input output response

```
control.iosys.input_output_response (sys, T, U=0.0, X0=0, params={}, method='RK45', return x=False, squeeze=True)
```

Compute the output response of a system to a given input.

Simulate a dynamical system with a given input and return its output and state values.

### **Parameters**

- **sys** (InputOutputSystem) Input/output system to simulate.
- **T** (array-like) Time steps at which the input is defined; values must be evenly spaced.
- **U** (array-like or number, optional) Input array giving input at each time *T* (default = 0).
- **XO** (array-like or number, optional) Initial condition (default = 0).
- return\_x (bool, optional) If True, return the values of the state at each time (default = False).
- **squeeze** (bool, optional) If True (default), squeeze unused dimensions out of the output response. In particular, for a single output system, return a vector of shape (nsteps) instead of (nsteps, 1).

### Returns

• **T** (*array*) – Time values of the output.

- yout (array) Response of the system.
- **xout** (*array*) Time evolution of the state vector (if return\_x=True)

#### Raises

- **TypeError** If the system is not an input/output system.
- **ValueError** If time step does not match sampling time (for discrete time systems)

## 3.10.4 control.iosys.ss2io

```
control.iosys.ss2io(*args, **kw)
```

Create an I/O system from a state space linear system.

Converts a *StateSpace* system into an InputOutputSystem with the same inputs, outputs, and states. The new system can be a continuous or discrete time system

#### **Parameters**

- linsys (StateSpace) LTI StateSpace system to be converted
- **inputs** (*int*, *list* of *str* or *None*, *optional*) Description of the system inputs. This can be given as an integer count or as a list of strings that name the individual signals. If an integer count is specified, the names of the signal will be of the form *s[i]* (where *s* is one of *u*, *y*, or *x*). If this parameter is not given or given as *None*, the relevant quantity will be determined when possible based on other information provided to functions using the system.
- outputs (int, list of str or None, optional) Description of the system outputs. Same format as *inputs*.
- **states** (int, list of str, or None, optional) Description of the system states. Same format as *inputs*.
- dt (None, True or float, optional) System timebase. None (default) indicates continuous time, True indicates discrete time with undefined sampling time, positive number is discrete time with specified sampling time.
- params (dict, optional) Parameter values for the systems. Passed to the evaluation functions for the system as default values, overriding internal defaults.
- name (string, optional) System name (used for specifying signals). If unspecified, a generic name <sys[id]> is generated with a unique integer id.

**Returns** iosys – Linear system represented as an input/output system

Return type LinearIOSystem

## 3.10.5 control.iosys.tf2io

```
control.iosys.tf2io(*args, **kw)
```

Convert a transfer function into an I/O system

## 3.10.6 control.flatsys.point to point

control.flatsys.**point\_to\_point** (*sys*, *x0*, *u0*, *xf*, *uf*, *Tf*, *T0*=0, *basis*=*None*, *cost*=*None*) Compute trajectory between an initial and final conditions.

Compute a feasible trajectory for a differentially flat system between an initial condition and a final condition.

#### **Parameters**

- **flatsys** (FlatSystem object) Description of the differentially flat system. This object must define a function flatsys.forward() that takes the system state and produceds the flag of flat outputs and a system flatsys.reverse() that takes the flag of the flat output and prodes the state and input.
- **x0** (1D arrays) Define the desired initial and final conditions for the system. If any of the values are given as None, they are replaced by a vector of zeros of the appropriate dimension.
- **u0** (1D arrays) Define the desired initial and final conditions for the system. If any of the values are given as None, they are replaced by a vector of zeros of the appropriate dimension.
- **xf** (1D arrays) Define the desired initial and final conditions for the system. If any of the values are given as None, they are replaced by a vector of zeros of the appropriate dimension.
- **uf** (1D arrays) Define the desired initial and final conditions for the system. If any of the values are given as None, they are replaced by a vector of zeros of the appropriate dimension.
- **Tf** (float) The final time for the trajectory (corresponding to xf)
- **TO** (float (optional)) The initial time for the trajectory (corresponding to x0). If not specified, its value is taken to be zero.
- **basis** (BasisFamily object (optional)) The basis functions to use for generating the trajectory. If not specified, the PolyFamily basis family will be used, with the minimal number of elements required to find a feasible trajectory (twice the number of system states)

**Returns traj** – The system trajectory is returned as an object that implements the eval() function, we can be used to compute the value of the state and input and a given time t.

Return type SystemTrajectory object

# 3.11 Utility functions and conversions

augw(g[, w1, w2, w3])	Augment plant for mixed sensitivity problem.
canonical_form(xsys[, form])	Convert a system into canonical form
damp(sys[, doprint])	Compute natural frequency, damping ratio, and poles of
	a system
db2mag(db)	Convert a gain in decibels (dB) to a magnitude
<pre>isctime(sys[, strict])</pre>	Check to see if a system is a continuous-time system
isdtime(sys[, strict])	Check to see if a system is a discrete time system
issiso(sys[, strict])	Check to see if a system is single input, single output
issys(obj)	Return True if an object is a system, otherwise False

continues on next page

Table 11 – continued from previous page

	i i i
mag2db(mag)	Convert a magnitude to decibels (dB)
observable_form(xsys)	Convert a system into observable canonical form
pade(T[, n, numdeg])	Create a linear system that approximates a delay.
reachable_form(xsys)	Convert a system into reachable canonical form
reset_defaults()	Reset configuration values to their default (initial) val-
	ues.
sample_system(sysc, Ts[, method, alpha,])	Convert a continuous time system to discrete time
ss2tf(sys)	Transform a state space system to a transfer function.
ssdata(sys)	Return state space data objects for a system
tf2ss(sys)	Transform a transfer function to a state space system.
tfdata(sys)	Return transfer function data objects for a system
timebase(sys[, strict])	Return the timebase for an LTI system
timebaseEqual(sys1, sys2)	Check to see if two systems have the same timebase
unwrap(angle[, period])	Unwrap a phase angle to give a continuous curve
use_fbs_defaults()	Use Feedback Systems (FBS) compatible settings.
use_matlab_defaults()	Use MATLAB compatible configuration settings.
use_numpy_matrix([flag, warn])	Turn on/off use of Numpy matrix class for state space
	operations.

## 3.11.1 control.augw

control.augw (g, w1=None, w2=None, w3=None)Augment plant for mixed sensitivity problem.

### **Parameters**

- g(LTI object, ny-by-nu)-
- w1 (weighting on S; None, scalar, or k1-by-ny LTI object) -
- w2 (weighting on KS; None, scalar, or k2-by-nu LTI object)-
- w3 (weighting on T; None, scalar, or k3-by-ny LTI object)-
- p(augmented plant; StateSpace object) -
- a weighting is None (If) -
- augmentation is done for it. At least (no) -
- weighting must not be None. (one)-
- a weighting w is scalar (If) -
- will be replaced by I\*w(it)-
- I is (where) -
- for w1 and w3 (ny-by-ny) -
- nu-by-nu for w2. (and) -

## Returns p

**Return type** plant augmented with weightings, suitable for submission to hinfsyn or h2syn.

### Raises ValueError -

• if all weightings are None

### See also:

h2syn, hinfsyn, mixsyn

## 3.11.2 control.canonical\_form

```
control.canonical form (xsys, form='reachable')
```

Convert a system into canonical form

#### **Parameters**

- xsys (StateSpace object) System to be transformed, with state 'x'
- form (String) -

### Canonical form for transformation. Chosen from:

- 'reachable' reachable canonical form
- 'observable' observable canonical form
- 'modal' modal canonical form

### Returns

- zsys (StateSpace object) System in desired canonical form, with state 'z'
- T(matrix) Coordinate transformation matrix, z = T \* x

## 3.11.3 control.damp

```
control.damp(sys, doprint=True)
```

Compute natural frequency, damping ratio, and poles of a system

The function takes 1 or 2 parameters

### **Parameters**

- sys (LTI (StateSpace or TransferFunction)) A linear system object
- **doprint** if true, print table with values

### Returns

- wn (array) Natural frequencies of the poles
- damping (array) Damping values
- **poles** (*array*) Pole locations
- Algorithm
- \_\_\_\_
- If the system is continuous, -wn = abs(poles) Z = -real(poles)/poles.
- If the system is discrete, the discrete poles are mapped to their
- equivalent location in the s-plane via s = log 10(poles)/dt
- and wn = abs(s) Z = -real(s)/wn.

### See also:

pole

## 3.11.4 control.db2mag

```
control.db2mag(db)
```

Convert a gain in decibels (dB) to a magnitude

If A is magnitude,

```
db = 20 * log10(A)
```

Parameters db (float or ndarray) - input value or array of values, given in decibels

Returns mag – corresponding magnitudes

Return type float or ndarray

## 3.11.5 control.isctime

```
control.isctime (sys, strict=False)
```

Check to see if a system is a continuous-time system

### **Parameters**

- **sys** (LTI system) System to be checked
- **strict** (bool (default = False)) If strict is True, make sure that timebase is not None

## 3.11.6 control.isdtime

```
control.isdtime (sys, strict=False)
```

Check to see if a system is a discrete time system

### **Parameters**

- **sys** (LTI system) System to be checked
- **strict** (bool (default = False)) If strict is True, make sure that timebase is not None

## 3.11.7 control.issiso

```
control.issiso(sys, strict=False)
```

Check to see if a system is single input, single output

#### **Parameters**

- **sys** (LTI system) System to be checked
- strict (bool (default = False)) If strict is True, do not treat scalars as SISO

## 3.11.8 control.issys

```
control.issys(obj)
```

Return True if an object is a system, otherwise False

## 3.11.9 control.mag2db

```
control.mag2db (mag)
```

Convert a magnitude to decibels (dB)

If A is magnitude,

```
db = 20 * log10(A)
```

Parameters mag (float or ndarray) - input magnitude or array of magnitudes

**Returns db** – corresponding values in decibels

Return type float or ndarray

## 3.11.10 control.observable\_form

```
control.observable_form(xsys)
```

Convert a system into observable canonical form

**Parameters** xsys (StateSpace object) – System to be transformed, with state x

### Returns

- zsys (StateSpace object) System in observable canonical form, with state z
- T(matrix) Coordinate transformation: z = T \* x

## 3.11.11 control.pade

```
control.pade(T, n=1, numdeg=None)
```

Create a linear system that approximates a delay.

Return the numerator and denominator coefficients of the Pade approximation.

### **Parameters**

- **T** (number) time delay
- n (positive integer) degree of denominator of approximation
- numdeg (integer, or None (the default)) If None, numerator degree equals denominator degree If >= 0, specifies degree of numerator If < 0, numerator degree is n+numdeg

**Returns num, den** – Polynomial coefficients of the delay model, in descending powers of s.

Return type array

#### Based on:

- 1. Algorithm 11.3.1 in Golub and van Loan, "Matrix Computation" 3rd. Ed. pp. 572-574
- 2. M. Vajta, "Some remarks on Padé-approximations", 3rd TEMPUS-INTCOM Symposium

## 3.11.12 control.reachable\_form

```
control.reachable_form(xsys)
```

Convert a system into reachable canonical form

**Parameters** xsys (StateSpace object) – System to be transformed, with state x

#### Returns

- zsys (StateSpace object) System in reachable canonical form, with state z
- T(matrix) Coordinate transformation: z = T \* x

## 3.11.13 control.sample\_system

```
control.sample_system (sysc, Ts, method='zoh', alpha=None, prewarp_frequency=None) Convert a continuous time system to discrete time
```

Creates a discrete time system from a continuous time system by sampling. Multiple methods of conversion are supported.

### **Parameters**

- **sysc** (linsys) Continuous time system to be converted
- **Ts** (real) Sampling period
- method (string) Method to use for conversion: 'matched', 'tustin', 'zoh' (default)
- **prewarp\_frequency** (*float within [0, infinity)*) The frequency [rad/s] at which to match with the input continuous- time system's magnitude and phase

Returns sysd – Discrete time system, with sampling rate Ts

Return type linsys

## **Notes**

See *TransferFunction.sample* and *StateSpace.sample* for further details.

## **Examples**

```
>>> sysc = TransferFunction([1], [1, 2, 1])
>>> sysd = sample_system(sysc, 1, method='matched')
```

### 3.11.14 control.ss2tf

```
control.ss2tf(sys)
```

Transform a state space system to a transfer function.

The function accepts either 1 or 4 parameters:

**ss2tf(sys)** Convert a linear system into space system form. Always creates a new system, even if sys is already a StateSpace object.

ss2tf(A, B, C, D) Create a state space system from the matrices of its state and output equations.

For details see: ss()

#### **Parameters**

- sys (StateSpace) A linear system
- A (array\_like or string) System matrix
- B (array\_like or string) Control matrix
- C (array\_like or string) Output matrix
- D (array\_like or string) Feedthrough matrix

**Returns out** – New linear system in transfer function form

Return type TransferFunction

#### Raises

- **ValueError** if matrix sizes are not self-consistent, or if an invalid number of arguments is passed in
- **TypeError** if sys is not a StateSpace object

### See also:

```
tf, ss, tf2ss
```

## **Examples**

```
>>> A = [[1., -2], [3, -4]]

>>> B = [[5.], [7]]

>>> C = [[6., 8]]

>>> D = [[9.]]

>>> sys1 = ss2tf(A, B, C, D)
```

```
>>> sys_ss = ss(A, B, C, D)
>>> sys2 = ss2tf(sys_ss)
```

## 3.11.15 control.ssdata

```
control.ssdata(sys)
```

Return state space data objects for a system

Parameters sys (LTI (StateSpace, or TransferFunction)) - LTI system whose data will be returned

Returns (A, B, C, D) – State space data for the system

Return type list of matrices

## 3.11.16 control.tf2ss

```
control.tf2ss(sys)
```

Transform a transfer function to a state space system.

The function accepts either 1 or 2 parameters:

- **tf2ss(sys)** Convert a linear system into transfer function form. Always creates a new system, even if sys is already a TransferFunction object.
- **tf2ss(num, den)** Create a transfer function system from its numerator and denominator polynomial coefficients.

For details see: tf()

#### **Parameters**

- sys(LTI (StateSpace or TransferFunction)) A linear system
- num (array\_like, or list of list of array\_like) Polynomial coefficients of the numerator
- den (array\_like, or list of list of array\_like) Polynomial coefficients of the denominator

**Returns out** – New linear system in state space form

Return type StateSpace

### Raises

- ValueError if num and den have invalid or unequal dimensions, or if an invalid number
  of arguments is passed in
- **TypeError** if *num* or *den* are of incorrect type, or if sys is not a TransferFunction object

### See also:

ss, tf, ss2tf

## **Examples**

```
>>> num = [[[1., 2.], [3., 4.]], [[5., 6.], [7., 8.]]]
>>> den = [[[9., 8., 7.], [6., 5., 4.]], [[3., 2., 1.], [-1., -2., -3.]]]
>>> sys1 = tf2ss(num, den)
```

```
>>> sys_tf = tf(num, den)
>>> sys2 = tf2ss(sys_tf)
```

## 3.11.17 control.tfdata

```
control.tfdata(sys)
```

Return transfer function data objects for a system

Parameters sys (LTI (StateSpace, or TransferFunction)) - LTI system whose data will be returned

**Returns** (num, den) – Transfer function coefficients (SISO only)

Return type numerator and denominator arrays

## 3.11.18 control.timebase

```
control.timebase(sys, strict=True)
```

Return the timebase for an LTI system

dt = timebase(sys)

returns the timebase for a system 'sys'. If the strict option is set to False, dt = True will be returned as 1.

## 3.11.19 control.timebaseEqual

```
control.timebaseEqual (sys1, sys2)
```

Check to see if two systems have the same timebase

timebaseEqual(sys1, sys2)

returns True if the timebases for the two systems are compatible. By default, systems with timebase 'None' are compatible with either discrete or continuous timebase systems. If two systems have a discrete timebase (dt > 0) then their timebases must be equal.

## 3.11.20 control.unwrap

```
control.unwrap(angle, period=6.283185307179586)
```

Unwrap a phase angle to give a continuous curve

### **Parameters**

- angle (array\_like) Array of angles to be unwrapped
- **period** (float, optional) Period (defaults to 2\*pi)

Returns angle\_out - Output array, with jumps of period/2 eliminated

Return type array\_like

## **Examples**

```
>>> import numpy as np
>>> theta = [5.74, 5.97, 6.19, 0.13, 0.35, 0.57]
>>> unwrap(theta, period=2 * np.pi)
[5.74, 5.97, 6.19, 6.413185307179586, 6.633185307179586, 6.8531853071795865]
```

## CONTROL SYSTEM CLASSES

The classes listed below are used to represent models of linear time-invariant (LTI) systems. They are usually created from factory functions such as tf() and ss(), so the user should normally not need to instantiate these directly.

TransferFunction(num, den[, dt])	A class for representing transfer functions
StateSpace(A, B, C, D[, dt])	A class for representing state-space models
FrequencyResponseData(d, w)	A class for models defined by frequency response data (FRD)
<pre>InputOutputSystem([inputs, outputs, states,])</pre>	A class for representing input/output systems.

## 4.1 control.TransferFunction

class control. TransferFunction (num, den[, dt])

A class for representing transfer functions

The TransferFunction class is used to represent systems in transfer function form.

The main data members are 'num' and 'den', which are 2-D lists of arrays containing MIMO numerator and denominator coefficients. For example,

```
>>> num[2][5] = numpy.array([1., 4., 8.])
```

means that the numerator of the transfer function from the 6th input to the 3rd output is set to  $s^2 + 4s + 8$ .

Discrete-time transfer functions are implemented by using the 'dt' instance variable and setting it to something other than 'None'. If 'dt' has a non-zero value, then it must match whenever two transfer functions are combined. If 'dt' is set to True, the system will be treated as a discrete time system with unspecified sampling time. The default value of 'dt' is None and can be changed by changing the value of control.config.defaults['xferfcn.default\_dt'].

The TransferFunction class defines two constants s and z that represent the differentiation and delay operators in continuous and discrete time. These can be used to create variables that allow algebraic creation of transfer functions. For example,

```
>>> s = TransferFunction.s
>>> G = (s + 1)/(s**2 + 2*s + 1)
```

```
___init___(*args)
```

TransferFunction(num, den[, dt])

Construct a transfer function.

The default constructor is TransferFunction(num, den), where num and den are lists of lists of arrays containing polynomial coefficients. To create a discrete time transfer funtion, use TransferFunction(num, den, dt) where 'dt' is the sampling time (or True for unspecified sampling time). To call the copy constructor, call TransferFunction(sys), where sys is a TransferFunction object (continuous or discrete).

### **Methods**

TransferFunction(num, den[, dt])
Natural frequency, damping ratio of system poles
Return the zero-frequency (or DC) gain
Evaluate a transfer function at a single angular fre-
quency.
Feedback interconnection between two LTI objects.
Evaluate the transfer function at a list of angular fre-
quencies.
Evaluate the systems's transfer function for a com-
plex variable
returns True if and only if all of the numerator and
denominator polynomials of the (possibly MIMO)
transfer function are zeroth order, that is, if the sys-
tem has no dynamics.
Check to see if a system is a continuous-time system
Check to see if a system is a discrete-time system
Check to see if a system is single input, single output
Remove cancelling pole/zero pairs from a transfer
function
Compute the poles of a transfer function.
Return a list of a list of scipy.signal.lti ob-
jects.
Convert a continuous-time system to discrete time
Compute the zeros of a transfer function.

### **Attributes**

S		
Z		

#### damp()

Natural frequency, damping ratio of system poles

## Returns

- wn (array) Natural frequencies for each system pole
- zeta (array) Damping ratio for each system pole
- poles (array) Array of system poles

## dcgain()

Return the zero-frequency (or DC) gain

For a continous-time transfer function G(s), the DC gain is G(0) For a discrete-time transfer function G(z), the DC gain is G(1)

**Returns** gain – The zero-frequency gain

Return type ndarray

### evalfr(omega)

Evaluate a transfer function at a single angular frequency.

self.\_evalfr(omega) returns the value of the transfer function matrix with input value s = i \* omega.

### feedback (other=1, sign=- 1)

Feedback interconnection between two LTI objects.

### freqresp(omega)

Evaluate the transfer function at a list of angular frequencies.

Reports the frequency response of the system,

```
G(j*omega) = mag*exp(j*phase)
```

for continuous time. For discrete time systems, the response is evaluated around the unit circle such that

 $G(\exp(j*\circ mega*dt)) = mag*\exp(j*phase).$ 

**Parameters omega** (array\_like) – A list of frequencies in radians/sec at which the system should be evaluated. The list can be either a python list or a numpy array and will be sorted before evaluation.

### Returns

- mag ((self.outputs, self.inputs, len(omega)) ndarray) The magnitude (absolute value, not dB or log10) of the system frequency response.
- **phase** ((*self.outputs*, *self.inputs*, *len(omega)*) *ndarray*) The wrapped phase in radians of the system frequency response.
- omega (ndarray or list or tuple) The list of sorted frequencies at which the response was
  evaluated.

#### horner(s)

Evaluate the systems's transfer function for a complex variable

Returns a matrix of values evaluated at complex variable s.

### is\_static\_gain()

returns True if and only if all of the numerator and denominator polynomials of the (possibly MIMO) transfer function are zeroth order, that is, if the system has no dynamics.

### isctime (strict=False)

Check to see if a system is a continuous-time system

### **Parameters**

- **sys** (LTI system) System to be checked
- **strict** (bool, optional) If strict is True, make sure that timebase is not None. Default is False.

#### isdtime (strict=False)

Check to see if a system is a discrete-time system

**Parameters strict** (bool, optional) – If strict is True, make sure that timebase is not None. Default is False.

#### issiso()

Check to see if a system is single input, single output

```
minreal (tol=None)
```

Remove cancelling pole/zero pairs from a transfer function

### pole()

Compute the poles of a transfer function.

### returnScipySignalLTI()

Return a list of a list of scipy.signal.lti objects.

For instance.

```
>>> out = tfobject.returnScipySignalLTI()
>>> out[3][5]
```

is a class: scipy.signal.lti object corresponding to the transfer function from the 6th input to the 4th output.

```
sample (Ts, method='zoh', alpha=None, prewarp_frequency=None)
```

Convert a continuous-time system to discrete time

Creates a discrete-time system from a continuous-time system by sampling. Multiple methods of conversion are supported.

### **Parameters**

- Ts (float) Sampling period
- method ({ "gbt ", "bilinear", "euler", "backward\_diff",) "zoh", "matched"} Method to use for sampling:
  - gbt: generalized bilinear transformation
  - bilinear: Tustin's approximation ("gbt" with alpha=0.5)
  - euler: Euler (or forward difference) method ("gbt" with alpha=0)
  - backward\_diff: Backwards difference ("gbt" with alpha=1.0)
  - zoh: zero-order hold (default)
- alpha (float within [0, 1]) The generalized bilinear transformation weighting parameter, which should only be specified with method="gbt", and is ignored otherwise.
- **prewarp\_frequency** (float within [0, infinity)) The frequency [rad/s] at which to match with the input continuous- time system's magnitude and phase (the gain=1 crossover frequency, for example). Should only be specified with method='bilinear' or 'gbt' with alpha=0.5 and ignored otherwise.

**Returns** sysd – Discrete time system, with sampling rate Ts

Return type StateSpace system

### **Notes**

- 1. Available only for SISO systems
- 2. Uses scipy.signal.cont2discrete()

## **Examples**

```
>>> sys = TransferFunction(1, [1,1])
>>> sysd = sys.sample(0.5, method='bilinear')
```

zero()

Compute the zeros of a transfer function.

# 4.2 control.StateSpace

```
class control. StateSpace (A, B, C, D[, dt])
```

A class for representing state-space models

The StateSpace class is used to represent state-space realizations of linear time-invariant (LTI) systems:

$$dx/dt = A x + B u y = C x + D u$$

where u is the input, y is the output, and x is the state.

The main data members are the A, B, C, and D matrices. The class also keeps track of the number of states (i.e., the size of A). The data format used to store state space matrices is set using the value of *config.defaults['use\_numpy\_matrix']*. If True (default), the state space elements are stored as *numpy.matrix* objects; otherwise they are *numpy.ndarray* objects. The <code>use\_numpy\_matrix()</code> function can be used to set the storage type.

Discrete-time state space system are implemented by using the 'dt' instance variable and setting it to the sampling period. If 'dt' is not None, then it must match whenever two state space systems are combined. Setting dt = 0 specifies a continuous system, while leaving dt = None means the system timebase is not specified. If 'dt' is set to True, the system will be treated as a discrete time system with unspecified sampling time. The default value of 'dt' is None and can be changed by changing the value of control.config.defaults['statesp.default dt'].

```
__init__ (*args, **kw)
StateSpace(A, B, C, D[, dt])
```

Construct a state space object.

The default constructor is StateSpace(A, B, C, D), where A, B, C, D are matrices or equivalent objects. To create a discrete time system, use StateSpace(A, B, C, D, dt) where 'dt' is the sampling time (or True for unspecified sampling time). To call the copy constructor, call StateSpace(sys), where sys is a StateSpace object.

## **Methods**

init(*args, **kw)	StateSpace(A, B, C, D[, dt])
append(other)	Append a second model to the present model.
damp()	Natural frequency, damping ratio of system poles
dcgain()	Return the zero-frequency gain
evalfr(omega)	Evaluate a SS system's transfer function at a single
	frequency.
feedback([other, sign])	Feedback interconnection between two LTI systems.
freqresp(omega)	Evaluate the system's transfer function at a list of fre-
	quencies

continues on next page

Table 4 – continued from previous page

horner(s)	Evaluate the systems's transfer function for a com-
	plex variable
is_static_gain()	True if and only if the system has no dynamics, that
	is, if A and B are zero.
isctime([strict])	Check to see if a system is a continuous-time system
isdtime([strict])	Check to see if a system is a discrete-time system
issiso()	Check to see if a system is single input, single output
1ft(other[, nu, ny])	Return the Linear Fractional Transformation.
minreal([tol])	Calculate a minimal realization, removes unobserv-
	able and uncontrollable states
pole()	Compute the poles of a state space system.
returnScipySignalLTI()	Return a list of a list of scipy.signal.lti ob-
	jects.
sample(Ts[, method, alpha, prewarp_frequency])	Convert a continuous time system to discrete time
zero()	Compute the zeros of a state space system.

#### append (other)

Append a second model to the present model.

The second model is converted to state-space if necessary, inputs and outputs are appended and their order is preserved

### damp()

Natural frequency, damping ratio of system poles

#### Returns

- wn (array) Natural frequencies for each system pole
- zeta (array) Damping ratio for each system pole
- poles (array) Array of system poles

## dcgain()

Return the zero-frequency gain

The zero-frequency gain of a continuous-time state-space system is given by:

and of a discrete-time state-space system by:

**Returns** gain – An array of shape (outputs,inputs); the array will either be the zero-frequency (or DC) gain, or, if the frequency response is singular, the array will be filled with np.nan.

### Return type ndarray

### evalfr(omega)

Evaluate a SS system's transfer function at a single frequency.

self.\_evalfr(omega) returns the value of the transfer function matrix with input value s = i \* omega.

### feedback (other=1, sign=- 1)

Feedback interconnection between two LTI systems.

### freqresp(omega)

Evaluate the system's transfer function at a list of frequencies

Reports the frequency response of the system,

$$G(j*omega) = mag*exp(j*phase)$$

for continuous time. For discrete time systems, the response is evaluated around the unit circle such that

 $G(\exp(j*omega*dt)) = mag*exp(j*phase).$ 

**Parameters omega** (array\_like) – A list of frequencies in radians/sec at which the system should be evaluated. The list can be either a python list or a numpy array and will be sorted before evaluation.

#### Returns

- mag ((self.outputs, self.inputs, len(omega)) ndarray) The magnitude (absolute value, not dB or log10) of the system frequency response.
- **phase** ((self.outputs, self.inputs, len(omega)) ndarray) The wrapped phase in radians of the system frequency response.
- omega (ndarray) The list of sorted frequencies at which the response was evaluated.

## horner(s)

Evaluate the systems's transfer function for a complex variable

Returns a matrix of values evaluated at complex variable s.

### is\_static\_gain()

True if and only if the system has no dynamics, that is, if A and B are zero.

### isctime (strict=False)

Check to see if a system is a continuous-time system

#### **Parameters**

- **sys** (LTI system) System to be checked
- **strict** (bool, optional) If strict is True, make sure that timebase is not None. Default is False.

### isdtime (strict=False)

Check to see if a system is a discrete-time system

**Parameters strict** (bool, optional) – If strict is True, make sure that timebase is not None. Default is False.

### issiso()

Check to see if a system is single input, single output

## **lft** (*other*, *nu=-1*, *ny=-1*)

Return the Linear Fractional Transformation.

A definition of the LFT operator can be found in Appendix A.7, page 512 in the 2nd Edition, Multivariable Feedback Control by Sigurd Skogestad.

An alternative definition can be found here: https://www.mathworks.com/help/control/ref/lft.html

### **Parameters**

- other (LTI) The lower LTI system
- ny (int, optional) Dimension of (plant) measurement output.
- nu (int, optional) Dimension of (plant) control input.

## minreal(tol=0.0)

Calculate a minimal realization, removes unobservable and uncontrollable states

### pole()

Compute the poles of a state space system.

### returnScipySignalLTI()

Return a list of a list of scipy.signal.lti objects.

For instance,

```
>>> out = ssobject.returnScipySignalLTI()
>>> out[3][5]
```

is a scipy.signal.lti object corresponding to the transfer function from the 6th input to the 4th output.

```
sample (Ts, method='zoh', alpha=None, prewarp_frequency=None)
```

Convert a continuous time system to discrete time

Creates a discrete-time system from a continuous-time system by sampling. Multiple methods of conversion are supported.

### **Parameters**

- **Ts** (*float*) Sampling period
- method ({ "gbt", "bilinear", "euler", "backward\_diff", "zoh"})
  - Which method to use:
  - gbt: generalized bilinear transformation
  - bilinear: Tustin's approximation ("gbt" with alpha=0.5)
  - euler: Euler (or forward differencing) method ("gbt" with alpha=0)
  - backward\_diff: Backwards differencing ("gbt" with alpha=1.0)
  - zoh: zero-order hold (default)
- alpha (float within [0, 1]) The generalized bilinear transformation weighting parameter, which should only be specified with method="gbt", and is ignored otherwise
- **prewarp\_frequency** (float within [0, infinity)) The frequency [rad/s] at which to match with the input continuous- time system's magnitude and phase (the gain=1 crossover frequency, for example). Should only be specified with method='bilinear' or 'gbt' with alpha=0.5 and ignored otherwise.

**Returns** sysd – Discrete time system, with sampling rate Ts

Return type StateSpace

### **Notes**

Uses scipy.signal.cont2discrete()

## **Examples**

```
>>> sys = StateSpace(0, 1, 1, 0)
>>> sysd = sys.sample(0.5, method='bilinear')
```

zero()

Compute the zeros of a state space system.

# 4.3 control.FrequencyResponseData

```
class control.FrequencyResponseData (d, w)
```

A class for models defined by frequency response data (FRD)

The FrequencyResponseData (FRD) class is used to represent systems in frequency response data form.

The main data members are 'omega' and 'fresp', where *omega* is a 1D array with the frequency points of the response, and *fresp* is a 3D array, with the first dimension corresponding to the output index of the FRD, the second dimension corresponding to the input index, and the 3rd dimension corresponding to the frequency points in omega. For example,

```
>>> frdata[2,5,:] = numpy.array([1., 0.8-0.2j, 0.2-0.8j])
```

means that the frequency response from the 6th input to the 3rd output at the frequencies defined in omega is set to the array above, i.e. the rows represent the outputs and the columns represent the inputs.

```
__init__ (*args, **kwargs)
Construct an FRD object.
```

The default constructor is FRD(d, w), where w is an iterable of frequency points, and d is the matching frequency data.

If d is a single list, 1d array, or tuple, a SISO system description is assumed. d can also be

To call the copy constructor, call FRD(sys), where sys is a FRD object.

To construct frequency response data for an existing LTI object, other than an FRD, call FRD(sys, omega)

## Methods

init(*args, **kwargs)	Construct an FRD object.
damp()	Natural frequency, damping ratio of system poles
dcgain()	Return the zero-frequency gain
eval(omega)	Evaluate a transfer function at a single angular fre-
	quency.
evalfr(omega)	Evaluate a transfer function at a single angular fre-
	quency.
feedback([other, sign])	Feedback interconnection between two FRD objects.
freqresp(omega)	Evaluate the frequency response at a list of angular
	frequencies.
isctime([strict])	Check to see if a system is a continuous-time system
isdtime([strict])	Check to see if a system is a discrete-time system
issiso()	Check to see if a system is single input, single output

### **Attributes**

epsw

#### damp()

Natural frequency, damping ratio of system poles

### **Returns**

- wn (array) Natural frequencies for each system pole
- zeta (array) Damping ratio for each system pole
- **poles** (*array*) Array of system poles

### dcgain()

Return the zero-frequency gain

#### eval (omega)

Evaluate a transfer function at a single angular frequency.

self.evalfr(omega) returns the value of the frequency response at frequency omega.

Note that a "normal" FRD only returns values for which there is an entry in the omega vector. An interpolating FRD can return intermediate values.

### evalfr(omega)

Evaluate a transfer function at a single angular frequency.

self.\_evalfr(omega) returns the value of the frequency response at frequency omega.

Note that a "normal" FRD only returns values for which there is an entry in the omega vector. An interpolating FRD can return intermediate values.

### **feedback** (other=1, sign=-1)

Feedback interconnection between two FRD objects.

## freqresp(omega)

Evaluate the frequency response at a list of angular frequencies.

Reports the value of the magnitude, phase, and angular frequency of the requency response evaluated at omega, where omega is a list of angular frequencies, and is a sorted version of the input omega.

**Parameters** omega (array\_like) – A list of frequencies in radians/sec at which the system should be evaluated. The list can be either a python list or a numpy array and will be sorted before evaluation.

#### Returns

- mag ((self.outputs, self.inputs, len(omega)) ndarray) The magnitude (absolute value, not dB or log10) of the system frequency response.
- **phase** ((*self.outputs*, *self.inputs*, *len(omega)*) *ndarray*) The wrapped phase in radians of the system frequency response.
- **omega** (*ndarray or list or tuple*) The list of sorted frequencies at which the response was evaluated.

## isctime (strict=False)

Check to see if a system is a continuous-time system

#### **Parameters**

• **sys** (LTI system) - System to be checked

• **strict** (bool, optional) - If strict is True, make sure that timebase is not None. Default is False.

#### isdtime (strict=False)

Check to see if a system is a discrete-time system

**Parameters strict** (bool, optional) – If strict is True, make sure that timebase is not None. Default is False.

#### issiso()

Check to see if a system is single input, single output

# 4.4 control.iosys.InputOutputSystem

class control.iosys.InputOutputSystem(inputs=None, outputs=None, params={}, dt=None, name=None)
states=None,

A class for representing input/output systems.

The InputOutputSystem class allows (possibly nonlinear) input/output systems to be represented in Python. It is intended as a parent class for a set of subclasses that are used to implement specific structures and operations for different types of input/output dynamical systems.

### **Parameters**

- inputs (int, list of str, or None) Description of the system inputs. This can be given as an integer count or as a list of strings that name the individual signals. If an integer count is specified, the names of the signal will be of the form s[i] (where s is one of u, y, or x). If this parameter is not given or given as None, the relevant quantity will be determined when possible based on other information provided to functions using the system.
- outputs (int, list of str, or None) Description of the system outputs. Same format as *inputs*.
- **states** (int, list of str, or None) Description of the system states. Same format as inputs.
- dt (None, True or float, optional) System timebase. None (default) indicates continuous time, True indicates discrete time with undefined sampling time, positive number is discrete time with specified sampling time.
- params (dict, optional) Parameter values for the systems. Passed to the evaluation functions for the system as default values, overriding internal defaults.
- name (string, optional) System name (used for specifying signals). If unspecified, a generic name <sys[id]> is generated with a unique integer id.

### ninputs, noutputs, nstates

Number of input, output and state variables

Type int

## input\_index, output\_index, state\_index

Dictionary of signal names for the inputs, outputs and states and the index of the corresponding array

Type dict

dt

System timebase. None (default) indicates continuous time, True indicates discrete time with undefined sampling time, positive number is discrete time with specified sampling time.

**Type** None, True or float

#### params

Parameter values for the systems. Passed to the evaluation functions for the system as default values, overriding internal defaults.

**Type** dict, optional

#### name

System name (used for specifying signals)

**Type** string, optional

#### **Notes**

The InputOuputSystem class (and its subclasses) makes use of two special methods for implementing much of the work of the class:

- \_rhs(t, x, u): compute the right hand side of the differential or difference equation for the system. This must be specified by the subclass for the system.
- \_out(t, x, u): compute the output for the current state of the system. The default is to return the entire system state.

\_\_init\_\_ (inputs=None, outputs=None, states=None, params={}, dt=None, name=None)
Create an input/output system.

The InputOutputSystem contructor is used to create an input/output object with the core information required for all input/output systems. Instances of this class are normally created by one of the input/output subclasses: LinearIOSystem, NonlinearIOSystem, InterconnectedSystem.

### **Parameters**

- **inputs** (*int*, *list* of *str*, or *None*) Description of the system inputs. This can be given as an integer count or as a list of strings that name the individual signals. If an integer count is specified, the names of the signal will be of the form *s[i]* (where *s* is one of *u*, *y*, or *x*). If this parameter is not given or given as *None*, the relevant quantity will be determined when possible based on other information provided to functions using the system.
- outputs (int, list of str, or None) Description of the system outputs. Same format as *inputs*.
- **states** (int, list of str, or None) Description of the system states. Same format as *inputs*.
- dt (None, True or float, optional) System timebase. None (default) indicates continuous time, True indicates discrete time with undefined sampling time, positive number is discrete time with specified sampling time.
- params (dict, optional) Parameter values for the systems. Passed to the evaluation functions for the system as default values, overriding internal defaults.
- name (string, optional) System name (used for specifying signals). If unspecified, a generic name <sys[id]> is generated with a unique integer id.

**Returns** Input/output system object

Return type InputOutputSystem

## **Methods**

init([inputs, outputs, states, params,])	Create an input/output system.
copy([newname])	Make a copy of an input/output system.
feedback([other, sign, params])	Feedback interconnection between two input/output
	systems
find_input(name)	Find the index for an input given its name ( <i>None</i> if
	not found)
find_output(name)	Find the index for an output given its name ( <i>None</i> if
	not found)
find_state(name)	Find the index for a state given its name ( <i>None</i> if not
	found)
linearize(x0, u0[, t, params, eps])	Linearize an input/output system at a given state and
	input.
name_or_default([name])	
<pre>set_inputs(inputs[, prefix])</pre>	Set the number/names of the system inputs.
set_outputs(outputs[, prefix])	Set the number/names of the system outputs.
set_states(states[, prefix])	Set the number/names of the system states.

### **Attributes**

### idCounter

### copy (newname=None)

Make a copy of an input/output system.

## feedback (other=1, sign=-1, params={})

Feedback interconnection between two input/output systems

### **Parameters**

- **sys1** (InputOutputSystem) The primary process.
- **sys2** (InputOutputSystem) The feedback process (often a feedback controller).
- **sign** (*scalar*, *optional*) The sign of feedback. *sign* = -1 indicates negative feedback, and *sign* = 1 indicates positive feedback. *sign* is an optional argument; it assumes a value of -1 if not specified.

### Returns out

Return type InputOutputSystem

Raises ValueError – if the inputs, outputs, or timebases of the systems are incompatible.

### find input (name)

Find the index for an input given its name (*None* if not found)

### find\_output (name)

Find the index for an output given its name (*None* if not found)

## find\_state(name)

Find the index for a state given its name (*None* if not found)

### **linearize** (x0, u0, t=0, $params={}$ }, eps=1e-06)

Linearize an input/output system at a given state and input.

Return the linearization of an input/output system at a given state and input value as a StateSpace system. See linearize() for complete documentation.

### set\_inputs (inputs, prefix='u')

Set the number/names of the system inputs.

### **Parameters**

- **inputs** (*int*, *list* of *str*, or *None*) Description of the system inputs. This can be given as an integer count or as a list of strings that name the individual signals. If an integer count is specified, the names of the signal will be of the form *u[i]* (where the prefix *u* can be changed using the optional prefix parameter).
- **prefix** (string, optional) If inputs is an integer, create the names of the states using the given prefix (default = 'u'). The names of the input will be of the form prefix[i].

### set\_outputs (outputs, prefix='y')

Set the number/names of the system outputs.

### **Parameters**

- **outputs** (*int*, *list* of *str*, or *None*) Description of the system outputs. This can be given as an integer count or as a list of strings that name the individual signals. If an integer count is specified, the names of the signal will be of the form u[i] (where the prefix u can be changed using the optional prefix parameter).
- **prefix** (string, optional) If outputs is an integer, create the names of the states using the given prefix (default = 'y'). The names of the input will be of the form prefix[i].

### set\_states (states, prefix='x')

Set the number/names of the system states.

## **Parameters**

- **states** (*int*, *list* of *str*, or *None*) Description of the system states. This can be given as an integer count or as a list of strings that name the individual signals. If an integer count is specified, the names of the signal will be of the form *u[i]* (where the prefix *u* can be changed using the optional prefix parameter).
- **prefix** (string, optional) If states is an integer, create the names of the states using the given prefix (default = 'x'). The names of the input will be of the form prefix[i].

# 4.5 Input/output system subclasses

Input/output systems are accessed primarily via a set of subclasses that allow for linear, nonlinear, and interconnected elements:

LinearIOSystem(linsys[, inputs, outp	outs, ])	Input/output representation of a linear (state space) sys-
		tem.
NonlinearIOSystem(updfcn[, outfcr	n, inputs,])	Nonlinear I/O system.
InterconnectedSystem(syslist[,	connections,	Interconnection of a set of input/output systems.
])		

## 4.5.1 control.iosys.LinearIOSystem

**class** control.iosys.**LinearIOSystem**(linsys, inputs=None, outputs=None, states=None, name=None)

Input/output representation of a linear (state space) system.

This class is used to implementat a system that is a linear state space system (defined by the StateSpace system object).

\_\_init\_\_ (linsys, inputs=None, outputs=None, states=None, name=None)
Create an I/O system from a state space linear system.

Converts a StateSpace system into an InputOutputSystem with the same inputs, outputs, and states. The new system can be a continuous or discrete time system

#### **Parameters**

- linsys (StateSpace) LTI StateSpace system to be converted
- **inputs** (*int*, *list* of *str* or *None*, *optional*) Description of the system inputs. This can be given as an integer count or as a list of strings that name the individual signals. If an integer count is specified, the names of the signal will be of the form *s[i]* (where *s* is one of *u*, *y*, or *x*). If this parameter is not given or given as *None*, the relevant quantity will be determined when possible based on other information provided to functions using the system.
- outputs (int, list of str or None, optional) Description of the system outputs. Same format as *inputs*.
- **states** (int, list of str, or None, optional) Description of the system states. Same format as *inputs*.
- dt (None, True or float, optional) System timebase. None (default) indicates continuous time, True indicates discrete time with undefined sampling time, positive number is discrete time with specified sampling time.
- params (dict, optional) Parameter values for the systems. Passed to the evaluation functions for the system as default values, overriding internal defaults.
- name (string, optional) System name (used for specifying signals). If unspecified, a generic name <sys[id]> is generated with a unique integer id.

**Returns** iosys – Linear system represented as an input/output system

Return type LinearIOSystem

### **Methods**

init(linsys[, inputs, outputs, states, name])	Create an I/O system from a state space linear sys-
(imsyst, inputs, outputs, states, name)	•
	tem.
append(other)	Append a second model to the present model.
copy([newname])	Make a copy of an input/output system.
damp()	Natural frequency, damping ratio of system poles
dcgain()	Return the zero-frequency gain
evalfr(omega)	Evaluate a SS system's transfer function at a single
	frequency.
feedback([other, sign, params])	Feedback interconnection between two input/output
	systems
	continues on next page

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Table 10 – continued from previous page

Table 10 - Continue	u irom previous page
find_input(name)	Find the index for an input given its name ( <i>None</i> if
	not found)
find_output(name)	Find the index for an output given its name ( <i>None</i> if
	not found)
find_state(name)	Find the index for a state given its name ( <i>None</i> if not
	found)
freqresp(omega)	Evaluate the system's transfer function at a list of fre-
	quencies
horner(s)	Evaluate the systems's transfer function for a com-
	plex variable
is_static_gain()	True if and only if the system has no dynamics, that
	is, if A and B are zero.
isctime([strict])	Check to see if a system is a continuous-time system
isdtime([strict])	Check to see if a system is a discrete-time system
issiso()	Check to see if a system is single input, single output
1ft(other[, nu, ny])	Return the Linear Fractional Transformation.
linearize(x0, u0[, t, params, eps])	Linearize an input/output system at a given state and
	input.
minreal([tol])	Calculate a minimal realization, removes unobserv-
	able and uncontrollable states
name_or_default([name])	
pole()	Compute the poles of a state space system.
returnScipySignalLTI()	Return a list of a list of scipy.signal.lti ob-
	jects.
sample(Ts[, method, alpha, prewarp_frequency])	Convert a continuous time system to discrete time
<pre>set_inputs(inputs[, prefix])</pre>	Set the number/names of the system inputs.
set_outputs(outputs[, prefix])	Set the number/names of the system outputs.
set_states(states[, prefix])	Set the number/names of the system states.
zero()	Compute the zeros of a state space system.

## **Attributes**

## idCounter

## append(other)

Append a second model to the present model.

The second model is converted to state-space if necessary, inputs and outputs are appended and their order is preserved

## copy (newname=None)

Make a copy of an input/output system.

### damp()

Natural frequency, damping ratio of system poles

## Returns

- wn (array) Natural frequencies for each system pole
- zeta (array) Damping ratio for each system pole
- **poles** (*array*) Array of system poles

#### dcgain()

Return the zero-frequency gain

The zero-frequency gain of a continuous-time state-space system is given by:

and of a discrete-time state-space system by:

**Returns gain** – An array of shape (outputs,inputs); the array will either be the zero-frequency (or DC) gain, or, if the frequency response is singular, the array will be filled with np.nan.

### Return type ndarray

### evalfr(omega)

Evaluate a SS system's transfer function at a single frequency.

self.\_evalfr(omega) returns the value of the transfer function matrix with input value s = i \* omega.

### feedback (other=1, sign=-1, params={})

Feedback interconnection between two input/output systems

### **Parameters**

- **sys1** (InputOutputSystem) The primary process.
- **sys2** (InputOutputSystem) The feedback process (often a feedback controller).
- **sign** (*scalar*, *optional*) The sign of feedback. *sign* = -1 indicates negative feedback, and *sign* = 1 indicates positive feedback. *sign* is an optional argument; it assumes a value of -1 if not specified.

#### Returns out

Return type InputOutputSystem

Raises ValueError – if the inputs, outputs, or timebases of the systems are incompatible.

### find input (name)

Find the index for an input given its name (*None* if not found)

## find\_output (name)

Find the index for an output given its name (*None* if not found)

#### find\_state(name)

Find the index for a state given its name (*None* if not found)

## freqresp(omega)

Evaluate the system's transfer function at a list of frequencies

Reports the frequency response of the system,

```
G(j*omega) = mag*exp(j*phase)
```

for continuous time. For discrete time systems, the response is evaluated around the unit circle such that

```
G(\exp(j*omega*dt)) = mag*exp(j*phase).
```

**Parameters omega** (array\_like) – A list of frequencies in radians/sec at which the system should be evaluated. The list can be either a python list or a numpy array and will be sorted before evaluation.

## Returns

• mag ((self.outputs, self.inputs, len(omega)) ndarray) – The magnitude (absolute value, not dB or log10) of the system frequency response.

- **phase** ((*self.outputs*, *self.inputs*, *len(omega)*) *ndarray*) The wrapped phase in radians of the system frequency response.
- **omega** (*ndarray*) The list of sorted frequencies at which the response was evaluated.

### horner(s)

Evaluate the systems's transfer function for a complex variable

Returns a matrix of values evaluated at complex variable s.

## is\_static\_gain()

True if and only if the system has no dynamics, that is, if A and B are zero.

### isctime (strict=False)

Check to see if a system is a continuous-time system

#### **Parameters**

- **sys** (LTI system) System to be checked
- **strict** (bool, optional) If strict is True, make sure that timebase is not None. Default is False.

#### isdtime (strict=False)

Check to see if a system is a discrete-time system

**Parameters strict** (bool, optional) – If strict is True, make sure that timebase is not None. Default is False.

#### issiso()

Check to see if a system is single input, single output

### **lft** (other, nu=-1, ny=-1)

Return the Linear Fractional Transformation.

A definition of the LFT operator can be found in Appendix A.7, page 512 in the 2nd Edition, Multivariable Feedback Control by Sigurd Skogestad.

An alternative definition can be found here: https://www.mathworks.com/help/control/ref/lft.html

### **Parameters**

- other (LTI) The lower LTI system
- ny (int, optional) Dimension of (plant) measurement output.
- nu (int, optional) Dimension of (plant) control input.

### **linearize** (x0, u0, t=0, $params={}$ }, eps=1e-06)

Linearize an input/output system at a given state and input.

Return the linearization of an input/output system at a given state and input value as a StateSpace system. See linearize() for complete documentation.

### minreal(tol=0.0)

Calculate a minimal realization, removes unobservable and uncontrollable states

### pole()

Compute the poles of a state space system.

## returnScipySignalLTI()

Return a list of a list of scipy.signal.lti objects.

For instance,

```
>>> out = ssobject.returnScipySignalLTI()
>>> out[3][5]
```

is a scipy.signal.lti object corresponding to the transfer function from the 6th input to the 4th output.

```
sample (Ts, method='zoh', alpha=None, prewarp_frequency=None)
```

Convert a continuous time system to discrete time

Creates a discrete-time system from a continuous-time system by sampling. Multiple methods of conversion are supported.

### **Parameters**

- Ts (float) Sampling period
- method ({"gbt", "bilinear", "euler", "backward\_diff", "zoh"})
  - Which method to use:
  - gbt: generalized bilinear transformation
  - bilinear: Tustin's approximation ("gbt" with alpha=0.5)
  - euler: Euler (or forward differencing) method ("gbt" with alpha=0)
  - backward\_diff: Backwards differencing ("gbt" with alpha=1.0)
  - zoh: zero-order hold (default)
- alpha (float within [0, 1]) The generalized bilinear transformation weighting parameter, which should only be specified with method="gbt", and is ignored otherwise
- **prewarp\_frequency** (float within [0, infinity)) The frequency [rad/s] at which to match with the input continuous- time system's magnitude and phase (the gain=1 crossover frequency, for example). Should only be specified with method='bilinear' or 'gbt' with alpha=0.5 and ignored otherwise.

**Returns** sysd – Discrete time system, with sampling rate Ts

Return type StateSpace

## **Notes**

Uses scipy.signal.cont2discrete()

## **Examples**

```
>>> sys = StateSpace(0, 1, 1, 0)
>>> sysd = sys.sample(0.5, method='bilinear')
```

set inputs (inputs, prefix='u')

Set the number/names of the system inputs.

### **Parameters**

• **inputs** (*int*, *list* of *str*, or *None*) – Description of the system inputs. This can be given as an integer count or as a list of strings that name the individual signals. If an integer count is specified, the names of the signal will be of the form *u[i]* (where the prefix *u* can be changed using the optional prefix parameter).

• **prefix** (*string*, *optional*) – If *inputs* is an integer, create the names of the states using the given prefix (default = 'u'). The names of the input will be of the form *prefix[i]*.

```
set_outputs (outputs, prefix='y')
```

Set the number/names of the system outputs.

#### **Parameters**

- **outputs** (*int*, *list* of *str*, or *None*) Description of the system outputs. This can be given as an integer count or as a list of strings that name the individual signals. If an integer count is specified, the names of the signal will be of the form u[i] (where the prefix u can be changed using the optional prefix parameter).
- **prefix** (string, optional) If outputs is an integer, create the names of the states using the given prefix (default = 'y'). The names of the input will be of the form prefix[i].

```
set_states (states, prefix='x')
```

Set the number/names of the system states.

### **Parameters**

- **states** (*int*, *list* of *str*, or *None*) Description of the system states. This can be given as an integer count or as a list of strings that name the individual signals. If an integer count is specified, the names of the signal will be of the form *u[i]* (where the prefix *u* can be changed using the optional prefix parameter).
- **prefix** (string, optional) If states is an integer, create the names of the states using the given prefix (default = 'x'). The names of the input will be of the form prefix[i].

zero()

Compute the zeros of a state space system.

## 4.5.2 control.iosys.NonlinearIOSystem

Nonlinear I/O system.

This class is used to implement a system that is a nonlinear state space system (defined by and update function and an output function).

```
__init__ (updfcn, outfcn=None, inputs=None, outputs=None, states=None, params={}, dt=None, name=None)
```

Create a nonlinear I/O system given update and output functions.

Creates an *InputOutputSystem* for a nonlinear system by specifying a state update function and an output function. The new system can be a continuous or discrete time system (Note: discrete-time systems not yet supported by most function.)

#### **Parameters**

• updfcn (callable) – Function returning the state update function

```
updfcn(t, x, u[, param]) \rightarrow array
```

where x is a 1-D array with shape (nstates,), u is a 1-D array with shape (ninputs,), t is a float representing the current time, and param is an optional dict containing the values of parameters used by the function.

• outfcn (callable) - Function returning the output at the given state

```
outfcn(t, x, u[, param]) \rightarrow array
```

where the arguments are the same as for *upfcn*.

- **inputs** (*int*, *list* of *str* or *None*, *optional*) Description of the system inputs. This can be given as an integer count or as a list of strings that name the individual signals. If an integer count is specified, the names of the signal will be of the form *s[i]* (where *s* is one of *u*, *y*, or *x*). If this parameter is not given or given as *None*, the relevant quantity will be determined when possible based on other information provided to functions using the system.
- outputs (int, list of str or None, optional) Description of the system outputs. Same format as *inputs*.
- **states** (int, list of str, or None, optional) Description of the system states. Same format as *inputs*.
- params (dict, optional) Parameter values for the systems. Passed to the evaluation functions for the system as default values, overriding internal defaults.
- dt (timebase, optional) The timebase for the system, used to specify whether the system is operating in continuous or discrete time. It can have the following values:
  - dt = None No timebase specified
  - dt = 0 Continuous time system
  - dt > 0 Discrete time system with sampling time dt
  - dt = True Discrete time with unspecified sampling time
- name (string, optional) System name (used for specifying signals). If unspecified, a generic name <sys[id]> is generated with a unique integer id.

**Returns** iosys – Nonlinear system represented as an input/output system.

**Return type** *NonlinearIOSystem* 

### Methods

init(updfcn[, outfcn, inputs, outputs,])	Create a nonlinear I/O system given update and out-
	put functions.
copy([newname])	Make a copy of an input/output system.
feedback([other, sign, params])	Feedback interconnection between two input/output
	systems
find_input(name)	Find the index for an input given its name ( <i>None</i> if
	not found)
find_output(name)	Find the index for an output given its name ( <i>None</i> if
	not found)
find_state(name)	Find the index for a state given its name ( <i>None</i> if not
	found)
linearize(x0, u0[, t, params, eps])	Linearize an input/output system at a given state and
	input.
name_or_default([name])	
set_inputs(inputs[, prefix])	Set the number/names of the system inputs.
set_outputs(outputs[, prefix])	Set the number/names of the system outputs.
set_states(states[, prefix])	Set the number/names of the system states.

### **Attributes**

#### idCounter

### copy (newname=None)

Make a copy of an input/output system.

## feedback (other=1, sign=- 1, params={})

Feedback interconnection between two input/output systems

#### **Parameters**

- **sys1** (InputOutputSystem) The primary process.
- **sys2** (InputOutputSystem) The feedback process (often a feedback controller).
- **sign** (*scalar*, *optional*) The sign of feedback. *sign* = -1 indicates negative feedback, and *sign* = 1 indicates positive feedback. *sign* is an optional argument; it assumes a value of -1 if not specified.

#### Returns out

Return type InputOutputSystem

Raises ValueError – if the inputs, outputs, or timebases of the systems are incompatible.

#### find input(name)

Find the index for an input given its name (*None* if not found)

### find\_output (name)

Find the index for an output given its name (*None* if not found)

#### find state(name)

Find the index for a state given its name (*None* if not found)

## **linearize** (x0, u0, t=0, $params={}$ }, eps=1e-06)

Linearize an input/output system at a given state and input.

Return the linearization of an input/output system at a given state and input value as a StateSpace system. See linearize() for complete documentation.

### set inputs (inputs, prefix='u')

Set the number/names of the system inputs.

## **Parameters**

- **inputs** (*int*, *list* of *str*, or *None*) Description of the system inputs. This can be given as an integer count or as a list of strings that name the individual signals. If an integer count is specified, the names of the signal will be of the form *u[i]* (where the prefix *u* can be changed using the optional prefix parameter).
- **prefix** (string, optional) If inputs is an integer, create the names of the states using the given prefix (default = 'u'). The names of the input will be of the form prefix[i].

## set\_outputs (outputs, prefix='y')

Set the number/names of the system outputs.

#### **Parameters**

• **outputs** (*int*, *list* of *str*, or *None*) – Description of the system outputs. This can be given as an integer count or as a list of strings that name the individual signals. If an integer count is specified, the names of the signal will be of the form *u[i]* (where the prefix *u* can be changed using the optional prefix parameter).

• **prefix** (string, optional) – If outputs is an integer, create the names of the states using the given prefix (default = 'y'). The names of the input will be of the form prefix[i].

```
set_states (states, prefix='x')
```

Set the number/names of the system states.

#### **Parameters**

- **states** (*int*, *list* of *str*, or *None*) Description of the system states. This can be given as an integer count or as a list of strings that name the individual signals. If an integer count is specified, the names of the signal will be of the form *u[i]* (where the prefix *u* can be changed using the optional prefix parameter).
- **prefix** (string, optional) If states is an integer, create the names of the states using the given prefix (default = 'x'). The names of the input will be of the form prefix[i].

## 4.5.3 control.iosys.InterconnectedSystem

Interconnection of a set of input/output systems.

This class is used to implement a system that is an interconnection of input/output systems. The sys consists of a collection of subsystems whose inputs and outputs are connected via a connection map. The overall system inputs and outputs are subsets of the subsystem inputs and outputs.

```
__init__ (syslist, connections=[], inplist=[], outlist=[], inputs=None, outputs=None, states=None, params={}, dt=None, name=None}

Create an I/O system from a list of systems + connection info.
```

The InterconnectedSystem class is used to represent an input/output system that consists of an interconnection between a set of subsystems. The outputs of each subsystem can be summed together to provide inputs to other subsystems. The overall system inputs and outputs can be any subset of subsystem inputs and outputs.

### **Parameters**

- **syslist** (array\_like of InputOutputSystems) The list of input/output systems to be connected
- connections (list of tuple of connection specifications, optional) Description of the internal connections between the subsystems.

```
[connection1, connection2, ...]
```

Each connection is a tuple that describes an input to one of the subsystems. The entries are of the form:

```
(input-spec, output-spec1, output-spec2, ...)
```

The input-spec should be a tuple of the form (subsys\_i, inp\_j) where subsys\_i is the index into syslist and inp\_j is the index into the input vector for the subsystem. If subsys\_i has a single input, then the subsystem index subsys\_i can be listed as the input-spec. If systems and signals are given names, then the form 'sys.sig' or ('sys', 'sig') are also recognized.

Each output-spec should be a tuple of the form (subsys\_i, out\_j, gain). The input will be constructed by summing the listed outputs after multiplying by the gain term. If the gain term is omitted, it is assumed to be 1. If the system has a single output, then the subsystem index subsys\_i can be listed as the input-spec. If systems and signals are given names, then

the form 'sys.sig', ('sys', 'sig') or ('sys', 'sig', gain) are also recognized, and the special form '-sys.sig' can be used to specify a signal with gain -1.

If omitted, the connection map (matrix) can be specified using the set\_connect\_map() method.

• inplist (List of tuple of input specifications, optional)—List of specifications for how the inputs for the overall system are mapped to the subsystem inputs. The input specification is similar to the form defined in the connection specification, except that connections do not specify an input-spec, since these are the system inputs. The entries are thus of the form:

```
(output-spec1, output-spec2, ...)
```

Each system input is added to the input for the listed subsystem.

If omitted, the input map can be specified using the *set\_input\_map* method.

• **outlist** (tuple of output specifications, optional)—List of specifications for how the outputs for the subsystems are mapped to overall system outputs. The output specification is the same as the form defined in the inplist specification (including the optional gain term). Numbered outputs must be chosen from the list of subsystem outputs, but named outputs can also be contained in the list of subsystem inputs.

If omitted, the output map can be specified using the *set\_output\_map* method.

- **inputs** (*int*, *list* of *str* or *None*, *optional*) Description of the system inputs. This can be given as an integer count or as a list of strings that name the individual signals. If an integer count is specified, the names of the signal will be of the form *s[i]* (where *s* is one of *u*, *y*, or *x*). If this parameter is not given or given as *None*, the relevant quantity will be determined when possible based on other information provided to functions using the system.
- **outputs** (*int*, *list* of *str* or *None*, *optional*) Description of the system outputs. Same format as *inputs*.
- **states** (*int*, *list* of *str*, or *None*, *optional*) Description of the system states. Same format as *inputs*, except the state names will be of the form '<subsys\_name>.<state\_name>', for each subsys in syslist and each state\_name of each subsys.
- params (dict, optional) Parameter values for the systems. Passed to the evaluation functions for the system as default values, overriding internal defaults.
- dt (timebase, optional) The timebase for the system, used to specify whether the system is operating in continuous or discrete time. It can have the following values:
  - dt = None No timebase specified
  - dt = 0 Continuous time system
  - dt > 0 Discrete time system with sampling time dt
  - dt = True Discrete time with unspecified sampling time
- name (string, optional) System name (used for specifying signals). If unspecified, a generic name <sys[id]> is generated with a unique integer id.

## **Methods**

init(syslist[, connections, inplist,])	Create an I/O system from a list of systems + con-
	nection info.
copy([newname])	Make a copy of an input/output system.
feedback([other, sign, params])	Feedback interconnection between two input/output systems
find_input(name)	Find the index for an input given its name ( <i>None</i> if not found)
find_output(name)	Find the index for an output given its name ( <i>None</i> if not found)
find_state(name)	Find the index for a state given its name ( <i>None</i> if not found)
linearize(x0, u0[, t, params, eps])	Linearize an input/output system at a given state and input.
<pre>name_or_default([name])</pre>	
set_connect_map(connect_map)	Set the connection map for an interconnected I/O system.
<pre>set_input_map(input_map)</pre>	Set the input map for an interconnected I/O system.
<pre>set_inputs(inputs[, prefix])</pre>	Set the number/names of the system inputs.
set_output_map(output_map)	Set the output map for an interconnected I/O system.
set_outputs(outputs[, prefix])	Set the number/names of the system outputs.
<pre>set_states(states[, prefix])</pre>	Set the number/names of the system states.

## **Attributes**

## idCounter

## copy (newname=None)

Make a copy of an input/output system.

### feedback (other=1, sign=-1, params={})

Feedback interconnection between two input/output systems

### **Parameters**

- **sys1** (InputOutputSystem) The primary process.
- **sys2** (InputOutputSystem) The feedback process (often a feedback controller).
- **sign** (*scalar*, *optional*) The sign of feedback. *sign* = -1 indicates negative feedback, and *sign* = 1 indicates positive feedback. *sign* is an optional argument; it assumes a value of -1 if not specified.

## Returns out

Return type InputOutputSystem

Raises ValueError – if the inputs, outputs, or timebases of the systems are incompatible.

#### find input (name)

Find the index for an input given its name (*None* if not found)

## find\_output (name)

Find the index for an output given its name (*None* if not found)

#### find state(name)

Find the index for a state given its name (*None* if not found)

## linearize (x0, u0, t=0, $params={}\}$ , eps=1e-06)

Linearize an input/output system at a given state and input.

Return the linearization of an input/output system at a given state and input value as a StateSpace system. See linearize() for complete documentation.

#### set connect map(connect map)

Set the connection map for an interconnected I/O system.

**Parameters connect\_map** (2D array) – Specify the matrix that will be used to multiply the vector of subsystem outputs to obtain the vector of subsystem inputs.

### set\_input\_map (input\_map)

Set the input map for an interconnected I/O system.

**Parameters input\_map** (2D array) – Specify the matrix that will be used to multiply the vector of system inputs to obtain the vector of subsystem inputs. These values are added to the inputs specified in the connection map.

### set\_inputs (inputs, prefix='u')

Set the number/names of the system inputs.

### **Parameters**

- **inputs** (*int*, *list* of *str*, or *None*) Description of the system inputs. This can be given as an integer count or as a list of strings that name the individual signals. If an integer count is specified, the names of the signal will be of the form *u[i]* (where the prefix *u* can be changed using the optional prefix parameter).
- **prefix** (string, optional) If inputs is an integer, create the names of the states using the given prefix (default = 'u'). The names of the input will be of the form prefix[i].

## set\_output\_map (output\_map)

Set the output map for an interconnected I/O system.

**Parameters output\_map** (2D array) – Specify the matrix that will be used to multiply the vector of subsystem outputs to obtain the vector of system outputs.

## set\_outputs (outputs, prefix='y')

Set the number/names of the system outputs.

#### **Parameters**

- **outputs** (*int*, *list* of *str*, or *None*) Description of the system outputs. This can be given as an integer count or as a list of strings that name the individual signals. If an integer count is specified, the names of the signal will be of the form u[i] (where the prefix u can be changed using the optional prefix parameter).
- **prefix** (string, optional) If outputs is an integer, create the names of the states using the given prefix (default = 'y'). The names of the input will be of the form prefix[i].

## set\_states (states, prefix='x')

Set the number/names of the system states.

## **Parameters**

• **states** (*int*, *list* of *str*, or *None*) – Description of the system states. This can be given as an integer count or as a list of strings that name the individual signals. If an integer count is specified, the names of the signal will be of the form *u[i]* (where the prefix *u* can be changed using the optional prefix parameter).

• **prefix** (*string*, *optional*) – If *states* is an integer, create the names of the states using the given prefix (default = 'x'). The names of the input will be of the form *prefix[i]*.

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## MATLAB COMPATIBILITY MODULE

The *control.matlab* module contains a number of functions that emulate some of the functionality of MATLAB. The intent of these functions is to provide a simple interface to the python control systems library (python-control) for people who are familiar with the MATLAB Control Systems Toolbox (tm).

# 5.1 Creating linear models

tf(num, den[, dt])	Create a transfer function system.
ss(A, B, C, D[, dt])	Create a state space system.
$frd(\mathbf{d}, \mathbf{w})$	Construct a frequency response data model
rss([states, outputs, inputs])	Create a stable <i>continuous</i> random state space object.
drss([states, outputs, inputs])	Create a stable <i>discrete</i> random state space object.

## 5.1.1 control.matlab.tf

control.matlab.tf(num, den[, dt])

Create a transfer function system. Can create MIMO systems.

The function accepts either 1, 2, or 3 parameters:

- **tf(sys)** Convert a linear system into transfer function form. Always creates a new system, even if sys is already a TransferFunction object.
- tf(num, den) Create a transfer function system from its numerator and denominator polynomial coefficients.

If *num* and *den* are 1D array\_like objects, the function creates a SISO system.

To create a MIMO system, *num* and *den* need to be 2D nested lists of array\_like objects. (A 3 dimensional data structure in total.) (For details see note below.)

- **tf (num, den, dt)** Create a discrete time transfer function system; dt can either be a positive number indicating the sampling time or 'True' if no specific timebase is given.
- tf('s') or tf('z') Create a transfer function representing the differential operator ('s') or delay operator ('z').

### **Parameters**

- sys(LTI (StateSpace or TransferFunction)) A linear system
- num (array\_like, or list of list of array\_like) Polynomial coefficients of the numerator

den (array\_like, or list of list of array\_like) - Polynomial coefficients of the denominator

**Returns out** – The new linear system

Return type TransferFunction

Raises

- **ValueError** if *num* and *den* have invalid or unequal dimensions
- **TypeError** if *num* or *den* are of incorrect type

### See also:

TransferFunction, ss, ss2tf, tf2ss

### **Notes**

num[i][j] contains the polynomial coefficients of the numerator for the transfer function from the (j+1)st input to the (i+1)st output. den[i][j] works the same way.

The list [2, 3, 4] denotes the polynomial  $2s^2 + 3s + 4$ .

The special forms tf('s') and tf('z') can be used to create transfer functions for differentiation and unit delays.

## **Examples**

```
>>> # Create a MIMO transfer function object
>>> # The transfer function from the 2nd input to the 1st output is
>>> # (3s + 4) / (6s^2 + 5s + 4).
>>> num = [[[1., 2.], [3., 4.]], [[5., 6.], [7., 8.]]]
>>> den = [[[9., 8., 7.], [6., 5., 4.]], [[3., 2., 1.], [-1., -2., -3.]]]
>>> sys1 = tf(num, den)
```

```
>>> # Create a variable 's' to allow algebra operations for SISO systems
>>> s = tf('s')
>>> G = (s + 1)/(s**2 + 2*s + 1)
```

```
>>> # Convert a StateSpace to a TransferFunction object.
>>> sys_ss = ss("1. -2; 3. -4", "5.; 7", "6. 8", "9.")
>>> sys2 = tf(sys1)
```

## 5.1.2 control.matlab.ss

```
control.matlab.ss(A, B, C, D[, dt])
```

Create a state space system.

The function accepts either 1, 4 or 5 parameters:

**ss(sys)** Convert a linear system into space system form. Always creates a new system, even if sys is already a StateSpace object.

ss (A, B, C, D) Create a state space system from the matrices of its state and output equations:

$$\dot{x} = A \cdot x + B \cdot u$$
$$u = C \cdot x + D \cdot u$$

ss(A, B, C, D, dt) Create a discrete-time state space system from the matrices of its state and output equations:

$$x[k+1] = A \cdot x[k] + B \cdot u[k]$$
$$y[k] = C \cdot x[k] + D \cdot u[ki]$$

The matrices can be given as *array like* data types or strings. Everything that the constructor of numpy. matrix accepts is permissible here too.

### **Parameters**

- sys (StateSpace or TransferFunction) A linear system
- A (array\_like or string) System matrix
- B (array\_like or string) Control matrix
- C (array\_like or string) Output matrix
- D (array\_like or string) Feed forward matrix
- dt (If present, specifies the sampling period and a discrete time) system is created

Returns out – The new linear system

Return type StateSpace

Raises ValueError – if matrix sizes are not self-consistent

## See also:

StateSpace, tf, ss2tf, tf2ss

## **Examples**

```
>>> # Create a StateSpace object from four "matrices".
>>> sys1 = ss("1. -2; 3. -4", "5.; 7", "6. 8", "9.")
```

```
>>> # Convert a TransferFunction to a StateSpace object.
>>> sys_tf = tf([2.], [1., 3])
>>> sys2 = ss(sys_tf)
```

### 5.1.3 control.matlab.frd

```
control.matlab.frd(d, w)
```

Construct a frequency response data model

frd models store the (measured) frequency response of a system.

This function can be called in different ways:

frd(response, freqs) Create an frd model with the given response data, in the form of complex response vector, at matching frequency freqs [in rad/s]

frd (sys, freqs) Convert an LTI system into an frd model with data at frequencies freqs.

#### **Parameters**

- response (array\_like, or list) complex vector with the system response
- freq(array\_lik or lis) vector with frequencies
- sys(LTI (StateSpace or TransferFunction)) A linear system

**Returns** sys – New frequency response system

Return type FRD

### See also:

FRD, ss, tf

### 5.1.4 control.matlab.rss

```
control.matlab.rss(states=1, outputs=1, inputs=1)
Create a stable continuous random state space object.
```

#### **Parameters**

- states (integer) Number of state variables
- inputs (integer) Number of system inputs
- outputs (integer) Number of system outputs

**Returns** sys – The randomly created linear system

Return type StateSpace

**Raises** ValueError – if any input is not a positive integer

## See also:

drss

## **Notes**

If the number of states, inputs, or outputs is not specified, then the missing numbers are assumed to be 1. The poles of the returned system will always have a negative real part.

## 5.1.5 control.matlab.drss

```
control.matlab.drss (states=1, outputs=1, inputs=1)

Create a stable discrete random state space object.
```

### **Parameters**

- states (integer) Number of state variables
- inputs (integer) Number of system inputs
- outputs (integer) Number of system outputs

**Returns** sys – The randomly created linear system

Return type StateSpace

**Raises** ValueError – if any input is not a positive integer

See also:

rss

### **Notes**

If the number of states, inputs, or outputs is not specified, then the missing numbers are assumed to be 1. The poles of the returned system will always have a magnitude less than 1.

# 5.2 Utility functions and conversions

mag2db(mag)	Convert a magnitude to decibels (dB)
db2mag(db)	Convert a gain in decibels (dB) to a magnitude
c2d(sysc, Ts[, method, prewarp_frequency])	Return a discrete-time system
ss2tf(sys)	Transform a state space system to a transfer function.
tf2ss(sys)	Transform a transfer function to a state space system.
tfdata(sys)	Return transfer function data objects for a system

## 5.2.1 control.matlab.mag2db

```
control.matlab.mag2db (mag)
```

Convert a magnitude to decibels (dB)

If A is magnitude,

$$db = 20 * log10(A)$$

Parameters mag (float or ndarray) - input magnitude or array of magnitudes

**Returns db** – corresponding values in decibels

Return type float or ndarray

## 5.2.2 control.matlab.db2mag

```
control.matlab.db2mag(db)
```

Convert a gain in decibels (dB) to a magnitude

If A is magnitude,

$$db = 20 * log10(A)$$

Parameters db (float or ndarray) - input value or array of values, given in decibels

Returns mag – corresponding magnitudes

Return type float or ndarray

## 5.2.3 control.matlab.c2d

control.matlab.c2d (sysc, Ts, method='zoh', prewarp\_frequency=None)
 Return a discrete-time system

#### **Parameters**

- **sysc** (LTI (StateSpace or TransferFunction), continuous) **System** to be converted
- **Ts** (number) Sample time for the conversion
- method (string, optional) Method to be applied, 'zoh' Zero-order hold on the inputs (default) 'foh' First-order hold, currently not implemented 'impulse' Impulse-invariant discretization, currently not implemented 'tustin' Bilinear (Tustin) approximation, only SISO 'matched' Matched pole-zero method, only SISO
- **prewarp\_frequency** (float within [0, infinity)) The frequency [rad/s] at which to match with the input continuous- time system's magnitude and phase

## 5.2.4 control.matlab.ss2tf

```
control.matlab.ss2tf(sys)
```

Transform a state space system to a transfer function.

The function accepts either 1 or 4 parameters:

**ss2tf(sys)** Convert a linear system into space system form. Always creates a new system, even if sys is already a StateSpace object.

ss2tf(A, B, C, D) Create a state space system from the matrices of its state and output equations.

For details see: ss()

## **Parameters**

- sys (StateSpace) A linear system
- A (array\_like or string) System matrix
- B(array\_like or string) Control matrix
- C (array\_like or string) Output matrix
- D (array\_like or string) Feedthrough matrix

Returns out – New linear system in transfer function form

Return type TransferFunction

#### Raises

- **ValueError** if matrix sizes are not self-consistent, or if an invalid number of arguments is passed in
- **TypeError** if sys is not a StateSpace object

### See also:

tf, ss, tf2ss

## **Examples**

```
>>> A = [[1., -2], [3, -4]]

>>> B = [[5.], [7]]

>>> C = [[6., 8]]

>>> D = [[9.]]

>>> sys1 = ss2tf(A, B, C, D)
```

```
>>> sys_ss = ss(A, B, C, D)
>>> sys2 = ss2tf(sys_ss)
```

## 5.2.5 control.matlab.tf2ss

```
control.matlab.tf2ss(sys)
```

Transform a transfer function to a state space system.

The function accepts either 1 or 2 parameters:

- **tf2ss(sys)** Convert a linear system into transfer function form. Always creates a new system, even if sys is already a TransferFunction object.
- **tf2ss (num, den)** Create a transfer function system from its numerator and denominator polynomial coefficients.

For details see: tf()

#### **Parameters**

- sys (LTI (StateSpace or TransferFunction)) A linear system
- num (array\_like, or list of list of array\_like) Polynomial coefficients of the numerator
- den (array\_like, or list of list of array\_like) Polynomial coefficients of the denominator

Returns out – New linear system in state space form

Return type StateSpace

### Raises

- **ValueError** if *num* and *den* have invalid or unequal dimensions, or if an invalid number of arguments is passed in
- **TypeError** if *num* or *den* are of incorrect type, or if sys is not a TransferFunction object

### See also:

```
ss, tf, ss2tf
```

## **Examples**

```
>>> num = [[[1., 2.], [3., 4.]], [[5., 6.], [7., 8.]]]
>>> den = [[[9., 8., 7.], [6., 5., 4.]], [[3., 2., 1.], [-1., -2., -3.]]]
>>> sys1 = tf2ss(num, den)
```

```
>>> sys_tf = tf(num, den)
>>> sys2 = tf2ss(sys_tf)
```

## 5.2.6 control.matlab.tfdata

```
control.matlab.tfdata(sys)
```

Return transfer function data objects for a system

Parameters sys (LTI (StateSpace, or TransferFunction)) - LTI system whose
 data will be returned

**Returns** (num, den) – Transfer function coefficients (SISO only)

Return type numerator and denominator arrays

# 5.3 System interconnections

series(sys1, *sysn)	Return the series connection (sysn * .
parallel(sys1, *sysn)	Return the parallel connection sys1 + sys2 (+.
feedback(sys1[, sys2, sign])	Feedback interconnection between two I/O systems.
negate(sys)	Return the negative of a system.
connect(sys, Q, inputv, outputv)	Index-based interconnection of an LTI system.
append(sys1, sys2,, sysn)	Group models by appending their inputs and outputs

## 5.3.1 control.matlab.series

```
control.matlab.series (sys1, *sysn)
Return the series connection (sysn * ... *) sys2 * sys1
```

### **Parameters**

- sys1(scalar, StateSpace, TransferFunction, or FRD)-
- \*sysn (other scalars, StateSpaces, TransferFunctions, or FRDs)

## Returns out

Return type scalar, StateSpace, or TransferFunction

**Raises ValueError** – if *sys2.inputs* does not equal *sys1.outputs* if *sys1.dt* is not compatible with *sys2.dt* 

### See also:

parallel, feedback

### **Notes**

This function is a wrapper for the \_\_mul\_\_ function in the StateSpace and TransferFunction classes. The output type is usually the type of *sys2*. If *sys2* is a scalar, then the output type is the type of *sys1*.

If both systems have a defined timebase (dt = 0 for continuous time, dt > 0 for discrete time), then the timebase for both systems must match. If only one of the system has a timebase, the return timebase will be set to match it

## **Examples**

```
>>> sys3 = series(sys1, sys2) # Same as sys3 = sys2 * sys1
>>> sys5 = series(sys1, sys2, sys3, sys4) # More systems
```

## 5.3.2 control.matlab.parallel

```
control.matlab.parallel (sys1, *sysn)
Return the parallel connection sys1 + sys2 (+ ... + sysn)
```

#### **Parameters**

sys1(scalar, StateSpace, TransferFunction, or FRD) \*sysn (other scalars, StateSpaces, TransferFunctions, or FRDs)

### Returns out

**Return type** scalar, *StateSpace*, or *TransferFunction* 

Raises ValueError – if sys1 and sys2 do not have the same numbers of inputs and outputs

### See also:

series, feedback

### **Notes**

This function is a wrapper for the \_\_add\_\_ function in the StateSpace and TransferFunction classes. The output type is usually the type of *sys1*. If *sys1* is a scalar, then the output type is the type of *sys2*.

If both systems have a defined timebase (dt = 0 for continuous time, dt > 0 for discrete time), then the timebase for both systems must match. If only one of the system has a timebase, the return timebase will be set to match it.

## **Examples**

```
>>> sys3 = parallel(sys1, sys2) # Same as sys3 = sys1 + sys2
```

```
>>> sys5 = parallel(sys1, sys2, sys3, sys4) # More systems
```

## 5.3.3 control.matlab.feedback

```
control.matlab.feedback(sys1, sys2=1, sign=-1)
```

Feedback interconnection between two I/O systems.

#### **Parameters**

- **sys1** (*scalar*, StateSpace, TransferFunction, *FRD*) The primary process.
- **sys2** (*scalar*, StateSpace, TransferFunction, *FRD*) The feedback process (often a feedback controller).
- **sign** (*scalar*) The sign of feedback. *sign* = -1 indicates negative feedback, and *sign* = 1 indicates positive feedback. *sign* is an optional argument; it assumes a value of -1 if not specified.

### Returns out

Return type StateSpace or TransferFunction

#### Raises

- **ValueError** if *sys1* does not have as many inputs as *sys2* has outputs, or if *sys2* does not have as many inputs as *sys1* has outputs
- NotImplementedError if an attempt is made to perform a feedback on a MIMO TransferFunction object

### See also:

series, parallel

## **Notes**

This function is a wrapper for the feedback function in the StateSpace and TransferFunction classes. It calls TransferFunction.feedback if *sys1* is a TransferFunction object, and StateSpace.feedback if *sys1* is a StateSpace object. If *sys1* is a scalar, then it is converted to *sys2*'s type, and the corresponding feedback function is used. If *sys1* and *sys2* are both scalars, then TransferFunction.feedback is used.

## 5.3.4 control.matlab.negate

```
control.matlab.negate(sys)

Return the negative of a system.
```

Parameters sys (StateSpace, TransferFunction or FRD) -

Returns out

Return type StateSpace or TransferFunction

### **Notes**

This function is a wrapper for the \_\_neg\_\_ function in the StateSpace and TransferFunction classes. The output type is the same as the input type.

## **Examples**

```
>>> sys2 = negate(sys1) # Same as sys2 = -sys1.
```

### 5.3.5 control.matlab.connect

```
control.matlab.connect (sys, Q, inputv, outputv)
Index-based interconnection of an LTI system.
```

The system sys is a system typically constructed with append, with multiple inputs and outputs. The inputs and outputs are connected according to the interconnection matrix Q, and then the final inputs and outputs are trimmed according to the inputs and outputs listed in inputv and outputv.

NOTE: Inputs and outputs are indexed starting at 1 and negative values correspond to a negative feedback interconnection.

#### **Parameters**

- sys (StateSpace Transferfunction) System to be connected
- Q (2D array) Interconnection matrix. First column gives the input to be connected. The second column gives the index of an output that is to be fed into that input. Each additional column gives the index of an additional input that may be optionally added to that input. Negative values mean the feedback is negative. A zero value is ignored. Inputs and outputs are indexed starting at 1 to communicate sign information.
- inputv (1D array) list of final external inputs, indexed starting at 1
- outputv (1D array) list of final external outputs, indexed starting at 1

**Returns** sys – Connected and trimmed LTI system

**Return type** LTI system

### **Examples**

```
>>> sys1 = ss([[1., -2], [3., -4]], [[5.], [7]], [[6, 8]], [[9.]])
>>> sys2 = ss([[-1.]], [[1.]], [[0.]])
>>> sys = append(sys1, sys2)
>>> Q = [[1, 2], [2, -1]] # negative feedback interconnection
>>> sysc = connect(sys, Q, [2], [1, 2])
```

## 5.3.6 control.matlab.append

```
control.matlab.append(sys1, sys2, ..., sysn)
```

Group models by appending their inputs and outputs

Forms an augmented system model, and appends the inputs and outputs together. The system type will be the type of the first system given; if you mix state-space systems and gain matrices, make sure the gain matrices are not first.

#### **Parameters**

- sys1 (StateSpace or Transferfunction) LTI systems to combine
- sys2 (StateSpace or Transferfunction) LTI systems to combine
- .. (StateSpace or Transferfunction) LTI systems to combine
- sysn (StateSpace or Transferfunction) LTI systems to combine

**Returns** sys – Combined LTI system, with input/output vectors consisting of all input/output vectors appended

Return type LTI system

## **Examples**

```
>>> sys1 = ss([[1., -2], [3., -4]], [[5.], [7]]", [[6., 8]], [[9.]])
>>> sys2 = ss([[-1.]], [[1.]], [[0.]])
>>> sys = append(sys1, sys2)
```

# 5.4 System gain and dynamics

dcgain(*args)	Compute the gain of the system in steady state.
pole(sys)	Compute system poles.
zero(sys)	Compute system zeros.
damp(sys[, doprint])	Compute natural frequency, damping ratio, and poles of
	a system
pzmap(sys[, plot, grid, title])	Plot a pole/zero map for a linear system.

## 5.4.1 control.matlab.dcgain

```
control.matlab.dcgain(*args)
```

Compute the gain of the system in steady state.

The function takes either 1, 2, 3, or 4 parameters:

### **Parameters**

- A (array-like) A linear system in state space form.
- **B** (array-like) A linear system in state space form.
- **C** (array-like) A linear system in state space form.
- **D** (array-like) A linear system in state space form.

- Z (array-like, array-like, number) A linear system in zero, pole, gain form.
- P (array-like, array-like, number) A linear system in zero, pole, gain form.
- k (array-like, array-like, number) A linear system in zero, pole, gain form.
- num (array-like) A linear system in transfer function form.
- den (array-like) A linear system in transfer function form.
- sys(LTI (StateSpace or TransferFunction)) A linear system object.

**Returns** gain – The gain of each output versus each input:  $y = gain \cdot u$ 

Return type ndarray

#### **Notes**

This function is only useful for systems with invertible system matrix A.

All systems are first converted to state space form. The function then computes:

$$qain = -C \cdot A^{-1} \cdot B + D$$

## 5.4.2 control.matlab.pole

control.matlab.pole(sys)

Compute system poles.

Parameters sys (StateSpace or TransferFunction) - Linear system

**Returns** poles – Array that contains the system's poles.

Return type ndarray

Raises NotImplementedError – when called on a TransferFunction object

See also:

zero, TransferFunction.pole, StateSpace.pole

## 5.4.3 control.matlab.zero

control.matlab.zero(sys)

Compute system zeros.

Parameters sys (StateSpace or TransferFunction) - Linear system

**Returns** zeros – Array that contains the system's zeros.

Return type ndarray

Raises NotImplementedError – when called on a MIMO system

See also:

pole, StateSpace.zero, TransferFunction.zero

## 5.4.4 control.matlab.damp

```
control.matlab.damp(sys, doprint=True)
```

Compute natural frequency, damping ratio, and poles of a system

The function takes 1 or 2 parameters

#### **Parameters**

- **sys** (LTI (StateSpace or TransferFunction)) A linear system object
- doprint if true, print table with values

### Returns

- wn (array) Natural frequencies of the poles
- damping (array) Damping values
- poles (array) Pole locations
- Algorithm
- \_\_\_\_
- *If the system is continuous*, wn = abs(poles) Z = -real(poles)/poles.
- If the system is discrete, the discrete poles are mapped to their
- equivalent location in the s-plane via s = log 10(poles)/dt
- and wn = abs(s) Z = -real(s)/wn.

### See also:

pole

## 5.4.5 control.matlab.pzmap

```
control.matlab.pzmap (sys, plot=None, grid=None, title='Pole Zero Map', **kwargs)
Plot a pole/zero map for a linear system.
```

### **Parameters**

- **sys** (*LTI* (StateSpace or TransferFunction)) Linear system for which poles and zeros are computed.
- plot (bool, optional) If True a graph is generated with Matplotlib, otherwise the poles and zeros are only computed and returned.
- grid (boolean (default = False)) If True plot omega-damping grid.

### Returns

- **pole** (*array*) The systems poles
- **zeros** (*array*) The system's zeros.

# 5.5 Time-domain analysis

step(sys[, T, X0, input, output, return_x])	Step response of a linear system
<pre>impulse(sys[, T, X0, input, output, return_x])</pre>	Impulse response of a linear system
<pre>initial(sys[, T, X0, input, output, return_x])</pre>	Initial condition response of a linear system
lsim(sys[, U, T, X0])	Simulate the output of a linear system.

# 5.5.1 control.matlab.step

control.matlab.**step** (*sys*, *T=None*, *X0=0.0*, *input=0*, *output=None*, *return\_x=False*) Step response of a linear system

If the system has multiple inputs or outputs (MIMO), one input has to be selected for the simulation. Optionally, one output may be selected. If no selection is made for the output, all outputs are given. The parameters *input* and *output* do this. All other inputs are set to 0, all other outputs are ignored.

#### **Parameters**

- sys (StateSpace, or TransferFunction) LTI system to simulate
- **T** (array-like or number, optional) Time vector, or simulation time duration if a number (time vector is autocomputed if not given)
- **X0** (array-like or number, optional) Initial condition (default = 0) Numbers are converted to constant arrays with the correct shape.
- input (int) Index of the input that will be used in this simulation.
- **output** (*int*) If given, index of the output that is returned by this simulation.

## Returns

- yout (array) Response of the system
- T (array) Time values of the output
- **xout** (*array* (*if selected*)) Individual response of each x variable

#### See also:

lsim, initial, impulse

## **Examples**

```
>>> yout, T = step(sys, T, X0)
```

# 5.5.2 control.matlab.impulse

```
control.matlab.impulse (sys, T=None, X0=0.0, input=0, output=None, return_x=False) Impulse response of a linear system
```

If the system has multiple inputs or outputs (MIMO), one input has to be selected for the simulation. Optionally, one output may be selected. If no selection is made for the output, all outputs are given. The parameters *input* and *output* do this. All other inputs are set to 0, all other outputs are ignored.

#### **Parameters**

- **sys** (StateSpace, TransferFunction) LTI system to simulate
- **T**(array-like or number, optional) Time vector, or simulation time duration if a number (time vector is autocomputed if not given)
- **x0** (array-like or number, optional) Initial condition (default = 0) Numbers are converted to constant arrays with the correct shape.
- input (int) Index of the input that will be used in this simulation.
- output (int) Index of the output that will be used in this simulation.

#### **Returns**

- yout (array) Response of the system
- T (array) Time values of the output
- **xout** (*array* (*if selected*)) Individual response of each x variable

#### See also:

1sim, step, initial

## **Examples**

```
>>> yout, T = impulse(sys, T)
```

## 5.5.3 control.matlab.initial

```
control.matlab.initial (sys, T=None, X0=0.0, input=None, output=None, return_x=False) Initial condition response of a linear system
```

If the system has multiple outputs (?IMO), optionally, one output may be selected. If no selection is made for the output, all outputs are given.

#### **Parameters**

- sys (StateSpace, or TransferFunction) LTI system to simulate
- **T**(array-like or number, optional) Time vector, or simulation time duration if a number (time vector is autocomputed if not given)
- **X0** (array-like object or number, optional) Initial condition (default = 0)

Numbers are converted to constant arrays with the correct shape.

- input (int) This input is ignored, but present for compatibility with step and impulse.
- **output** (*int*) If given, index of the output that is returned by this simulation.

#### Returns

- yout (array) Response of the system
- T (array) Time values of the output
- xout (array (if selected)) Individual response of each x variable

#### See also:

1sim, step, impulse

#### **Examples**

```
>>> yout, T = initial(sys, T, X0)
```

## 5.5.4 control.matlab.lsim

```
control.matlab.lsim(sys, U=0.0, T=None, X0=0.0)
```

Simulate the output of a linear system.

As a convenience for parameters U, X0: Numbers (scalars) are converted to constant arrays with the correct shape. The correct shape is inferred from arguments sys and T.

## **Parameters**

- sys (LTI (StateSpace, or TransferFunction)) LTI system to simulate
- U (array-like or number, optional) Input array giving input at each time T (default = 0).

If U is None or 0, a special algorithm is used. This special algorithm is faster than the general algorithm, which is used otherwise.

- **T** (array-like, optional for discrete LTI *sys*) Time steps at which the input is defined; values must be evenly spaced.
- **X0** (array-like or number, optional) Initial condition (default = 0).

## Returns

- yout (array) Response of the system.
- **T** (*array*) Time values of the output.
- **xout** (*array*) Time evolution of the state vector.

## See also:

```
step, initial, impulse
```

## **Examples**

```
>>> yout, T, xout = lsim(sys, U, T, X0)
```

# 5.6 Frequency-domain analysis

bode(syslist[, omega, dB, Hz, deg,])	Bode plot of the frequency response
nyquist(syslist[, omega, plot, label_freq,])	Nyquist plot for a system
nichols(sys_list[, omega, grid])	Nichols plot for a system
margin(sysdata)	Calculate gain and phase margins and associated
	crossover frequencies
freqresp(sys, omega)	Frequency response of an LTI system at multiple angu-
	lar frequencies.
evalfr(sys, x)	Evaluate the transfer function of an LTI system for a
	single complex number x.

## 5.6.1 control.matlab.bode

```
control.matlab.bode (syslist[, omega, dB, Hz, deg, ...])
Bode plot of the frequency response
```

Plots a bode gain and phase diagram

#### **Parameters**

- **sys** (*LTI*, *or list of LTI*) System for which the Bode response is plotted and give. Optionally a list of systems can be entered, or several systems can be specified (i.e. several parameters). The sys arguments may also be interspersed with format strings. A frequency argument (array\_like) may also be added, some examples: \*>>> bode(sys, w) # one system, freq vector \*>>> bode(sys1, sys2, ..., sysN) # several systems \*>>> bode(sys1, sys2, ..., sysN, w) \*>>> bode(sys1, 'plotstyle1', ..., sysN, 'plotstyleN') # + plot formats
- omega (freq\_range) Range of frequencies in rad/s
- dB (boolean) If True, plot result in dB
- Hz (boolean) If True, plot frequency in Hz (omega must be provided in rad/sec)
- deg (boolean) If True, return phase in degrees (else radians)
- Plot (boolean) If True, plot magnitude and phase

## **Examples**

```
>>> sys = ss("1. -2; 3. -4", "5.; 7", "6. 8", "9.")
>>> mag, phase, omega = bode(sys)
```

## **Todo:** Document these use cases

```
>>> bode(sys, w)
```

```
• >>> bode(sys1, sys2, ..., sysN)
• >>> bode(sys1, sys2, ..., sysN, w)
```

```
>>> bode(sys1, 'plotstyle1', ..., sysN, 'plotstyleN')
```

# 5.6.2 control.matlab.nyquist

control.matlab.nyquist(syslist, omega=None, plot=True, label\_freq=0, arrowhead\_length=0.1, arrowhead\_width=0.1, color=None, \*args, \*\*kwargs)

Nyquist plot for a system

Plots a Nyquist plot for the system over a (optional) frequency range.

#### **Parameters**

- syslist (list of LTI) List of linear input/output systems (single system is OK)
- omega (freq\_range) Range of frequencies (list or bounds) in rad/sec
- Plot (boolean) If True, plot magnitude
- **color** (*string*) Used to specify the color of the plot
- label\_freq (int) Label every nth frequency on the plot
- arrowhead\_width(arrow head width)-
- arrowhead length (arrow head length) -
- \*args (matplotlib.pyplot.plot() positional properties, optional) Additional arguments for *matplotlib* plots (color, linestyle, etc)
- \*\*kwargs (matplotlib.pyplot.plot() keyword properties, optional) Additional keywords (passed to *matplotlib*)

#### Returns

- real (array) real part of the frequency response array
- imag (array) imaginary part of the frequency response array
- **freq** (array) frequencies

# **Examples**

```
>>> sys = ss("1. -2; 3. -4", "5.; 7", "6. 8", "9.")
>>> real, imag, freq = nyquist_plot(sys)
```

## 5.6.3 control.matlab.nichols

control.matlab.nichols (sys\_list, omega=None, grid=None)
 Nichols plot for a system

Plots a Nichols plot for the system over a (optional) frequency range.

#### **Parameters**

- **sys\_list** (list of LTI, or LTI) List of linear input/output systems (single system is OK)
- omega (array\_like) Range of frequencies (list or bounds) in rad/sec
- grid (boolean, optional) True if the plot should include a Nichols-chart grid. Default is True.

#### Returns

Return type None

# 5.6.4 control.matlab.margin

```
control.matlab.margin(sysdata)
```

Calculate gain and phase margins and associated crossover frequencies

```
Parameters sysdata (LTI system or (mag, phase, omega) sequence) -
```

sys [StateSpace or TransferFunction] Linear SISO system

mag, phase, omega [sequence of array\_like] Input magnitude, phase (in deg.), and frequencies (rad/sec) from bode frequency response data

#### Returns

- gm (float) Gain margin
- **pm** (*float*) Phase margin (in degrees)
- $\mathbf{wg}$  (float) Frequency for gain margin (at phase crossover, phase = -180 degrees)
- **wp** (*float*) Frequency for phase margin (at gain crossover, gain = 1)
- Margins are calculated for a SISO open-loop system.
- If there is more than one gain crossover, the one at the smallest
- margin (deviation from gain = 1), in absolute sense, is
- returned. Likewise the smallest phase margin (in absolute sense)
- is returned.

## **Examples**

```
>>> sys = tf(1, [1, 2, 1, 0])
>>> gm, pm, wg, wp = margin(sys)
```

# 5.6.5 control.matlab.freqresp

```
control.matlab.freqresp(sys, omega)
```

Frequency response of an LTI system at multiple angular frequencies.

#### **Parameters**

- sys (StateSpace or TransferFunction) Linear system
- omega (array\_like) A list of frequencies in radians/sec at which the system should be evaluated. The list can be either a python list or a numpy array and will be sorted before evaluation.

#### Returns

- mag ((self.outputs, self.inputs, len(omega)) ndarray) The magnitude (absolute value, not dB or log10) of the system frequency response.
- **phase** ((self.outputs, self.inputs, len(omega)) ndarray) The wrapped phase in radians of the system frequency response.
- omega (ndarray or list or tuple) The list of sorted frequencies at which the response was
  evaluated.

#### See also:

evalfr, bode

#### **Notes**

This function is a wrapper for StateSpace.freqresp and TransferFunction.freqresp. The output omega is a sorted version of the input omega.

#### **Examples**

```
>>> sys = ss("1. -2; 3. -4", "5.; 7", "6. 8", "9.")
>>> mag, phase, omega = freqresp(sys, [0.1, 1., 10.])
>>> mag
array([[[ 58.8576682 , 49.64876635, 13.40825927]]])
>>> phase
array([[[-0.05408304, -0.44563154, -0.66837155]]])
```

Todo: Add example with MIMO system

#>> sys = rss(3, 2, 2) #>> mag, phase, omega = freqresp(sys, [0.1, 1., 10.]) #>> mag[0, 1, :] # array([55.43747231, 42.47766549, 1.97225895]) #>> phase[1, 0, :] # array([-0.12611087, -1.14294316, 2.5764547]) #>> # This is the magnitude of the frequency response from the 2nd #>> # input to the 1st output, and the phase (in radians) of the #>> # frequency response from the 1st input to the 2nd output, for #>> # s = 0.1i, i, 10i.

# 5.6.6 control.matlab.evalfr

```
control.matlab.evalfr(sys, x)
```

Evaluate the transfer function of an LTI system for a single complex number x.

To evaluate at a frequency, enter x = omega\*j, where omega is the frequency in radians

#### **Parameters**

- sys (StateSpace or TransferFunction) Linear system
- **x** (scalar) Complex number

## Returns fresp

Return type ndarray

#### See also:

freqresp, bode

#### **Notes**

This function is a wrapper for StateSpace.evalfr and TransferFunction.evalfr.

# **Examples**

```
>>> sys = ss("1. -2; 3. -4", "5.; 7", "6. 8", "9.")
>>> evalfr(sys, 1j)
array([[ 44.8-21.4j]])
>>> # This is the transfer function matrix evaluated at s = i.
```

**Todo:** Add example with MIMO system

# 5.7 Compensator design

rlocus(sys[, kvect, xlim, ylim, plotstr,])	Root locus plot
sisotool(sys[, kvect, xlim_rlocus,])	Sisotool style collection of plots inspired by MATLAB's
	sisotool.
place(A, B, p)	Place closed loop eigenvalues
Iqr(A, B, Q, R[, N])	Linear quadratic regulator design

## 5.7.1 control.matlab.rlocus

control.matlab.rlocus(sys, kvect=None, xlim=None, ylim=None, plotstr=None, plot=True, print\_gain=None, grid=None, ax=None, \*\*kwargs)

## Root locus plot

Calculate the root locus by finding the roots of 1+k\*TF(s) where TF is self.num(s)/self.den(s) and each k is an element of kvect.

#### **Parameters**

- **sys** (LTI object) Linear input/output systems (SISO only, for now).
- kvect (list or ndarray, optional) List of gains to use in computing diagram.
- **xlim** (tuple or list, optional) Set limits of x axis, normally with tuple (see matplotlib.axes).
- ylim (tuple or list, optional) Set limits of y axis, normally with tuple (see matplotlib.axes).
- plotstr (matplotlib.pyplot.plot() format string, optional) plotting style specification
- plot (boolean, optional) If True (default), plot root locus diagram.
- **print\_gain** (bool) If True (default), report mouse clicks when close to the root locus branches, calculate gain, damping and print.
- **grid** (bool) If True plot omega-damping grid. Default is False.
- ax (matplotlib.axes.Axes) Axes on which to create root locus plot

#### Returns

- rlist (ndarray) Computed root locations, given as a 2D array
- **klist** (*ndarray or list*) Gains used. Same as klist keyword argument if provided.

## 5.7.2 control.matlab.sisotool

Sisotool style collection of plots inspired by MATLAB's sisotool. The left two plots contain the bode magnitude and phase diagrams. The top right plot is a clickable root locus plot, clicking on the root locus will change the gain of the system. The bottom left plot shows a closed loop time response.

## **Parameters**

- **sys** (LTI object) Linear input/output systems (SISO only)
- **kvect** (list or ndarray, optional) List of gains to use for plotting root locus
- xlim\_rlocus (tuple or list, optional) control of x-axis range, normally with tuple (see matplotlib.axes).
- ylim\_rlocus (tuple or list, optional) control of y-axis range
- plotstr\_rlocus (matplotlib.pyplot.plot() format string, optional) plotting style for the root locus plot(color, linestyle, etc)
- rlocus\_grid(boolean (default = False)) If True plot s-plane grid.

- omega (freq\_range) Range of frequencies in rad/sec for the bode plot
- dB (boolean) If True, plot result in dB for the bode plot
- **Hz** (boolean) If True, plot frequency in Hz for the bode plot (omega must be provided in rad/sec)
- **deg** (boolean) If True, plot phase in degrees for the bode plot (else radians)
- omega\_limits (tuple, list, .. of two values) Limits of the to generate frequency vector. If Hz=True the limits are in Hz otherwise in rad/s.
- omega\_num (int) number of samples
- margins\_bode (boolean) If True, plot gain and phase margin in the bode plot
- tvect (list or ndarray, optional) List of timesteps to use for closed loop step response

## **Examples**

```
>>> sys = tf([1000], [1,25,100,0])
>>> sisotool(sys)
```

# 5.7.3 control.matlab.place

```
control.matlab.place (A, B, p)
Place closed loop eigenvalues
```

K = place(A, B, p)

#### **Parameters**

- A (2D array) Dynamics matrix
- B (2D array) Input matrix
- p (1D list) Desired eigenvalue locations

**Returns** K – Gain such that A - B K has eigenvalues given in p

**Return type** 2D array (or matrix)

#### **Notes**

Algorithm This is a wrapper function for <code>scipy.signal.place\_poles()</code>, which implements the Tits and Yang algorithm<sup>1</sup>. It will handle SISO, MISO, and MIMO systems. If you want more control over the algorithm, use <code>scipy.signal.place\_poles()</code> directly.

**Limitations** The algorithm will not place poles at the same location more than rank(B) times.

The return type for 2D arrays depends on the default class set for state space operations. See  $use\_numpy\_matrix()$ .

<sup>&</sup>lt;sup>1</sup> A.L. Tits and Y. Yang, "Globally convergent algorithms for robust pole assignment by state feedback, IEEE Transactions on Automatic Control, Vol. 41, pp. 1432-1452, 1996.

#### References

## **Examples**

```
>>> A = [[-1, -1], [0, 1]]
>>> B = [[0], [1]]
>>> K = place(A, B, [-2, -5])
```

## See also:

place\_varga, acker

## **Notes**

The return type for 2D arrays depends on the default class set for state space operations. See  $use\_numpy\_matrix()$ .

# 5.7.4 control.matlab.lqr

```
control.matlab.lqr(A, B, Q, R[, N])
```

Linear quadratic regulator design

The lqr() function computes the optimal state feedback controller that minimizes the quadratic cost

$$J = \int_0^\infty (x'Qx + u'Ru + 2x'Nu)dt$$

The function can be called with either 3, 4, or 5 arguments:

- lqr(sys, Q, R)
- lqr(sys, Q, R, N)
- lqr(A, B, Q, R)
- lqr(A, B, Q, R, N)

where sys is an LTI object, and A, B, Q, R, and N are 2d arrays or matrices of appropriate dimension.

#### **Parameters**

- A (2D array) Dynamics and input matrices
- B (2D array) Dynamics and input matrices
- sys (LTI (StateSpace or TransferFunction)) Linear I/O system
- Q (2D array) State and input weight matrices
- **R** (2D array) State and input weight matrices
- N (2D array, optional) Cross weight matrix

## Returns

- **K** (2D array (or matrix)) State feedback gains
- S (2D array (or matrix)) Solution to Riccati equation
- E (1D array) Eigenvalues of the closed loop system

#### See also:

lqe

## **Notes**

The return type for 2D arrays depends on the default class set for state space operations. See  $use\_numpy\_matrix()$ .

# **Examples**

```
>>> K, S, E = lqr(sys, Q, R, [N])
>>> K, S, E = lqr(A, B, Q, R, [N])
```

# 5.8 State-space (SS) models

rss([states, outputs, inputs])	Create a stable <i>continuous</i> random state space object.
drss([states, outputs, inputs])	Create a stable <i>discrete</i> random state space object.
ctrb(A, B)	Controllabilty matrix
obsv(A, C)	Observability matrix
gram(sys, type)	Gramian (controllability or observability)

## 5.8.1 control.matlab.ctrb

control.matlab.ctrb (A, B)Controllabilty matrix

## **Parameters**

- A (array\_like or string) Dynamics and input matrix of the system
- B (array\_like or string) Dynamics and input matrix of the system

**Returns** C – Controllability matrix

**Return type** 2D array (or matrix)

## **Notes**

The return type for 2D arrays depends on the default class set for state space operations. See  $use\_numpy\_matrix()$ .

## **Examples**

```
>>> C = ctrb(A, B)
```

## 5.8.2 control.matlab.obsv

```
control.matlab.obsv(A, C)
Observability matrix
```

#### **Parameters**

- A (array\_like or string) Dynamics and output matrix of the system
- C (array\_like or string) Dynamics and output matrix of the system

**Returns** O – Observability matrix

**Return type** 2D array (or matrix)

#### **Notes**

The return type for 2D arrays depends on the default class set for state space operations. See  $use\_numpy\_matrix()$ .

## **Examples**

```
>>> O = obsv(A, C)
```

# 5.8.3 control.matlab.gram

```
control.matlab.gram(sys, type)
Gramian(controllability or observability)
```

#### **Parameters**

- **sys** (StateSpace) System description
- **type** (*String*) Type of desired computation. *type* is either 'c' (controllability) or 'o' (observability). To compute the Cholesky factors of Gramians use 'cf' (controllability) or 'of' (observability)

Returns gram – Gramian of system

**Return type** 2D array (or matrix)

# Raises

- ValueError -
  - if system is not instance of StateSpace class \* if *type* is not 'c', 'o', 'cf' or 'of' \* if system is unstable (sys.A has eigenvalues not in left half plane)
- ImportError if slycot routine sb03md cannot be found if slycot routine sb03od cannot be found

## **Notes**

The return type for 2D arrays depends on the default class set for state space operations. See  $use\_numpy\_matrix()$ .

## **Examples**

```
>>> Wc = gram(sys, 'c')
>>> Wo = gram(sys, 'o')
>>> Rc = gram(sys, 'cf'), where Wc = Rc' * Rc
>>> Ro = gram(sys, 'of'), where Wo = Ro' * Ro
```

# 5.9 Model simplification

minreal(sys[, tol, verbose])	Eliminates uncontrollable or unobservable states in
	state-space models or cancelling pole-zero pairs in
	transfer functions.
hsvd(sys)	Calculate the Hankel singular values.
balred(sys, orders[, method, alpha])	Balanced reduced order model of sys of a given order.
modred(sys, ELIM[, method])	Model reduction of sys by eliminating the states in
	ELIM using a given method.
era(YY, m, n, nin, nout, r)	Calculate an ERA model of order r based on the
	impulse-response data YY.
markov(Y, U[, m, transpose])	Calculate the first <i>m</i> Markov parameters [D CB CAB
	$\dots$ ] from input $U$ , output $Y$ .

## 5.9.1 control.matlab.minreal

```
control.matlab.minreal(sys, tol=None, verbose=True)
```

Eliminates uncontrollable or unobservable states in state-space models or cancelling pole-zero pairs in transfer functions. The output sysr has minimal order and the same response characteristics as the original model sys.

#### **Parameters**

- sys (StateSpace or TransferFunction) Original system
- tol (real) Tolerance
- verbose (bool) Print results if True

Returns rsys - Cleaned model

**Return type** *StateSpace* or *TransferFunction* 

## 5.9.2 control.matlab.hsvd

```
control.matlab.hsvd(sys)
```

Calculate the Hankel singular values.

Parameters sys (StateSpace) – A state space system

**Returns** H – A list of Hankel singular values

Return type array

See also:

gram

#### **Notes**

The Hankel singular values are the singular values of the Hankel operator. In practice, we compute the square root of the eigenvalues of the matrix formed by taking the product of the observability and controllability gramians. There are other (more efficient) methods based on solving the Lyapunov equation in a particular way (more details soon).

# **Examples**

```
>>> H = hsvd(sys)
```

## 5.9.3 control.matlab.balred

```
control.matlab.balred(sys, orders, method='truncate', alpha=None)
```

Balanced reduced order model of sys of a given order. States are eliminated based on Hankel singular value. If sys has unstable modes, they are removed, the balanced realization is done on the stable part, then reinserted in accordance with the reference below.

Reference: Hsu, C.S., and Hou, D., 1991, Reducing unstable linear control systems via real Schur transformation. Electronics Letters, 27, 984-986.

## Parameters

- **sys** (StateSpace) Original system to reduce
- orders (integer or array of integer) Desired order of reduced order model (if a vector, returns a vector of systems)
- method (string) Method of removing states, either 'truncate' or 'matchdc'.
- alpha (float) Redefines the stability boundary for eigenvalues of the system matrix A. By default for continuous-time systems, alpha <= 0 defines the stability boundary for the real part of A's eigenvalues and for discrete-time systems, 0 <= alpha <= 1 defines the stability boundary for the modulus of A's eigenvalues. See SLICOT routines AB09MD and AB09ND for more information.

Returns rsys – A reduced order model or a list of reduced order models if orders is a list.

Return type StateSpace

Raises

• ValueError - If method is not 'truncate' or 'matchdc'

- ImportError if slycot routine ab09ad, ab09md, or ab09nd is not found
- ValueError if there are more unstable modes than any value in orders

## **Examples**

```
>>> rsys = balred(sys, orders, method='truncate')
```

## 5.9.4 control.matlab.modred

```
control.matlab.modred(sys, ELIM, method='matchdc')
```

Model reduction of sys by eliminating the states in *ELIM* using a given method.

## **Parameters**

- **sys** (StateSpace) Original system to reduce
- **ELIM** (array) Vector of states to eliminate
- method (string) Method of removing states in ELIM: either 'truncate' or 'matchdc'.

**Returns** rsys – A reduced order model

**Return type** *StateSpace* 

**Raises ValueError** – Raised under the following conditions:

- \* if method is not either 'matchdc' or 'truncate'
- \* if eigenvalues of sys.A are not all in left half plane (sys must be stable)

## **Examples**

```
>>> rsys = modred(sys, ELIM, method='truncate')
```

## 5.9.5 control.matlab.era

```
control.matlab.era (YY, m, n, nin, nout, r)
```

Calculate an ERA model of order *r* based on the impulse-response data *YY*.

**Note:** This function is not implemented yet.

## **Parameters**

- YY (array) nout x nin dimensional impulse-response data
- m (integer) Number of rows in Hankel matrix
- n (integer) Number of columns in Hankel matrix
- nin (integer) Number of input variables
- nout (integer) Number of output variables
- r (integer) Order of model

**Returns** sys – A reduced order model sys=ss(Ar,Br,Cr,Dr)

Return type StateSpace

## **Examples**

```
>>> rsys = era(YY, m, n, nin, nout, r)
```

## 5.9.6 control.matlab.markov

control.matlab.markov(Y, U, m=None, transpose=None)

Calculate the first m Markov parameters [D CB CAB ...] from input U, output Y.

This function computes the Markov parameters for a discrete time system

$$x[k+1] = Ax[k] + Bu[k]$$
$$y[k] = Cx[k] + Du[k]$$

given data for u and y. The algorithm assumes that that C  $A^k$  B = 0 for k > m-2 (see<sup>1</sup>). Note that the problem is ill-posed if the length of the input data is less than the desired number of Markov parameters (a warning message is generated in this case).

#### **Parameters**

- Y (array\_like) Output data. If the array is 1D, the system is assumed to be single input. If the array is 2D and transpose=False, the columns of Y are taken as time points, otherwise the rows of Y are taken as time points.
- **U** (array\_like) Input data, arranged in the same way as Y.
- m(int, optional) Number of Markov parameters to output. Defaults to len(U).
- **transpose** (bool, optional) Assume that input data is transposed relative to the standard *Time series data*. The default value is true for backward compatibility with legacy code.

**Returns** H – First m Markov parameters, [D CB CAB ...]

Return type ndarray

## References

## **Notes**

Currently only works for SISO systems.

This function does not currently comply with the Python Control Library *Time series data* for representation of time series data. Use *transpose=False* to make use of the standard convention (this will be updated in a future release).

<sup>&</sup>lt;sup>1</sup> J.-N. Juang, M. Phan, L. G. Horta, and R. W. Longman, Identification of observer/Kalman filter Markov parameters - Theory and experiments. Journal of Guidance Control and Dynamics, 16(2), 320-329, 2012. http://doi.org/10.2514/3.21006

## **Examples**

```
>>> T = numpy.linspace(0, 10, 100)

>>> U = numpy.ones((1, 100))

>>> T, Y, _ = forced_response(tf([1], [1, 0.5], True), T, U)

>>> H = markov(Y, U, 3, transpose=False)
```

# 5.10 Time delays

pade(T[, n, numdeg])	Create a linear system that approximates a delay.

# 5.10.1 control.matlab.pade

```
control.matlab.pade (T, n=1, numdeg=None)
```

Create a linear system that approximates a delay.

Return the numerator and denominator coefficients of the Pade approximation.

#### **Parameters**

- T (number) time delay
- n (positive integer) degree of denominator of approximation
- numdeg (integer, or None (the default)) If None, numerator degree equals denominator degree If >= 0, specifies degree of numerator If < 0, numerator degree is n+numdeg

**Returns num, den** – Polynomial coefficients of the delay model, in descending powers of s.

Return type array

#### **Notes**

#### Based on:

- 1. Algorithm 11.3.1 in Golub and van Loan, "Matrix Computation" 3rd. Ed. pp. 572-574
- 2. M. Vajta, "Some remarks on Padé-approximations", 3rd TEMPUS-INTCOM Symposium

# 5.11 Matrix equation solvers and linear algebra

lyap(A, Q[, C, E])	X = lyap(A, Q) solves the continuous-time Lyapunov equation
dlyap(A, Q[, C, E])	dlyap(A,Q) solves the discrete-time Lyapunov equation
care(A, B, Q[, R, S, E, stabilizing])	(X, L, G) = care(A, B, Q, R=None) solves the
	continuous-time algebraic Riccati equation
dare(A, B, Q, R[, S, E, stabilizing])	(X, L, G) = dare(A, B, Q, R) solves the discrete-time
	algebraic Riccati equation

# 5.11.1 control.matlab.lyap

control.matlab.lyap (A, Q, C=None, E=None)

X = Iyap(A, Q) solves the continuous-time Lyapunov equation

$$AX + XA^T + Q = 0$$

where A and Q are square matrices of the same dimension. Further, Q must be symmetric.

X = lyap(A, Q, C) solves the Sylvester equation

$$AX + XQ + C = 0$$

where A and Q are square matrices.

X = lyap(A, Q, None, E) solves the generalized continuous-time Lyapunov equation

$$AXE^T + EXA^T + Q = 0$$

where Q is a symmetric matrix and A, Q and E are square matrices of the same dimension.

#### **Parameters**

- A (2D array) Dynamics matrix
- C (2D array, optional) If present, solve the Slyvester equation
- E (2D array, optional) If present, solve the generalized Laypunov equation

Returns Q – Solution to the Lyapunov or Sylvester equation

**Return type** 2D array (or matrix)

#### **Notes**

The return type for 2D arrays depends on the default class set for state space operations. See  $use\_numpy\_matrix()$ .

# 5.11.2 control.matlab.dlyap

control.matlab.dlyap (A, Q, C=None, E=None)

dlyap(A,Q) solves the discrete-time Lyapunov equation

$$AXA^T - X + Q = 0$$

where A and Q are square matrices of the same dimension. Further Q must be symmetric.

dlyap(A,Q,C) solves the Sylvester equation

$$AXQ^T - X + C = 0$$

where A and Q are square matrices.

dlyap(A,Q,None,E) solves the generalized discrete-time Lyapunov equation

$$AXA^T - EXE^T + Q = 0$$

where Q is a symmetric matrix and A, Q and E are square matrices of the same dimension.

## 5.11.3 control.matlab.care

control.matlab.care (A, B, Q, R=None, S=None, E=None, stabilizing=True)

(X, L, G) = care(A, B, Q, R=None) solves the continuous-time algebraic Riccati equation

$$A^TX + XA - XBR^{-1}B^TX + Q = 0$$

where A and Q are square matrices of the same dimension. Further, Q and R are a symmetric matrices. If R is None, it is set to the identity matrix. The function returns the solution X, the gain matrix  $G = B^T X$  and the closed loop eigenvalues L, i.e., the eigenvalues of A - B G.

(X, L, G) = care(A, B, Q, R, S, E) solves the generalized continuous-time algebraic Riccati equation

$$A^{T}XE + E^{T}XA - (E^{T}XB + S)R^{-1}(B^{T}XE + S^{T}) + Q = 0$$

where A, Q and E are square matrices of the same dimension. Further, Q and R are symmetric matrices. If R is None, it is set to the identity matrix. The function returns the solution X, the gain matrix  $G = R^{-1}$  (B^T X E + S^T) and the closed loop eigenvalues L, i.e., the eigenvalues of A - B G, E.

#### **Parameters**

- A (2D arrays) Input matrices for the Riccati equation
- **B** (2D arrays) Input matrices for the Riccati equation
- Q (2D arrays) Input matrices for the Riccati equation
- R (2D arrays, optional) Input matrices for generalized Riccati equation
- **S** (2D arrays, optional) Input matrices for generalized Riccati equation
- E (2D arrays, optional) Input matrices for generalized Riccati equation

#### Returns

- X (2D array (or matrix)) Solution to the Ricatti equation
- L (1D array) Closed loop eigenvalues
- **G** (2D array (or matrix)) Gain matrix

#### **Notes**

The return type for 2D arrays depends on the default class set for state space operations. See  $use\_numpy\_matrix()$ .

## 5.11.4 control.matlab.dare

control.matlab.dare(A, B, Q, R, S=None, E=None, stabilizing=True)

(X, L, G) = dare(A, B, Q, R) solves the discrete-time algebraic Riccati equation

$$A^{T}XA - X - A^{T}XB(B^{T}XB + R)^{-1}B^{T}XA + Q = 0$$

where A and Q are square matrices of the same dimension. Further, Q is a symmetric matrix. The function returns the solution X, the gain matrix  $G = (B^T X B + R)^{-1} B^T X A$  and the closed loop eigenvalues L, i.e., the eigenvalues of A - B G.

(X, L, G) = dare(A, B, Q, R, S, E) solves the generalized discrete-time algebraic Riccati equation

$$A^{T}XA - E^{T}XE - (A^{T}XB + S)(B^{T}XB + R)^{-1}(B^{T}XA + S^{T}) + Q = 0$$

where A, Q and E are square matrices of the same dimension. Further, Q and R are symmetric matrices. The function returns the solution X, the gain matrix  $G = (B^T X B + R)^{-1} (B^T X A + S^T)$  and the closed loop eigenvalues L, i.e., the eigenvalues of A - B G, E.

#### **Parameters**

- A (2D arrays) Input matrices for the Riccati equation
- **B** (2D arrays) Input matrices for the Riccati equation
- Q (2D arrays) Input matrices for the Riccati equation
- R (2D arrays, optional) Input matrices for generalized Riccati equation
- **S** (2D arrays, optional) Input matrices for generalized Riccati equation
- **E** (2D arrays, optional) Input matrices for generalized Riccati equation

#### Returns

- **X** (2D array (or matrix)) Solution to the Ricatti equation
- L (1D array) Closed loop eigenvalues
- **G** (2D array (or matrix)) Gain matrix

#### **Notes**

The return type for 2D arrays depends on the default class set for state space operations. See  $use\_numpy\_matrix()$ .

# 5.12 Additional functions

gangof4(P, C[, omega])	Plot the "Gang of 4" transfer functions for a system
unwrap(angle[, period])	Unwrap a phase angle to give a continuous curve

## 5.12.1 control.matlab.gangof4

```
control.matlab.gangof4 (P, C, omega=None, **kwargs)
```

Plot the "Gang of 4" transfer functions for a system

Generates a 2x2 plot showing the "Gang of 4" sensitivity functions [T, PS; CS, S]

#### **Parameters**

- **P** (*LTI*) Linear input/output systems (process and control)
- **C** (LTI) Linear input/output systems (process and control)
- omega (array) Range of frequencies (list or bounds) in rad/sec
- \*\*kwargs (matplotlib.pyplot.plot() keyword properties, optional) Additional keywords (passed to *matplotlib*)

## Returns

Return type None

# 5.12.2 control.matlab.unwrap

```
control.matlab.unwrap (angle, period=6.283185307179586) Unwrap a phase angle to give a continuous curve
```

## **Parameters**

- angle (array\_like) Array of angles to be unwrapped
- period (float, optional) Period (defaults to 2\*pi)

**Returns angle\_out** – Output array, with jumps of period/2 eliminated **Return type** array\_like

# **Examples**

```
>>> import numpy as np
>>> theta = [5.74, 5.97, 6.19, 0.13, 0.35, 0.57]
>>> unwrap(theta, period=2 * np.pi)
[5.74, 5.97, 6.19, 6.413185307179586, 6.633185307179586, 6.8531853071795865]
```

# 5.13 Functions imported from other modules

linspace(start, stop[, num, endpoint,])	Return evenly spaced numbers over a specified interval.
logspace(start, stop[, num, endpoint, base,])	Return numbers spaced evenly on a log scale.
ss2zpk(A, B, C, D[, input])	State-space representation to zero-pole-gain representa-
	tion.
tf2zpk(b, a)	Return zero, pole, gain (z, p, k) representation from a
	numerator, denominator representation of a linear filter.
zpk2ss(z, p, k)	Zero-pole-gain representation to state-space representa-
	tion
zpk2tf(z, p, k)	Return polynomial transfer function representation from
	zeros and poles

# DIFFERENTIALLY FLAT SYSTEMS

The *control*. *flatsys* package contains a set of classes and functions that can be used to compute trajectories for differentially flat systems.

A differentially flat system is defined by creating an object using the FlatSystem class, which has member functions for mapping the system state and input into and out of flat coordinates. The point\_to\_point() function can be used to create a trajectory between two endpoints, written in terms of a set of basis functions defined using the BasisFamily class. The resulting trajectory is return as a SystemTrajectory object and can be evaluated using the eval() member function.

# 6.1 Overview of differential flatness

A nonlinear differential equation of the form

$$\dot{x} = f(x, u), \qquad x \in \mathbb{R}^n, u \in \mathbb{R}^m$$

is differentially flat if there exists a function  $\alpha$  such that

$$z = \alpha(x, u, \dot{u}, \dots, u^{(p)})$$

and we can write the solutions of the nonlinear system as functions of z and a finite number of derivatives

$$x = \beta(z, \dot{z}, \dots, z^{(q)}) u = \gamma(z, \dot{z}, \dots, z^{(q)}).$$
(6.1)

For a differentially flat system, all of the feasible trajectories for the system can be written as functions of a flat output  $z(\cdot)$  and its derivatives. The number of flat outputs is always equal to the number of system inputs.

Differentially flat systems are useful in situations where explicit trajectory generation is required. Since the behavior of a flat system is determined by the flat outputs, we can plan trajectories in output space, and then map these to appropriate inputs. Suppose we wish to generate a feasible trajectory for the the nonlinear system

$$\dot{x} = f(x, u), \qquad x(0) = x_0, x(T) = x_f.$$

If the system is differentially flat then

$$x(0) = \beta(z(0), \dot{z}(0), \dots, z^{(q)}(0)) = x_0,$$
  
$$x(T) = \gamma(z(T), \dot{z}(T), \dots, z^{(q)}(T)) = x_f,$$

and we see that the initial and final condition in the full state space depends on just the output z and its derivatives at the initial and final times. Thus any trajectory for z that satisfies these boundary conditions will be a feasible trajectory for the system, using equation (6.1) to determine the full state space and input trajectories.

In particular, given initial and final conditions on z and its derivatives that satisfy the initial and final conditions any curve  $z(\cdot)$  satisfying those conditions will correspond to a feasible trajectory of the system. We can parameterize the flat output trajectory using a set of smooth basis functions  $\psi_i(t)$ :

$$z(t) = \sum_{i=1}^{N} \alpha_i \psi_i(t), \qquad \alpha_i \in R$$

We seek a set of coefficients  $\alpha_i$ ,  $i=1,\ldots,N$  such that z(t) satisfies the boundary conditions for x(0) and x(T). The derivatives of the flat output can be computed in terms of the derivatives of the basis functions:

$$\dot{z}(t) = \sum_{i=1}^{N} \alpha_i \dot{\psi}_i(t)$$

:

$$\dot{z}^{(q)}(t) = \sum_{i=1}^{N} \alpha_i \psi_i^{(q)}(t).$$

We can thus write the conditions on the flat outputs and their derivatives as

$$\begin{bmatrix} \psi_{1}(0) & \psi_{2}(0) & \dots & \psi_{N}(0) \\ \dot{\psi}_{1}(0) & \dot{\psi}_{2}(0) & \dots & \dot{\psi}_{N}(0) \\ \vdots & \vdots & & \vdots \\ \psi_{1}^{(q)}(0) & \psi_{2}^{(q)}(0) & \dots & \psi_{N}^{(q)}(0) \\ \psi_{1}(T) & \psi_{2}(T) & \dots & \psi_{N}(T) \\ \dot{\psi}_{1}(T) & \dot{\psi}_{2}(T) & \dots & \dot{\psi}_{N}(T) \\ \vdots & \vdots & & \vdots \\ \psi_{1}^{(q)}(T) & \psi_{2}^{(q)}(T) & \dots & \psi_{N}^{(q)}(T) \end{bmatrix} \begin{bmatrix} \alpha_{1} \\ \vdots \\ \alpha_{N} \end{bmatrix} = \begin{bmatrix} z(0) \\ \dot{z}(0) \\ \vdots \\ z^{(q)}(0) \\ \vdots \\ z^{(q)}(T) \\ \vdots \\ z^{(q)}(T) \end{bmatrix}$$

This equation is a linear equation of the form

$$M\alpha = \begin{bmatrix} \bar{z}(0) \\ \bar{z}(T) \end{bmatrix}$$

where  $\bar{z}$  is called the *flat flag* for the system. Assuming that M has a sufficient number of columns and that it is full column rank, we can solve for a (possibly non-unique)  $\alpha$  that solves the trajectory generation problem.

# 6.2 Module usage

To create a trajectory for a differentially flat system, a FlatSystem object must be created. This is done by specifying the *forward* and *reverse* mappings between the system state/input and the differentially flat outputs and their derivatives ("flat flag").

The forward() method computes the flat flag given a state and input:

$$zflag = sys.forward(x, u)$$

The reverse () method computes the state and input given the flat flag:

$$x, u = sys.reverse(zflag)$$

The flag  $\bar{z}$  is implemented as a list of flat outputs  $z_i$  and their derivatives up to order  $q_i$ :

$$zflag[i][j] = z_i^{(j)}$$

The number of flat outputs must match the number of system inputs.

For a linear system, a flat system representation can be generated using the LinearFlatSystem class:

```
flatsys = control.flatsys.LinearFlatSystem(linsys)
```

For more general systems, the FlatSystem object must be created manually

```
flatsys = control.flatsys.FlatSystem(nstate, ninputs, forward, reverse)
```

In addition to the flat system descriptionn, a set of basis functions  $\phi_i(t)$  must be chosen. The *FlatBasis* class is used to represent the basis functions. A polynomial basis function of the form 1, t,  $t^2$ , ... can be computed using the *PolyBasis* class, which is initialized by passing the desired order of the polynomial basis set:

```
polybasis = control.flatsys.PolyBasis(N)
```

Once the system and basis function have been defined, the point\_to\_point() function can be used to compute a trajectory between initial and final states and inputs:

```
traj = control.flatsys.point_to_point(x0, u0, xf, uf, Tf, basis=polybasis)
```

The returned object has class SystemTrajectory and can be used to compute the state and input trajectory between the initial and final condition:

```
xd, ud = traj.eval(T)
```

where T is a list of times on which the trajectory should be evaluated (e.g., T = numpy.linspace(0, Tf, M).

# 6.3 Example

To illustrate how we can use a two degree-of-freedom design to improve the performance of the system, consider the problem of steering a car to change lanes on a road. We use the non-normalized form of the dynamics, which are derived *Feedback Systems* by Astrom and Murray, Example 3.11.

```
import control.flatsys as fs
# Function to take states, inputs and return the flat flag
def vehicle_flat_forward(x, u, params={}):
    # Get the parameter values
   b = params.get('wheelbase', 3.)
    # Create a list of arrays to store the flat output and its derivatives
    zflag = [np.zeros(3), np.zeros(3)]
    # Flat output is the x, y position of the rear wheels
    zflag[0][0] = x[0]
    zflag[1][0] = x[1]
    # First derivatives of the flat output
    zflag[0][1] = u[0] * np.cos(x[2]) # dx/dt
    zflag[1][1] = u[0] * np.sin(x[2]) # dy/dt
    # First derivative of the angle
   thdot = (u[0]/b) * np.tan(u[1])
    # Second derivatives of the flat output (setting vdot = 0)
    zflag[0][2] = -u[0] * thdot * np.sin(x[2])
    zflag[1][2] = u[0] * thdot * np.cos(x[2])
```

(continues on next page)

6.3. Example 129

(continued from previous page)

```
return zflag
# Function to take the flat flag and return states, inputs
def vehicle_flat_reverse(zflag, params={}):
    # Get the parameter values
   b = params.get('wheelbase', 3.)
    # Create a vector to store the state and inputs
   x = np.zeros(3)
   u = np.zeros(2)
    # Given the flat variables, solve for the state
   x[0] = zflag[0][0] # x position
   x[1] = zflag[1][0] # y position
   x[2] = np.arctan2(zflag[1][1], zflag[0][1]) # tan(theta) = ydot/xdot
    # And next solve for the inputs
   u[0] = zflag[0][1] * np.cos(x[2]) + zflag[1][1] * np.sin(x[2])
   u[1] = np.arctan2(
        (zflag[1][2] * np.cos(x[2]) - zflag[0][2] * np.sin(x[2])), u[0]/b)
   return x, u
vehicle_flat = fs.FlatSystem(
    3, 2, forward=vehicle_flat_forward, reverse=vehicle_flat_reverse)
```

To find a trajectory from an initial state  $x_0$  to a final state  $x_f$  in time  $T_f$  we solve a point-to-point trajectory generation problem. We also set the initial and final inputs, which sets the vehicle velocity v and steering wheel angle  $\delta$  at the endpoints.

```
# Define the endpoints of the trajectory
x0 = [0., -2., 0.]; u0 = [10., 0.]
xf = [100., 2., 0.]; uf = [10., 0.]
Tf = 10

# Define a set of basis functions to use for the trajectories
poly = fs.PolyFamily(6)

# Find a trajectory between the initial condition and the final condition
traj = fs.point_to_point(vehicle_flat, x0, u0, xf, uf, Tf, basis=poly)

# Create the trajectory
t = np.linspace(0, Tf, 100)
x, u = traj.eval(t)
```

# 6.4 Module classes and functions

# 6.4.1 Flat systems classes

BasisFamily(N)	Base class for implementing basis functions for flat sys-
	tems.
FlatSystem(forward, reverse[, updfcn,])	Base class for representing a differentially flat system.
	continues on next page

## Table 1 – continued from previous page

LinearFlatSystem(linsys[, inputs, outputs,])	
PolyFamily(N)	Polynomial basis functions.
SystemTrajectory(sys, basis[, coeffs, flaglen])	Class representing a system trajectory.

## control.flatsys.BasisFamily

## class control.flatsys.BasisFamily(N)

Base class for implementing basis functions for flat systems.

A BasisFamily object is used to construct trajectories for a flat system. The class must implement a single function that computes the jth derivative of the ith basis function at a time t:

$$z_i^{(q)}(t)$$
 = basis.eval\_deriv(self, i, j, t)
\_\_init\_\_(N)
Create a basis family of order N.

## **Methods**

init(N)	Create a basis family of order N.

## control.flatsys.FlatSystem

Base class for representing a differentially flat system.

The FlatSystem class is used as a base class to describe differentially flat systems for trajectory generation. The class must implement two functions:

- **zflag = flatsys.foward(x, u)** This function computes the flag (derivatives) of the flat output. The inputs to this function are the state 'x' and inputs 'u' (both 1D arrays). The output should be a 2D array with the first dimension equal to the number of system inputs and the second dimension of the length required to represent the full system dynamics (typically the number of states)
- $\mathbf{x}$ ,  $\mathbf{u} = \mathbf{flatsys.reverse}(\mathbf{zflag})$  This function system state and inputs give the flag (derivatives) of the flat output. The input to this function is an 2D array whose first dimension is equal to the number of system inputs and whose second dimension is of length required to represent the full system dynamics (typically the number of states). The output is the state x and inputs u (both 1D arrays).

A flat system is also an input/output system supporting simulation, composition, and linearization. If the update and output methods are given, they are used in place of the flat coordinates.

\_\_init\_\_ (forward, reverse, updfcn=None, outfcn=None, inputs=None, outputs=None, states=None, params={}, dt=None, name=None}

Create a differentially flat input/output system.

The FlatIOSystem constructor is used to create an input/output system object that also represents a differentially flat system. The output of the system does not need to be the differentially flat output.

#### **Parameters**

- **forward** (callable) A function to compute the flat flag given the states and input.
- reverse (callable) A function to compute the states and input given the flat flag.

• **updfcn** (callable, optional) – Function returning the state update function updfcn(t, x, u[, param]) -> array

where x is a 1-D array with shape (nstates,), u is a 1-D array with shape (ninputs,), t is a float representing the currrent time, and param is an optional dict containing the values of parameters used by the function. If not specified, the state space update will be computed using the flat system coordinates.

• **outfcn** (callable) – Function returning the output at the given state outfcn(t, x, u[, param]) -> array

where the arguments are the same as for *upfcn*. If not specified, the output will be the flat outputs.

- **inputs** (*int*, *list* of *str*, or *None*) Description of the system inputs. This can be given as an integer count or as a list of strings that name the individual signals. If an integer count is specified, the names of the signal will be of the form *s[i]* (where *s* is one of *u*, *y*, or *x*). If this parameter is not given or given as *None*, the relevant quantity will be determined when possible based on other information provided to functions using the system.
- outputs (int, list of str, or None) Description of the system outputs. Same format as *inputs*.
- **states** (*int*, *list* of *str*, or *None*) Description of the system states. Same format as *inputs*.
- dt (None, True or float, optional) System timebase. None (default) indicates continuous time, True indicates discrete time with undefined sampling time, positive number is discrete time with specified sampling time.
- params (dict, optional) Parameter values for the systems. Passed to the evaluation functions for the system as default values, overriding internal defaults.
- name (string, optional) System name (used for specifying signals)

Returns Input/output system object

Return type InputOutputSystem

## **Methods**

init(forward, reverse[, updfcn, outfcn,])	Create a differentially flat input/output system.
copy([newname])	Make a copy of an input/output system.
feedback([other, sign, params])	Feedback interconnection between two input/output
	systems
find_input(name)	Find the index for an input given its name (None if
	not found)
find_output(name)	Find the index for an output given its name ( <i>None</i> if
	not found)
find_state(name)	Find the index for a state given its name ( <i>None</i> if not
	found)
forward(x, u[, params])	Compute the flat flag given the states and input.
linearize(x0, u0[, t, params, eps])	Linearize an input/output system at a given state and
	input.
name_or_default([name])	
	continues on next page

Table 3 – continued from previous page

reverse(zflag[, params])	Compute the states and input given the flat flag.
<pre>set_inputs(inputs[, prefix])</pre>	Set the number/names of the system inputs.
set_outputs(outputs[, prefix])	Set the number/names of the system outputs.
<pre>set_states(states[, prefix])</pre>	Set the number/names of the system states.

#### **Attributes**

#### idCounter

#### copy (newname=None)

Make a copy of an input/output system.

## feedback (other=1, sign=-1, params={})

Feedback interconnection between two input/output systems

#### **Parameters**

- **sys1** (InputOutputSystem) The primary process.
- **sys2** (InputOutputSystem) The feedback process (often a feedback controller).
- **sign** (*scalar*, *optional*) The sign of feedback. *sign* = -1 indicates negative feedback, and *sign* = 1 indicates positive feedback. *sign* is an optional argument; it assumes a value of -1 if not specified.

#### Returns out

Return type InputOutputSystem

**Raises ValueError** – if the inputs, outputs, or timebases of the systems are incompatible.

#### find\_input (name)

Find the index for an input given its name (*None* if not found)

## find\_output (name)

Find the index for an output given its name (*None* if not found)

## find\_state(name)

Find the index for a state given its name (*None* if not found)

#### $forward(x, u, params = \{\})$

Compute the flat flag given the states and input.

Given the states and inputs for a system, compute the flat outputs and their derivatives (the flat "flag") for the system.

#### **Parameters**

- **x** (list or array) The state of the system.
- **u** (list or array) The input to the system.
- params (dict, optional) Parameter values for the system. Passed to the evaluation functions for the system as default values, overriding internal defaults.

**Returns** zflag – For each flat output  $z_i$ , zflag[i] should be an idarray of length  $q_i$  that contains the flat output and its first  $q_i$  derivatives.

**Return type** list of 1D arrays

#### linearize $(x0, u0, t=0, params=\{\}, eps=1e-06)$

Linearize an input/output system at a given state and input.

Return the linearization of an input/output system at a given state and input value as a StateSpace system. See linearize() for complete documentation.

#### reverse (zflag, params={})

Compute the states and input given the flat flag.

#### **Parameters**

- **zflag** (list of arrays) For each flat output  $z_i$ , zflag[i] should be an idarray of length  $q_i$  that contains the flat output and its first  $q_i$  derivatives.
- params (dict, optional) Parameter values for the system. Passed to the evaluation functions for the system as default values, overriding internal defaults.

#### Returns

- x (1D array) The state of the system corresponding to the flat flag.
- **u** (1D array) The input to the system corresponding to the flat flag.

#### set\_inputs (inputs, prefix='u')

Set the number/names of the system inputs.

#### **Parameters**

- **inputs** (*int*, *list* of *str*, or *None*) Description of the system inputs. This can be given as an integer count or as a list of strings that name the individual signals. If an integer count is specified, the names of the signal will be of the form *u[i]* (where the prefix *u* can be changed using the optional prefix parameter).
- **prefix** (string, optional) If inputs is an integer, create the names of the states using the given prefix (default = 'u'). The names of the input will be of the form prefix[i].

## set\_outputs (outputs, prefix='y')

Set the number/names of the system outputs.

#### **Parameters**

- **outputs** (*int*, *list* of *str*, or *None*) Description of the system outputs. This can be given as an integer count or as a list of strings that name the individual signals. If an integer count is specified, the names of the signal will be of the form u[i] (where the prefix u can be changed using the optional prefix parameter).
- **prefix** (string, optional) If outputs is an integer, create the names of the states using the given prefix (default = 'y'). The names of the input will be of the form prefix[i].

#### set states (states, prefix='x')

Set the number/names of the system states.

#### **Parameters**

- **states** (*int*, *list* of *str*, or *None*) Description of the system states. This can be given as an integer count or as a list of strings that name the individual signals. If an integer count is specified, the names of the signal will be of the form *u[i]* (where the prefix *u* can be changed using the optional prefix parameter).
- **prefix** (string, optional) If states is an integer, create the names of the states using the given prefix (default = 'x'). The names of the input will be of the form prefix[i].

#### control.flatsys.LinearFlatSystem

\_\_init\_\_ (linsys, inputs=None, outputs=None, states=None, name=None)
Define a flat system from a SISO LTI system.

Given a reachable, single-input/single-output, linear time-invariant system, create a differentially flat system representation.

#### **Parameters**

- linsys (StateSpace) LTI StateSpace system to be converted
- **inputs** (*int*, *list* of *str* or *None*, *optional*) Description of the system inputs. This can be given as an integer count or as a list of strings that name the individual signals. If an integer count is specified, the names of the signal will be of the form *s[i]* (where *s* is one of *u*, *y*, or *x*). If this parameter is not given or given as *None*, the relevant quantity will be determined when possible based on other information provided to functions using the system.
- **outputs** (int, list of str or None, optional) Description of the system outputs. Same format as *inputs*.
- **states** (int, list of str, or None, optional) Description of the system states. Same format as *inputs*.
- dt (None, True or float, optional) System timebase. None (default) indicates continuous time, True indicates discrete time with undefined sampling time, positive number is discrete time with specified sampling time.
- params (dict, optional) Parameter values for the systems. Passed to the evaluation functions for the system as default values, overriding internal defaults.
- name (string, optional) System name (used for specifying signals)

**Returns** iosys – Linear system represented as an flat input/output system

**Return type** *LinearFlatSystem* 

## **Methods**

init(linsys[, inputs, outputs, states, name])	Define a flat system from a SISO LTI system.
append(other)	Append a second model to the present model.
copy([newname])	Make a copy of an input/output system.
damp()	Natural frequency, damping ratio of system poles
dcgain()	Return the zero-frequency gain
evalfr(omega)	Evaluate a SS system's transfer function at a single
	frequency.
feedback([other, sign, params])	Feedback interconnection between two input/output
	systems
find_input(name)	Find the index for an input given its name (None if
	not found)
find_output(name)	Find the index for an output given its name (None if
	not found)

continues on next page

Table 5 – continued from previous page

Table 5 Continues	a nom previous page
find_state(name)	Find the index for a state given its name ( <i>None</i> if not
	found)
forward(x, u)	Compute the flat flag given the states and input.
freqresp(omega)	Evaluate the system's transfer function at a list of fre-
	quencies
horner(s)	Evaluate the systems's transfer function for a com-
	plex variable
is_static_gain()	True if and only if the system has no dynamics, that
	is, if A and B are zero.
isctime([strict])	Check to see if a system is a continuous-time system
isdtime([strict])	Check to see if a system is a discrete-time system
issiso()	Check to see if a system is single input, single output
1ft(other[, nu, ny])	Return the Linear Fractional Transformation.
linearize(x0, u0[, t, params, eps])	Linearize an input/output system at a given state and
	input.
minreal([tol])	Calculate a minimal realization, removes unobserv-
	able and uncontrollable states
name_or_default([name])	
pole()	Compute the poles of a state space system.
returnScipySignalLTI()	Return a list of a list of scipy.signal.lti ob-
	jects.
reverse(zflag)	Compute the states and input given the flat flag.
sample(Ts[, method, alpha, prewarp_frequency])	Convert a continuous time system to discrete time
set_inputs(inputs[, prefix])	Set the number/names of the system inputs.
set_outputs(outputs[, prefix])	Set the number/names of the system outputs.
set_states(states[, prefix])	Set the number/names of the system states.
zero()	Compute the zeros of a state space system.

## **Attributes**

# idCounter

## append(other)

Append a second model to the present model.

The second model is converted to state-space if necessary, inputs and outputs are appended and their order is preserved

## copy (newname=None)

Make a copy of an input/output system.

# ${\tt damp}\,(\,)$

Natural frequency, damping ratio of system poles

## Returns

- wn (array) Natural frequencies for each system pole
- zeta (array) Damping ratio for each system pole
- poles (array) Array of system poles

# dcgain()

Return the zero-frequency gain

The zero-frequency gain of a continuous-time state-space system is given by:

and of a discrete-time state-space system by:

**Returns gain** – An array of shape (outputs,inputs); the array will either be the zero-frequency (or DC) gain, or, if the frequency response is singular, the array will be filled with np.nan.

#### Return type ndarray

#### evalfr(omega)

Evaluate a SS system's transfer function at a single frequency.

self.\_evalfr(omega) returns the value of the transfer function matrix with input value s = i \* omega.

## feedback (other=1, sign=-1, params={})

Feedback interconnection between two input/output systems

#### **Parameters**

- sys1 (InputOutputSystem) The primary process.
- **sys2** (InputOutputSystem) The feedback process (often a feedback controller).
- **sign** (*scalar*, *optional*) The sign of feedback. *sign* = -1 indicates negative feedback, and *sign* = 1 indicates positive feedback. *sign* is an optional argument; it assumes a value of -1 if not specified.

#### Returns out

Return type InputOutputSystem

Raises ValueError – if the inputs, outputs, or timebases of the systems are incompatible.

## find\_input (name)

Find the index for an input given its name (*None* if not found)

## find\_output (name)

Find the index for an output given its name (*None* if not found)

#### find state(name)

Find the index for a state given its name (*None* if not found)

#### forward(x, u)

Compute the flat flag given the states and input.

See control.flatsys.FlatSystem.forward() for more info.

## freqresp(omega)

Evaluate the system's transfer function at a list of frequencies

Reports the frequency response of the system,

```
G(j*omega) = mag*exp(j*phase)
```

for continuous time. For discrete time systems, the response is evaluated around the unit circle such that

```
G(\exp(j*omega*dt)) = mag*exp(j*phase).
```

**Parameters omega** (array\_like) – A list of frequencies in radians/sec at which the system should be evaluated. The list can be either a python list or a numpy array and will be sorted before evaluation.

## Returns

• mag ((self.outputs, self.inputs, len(omega)) ndarray) – The magnitude (absolute value, not dB or log10) of the system frequency response.

- **phase** ((*self.outputs*, *self.inputs*, *len(omega)*) *ndarray*) The wrapped phase in radians of the system frequency response.
- omega (ndarray) The list of sorted frequencies at which the response was evaluated.

## horner(s)

Evaluate the systems's transfer function for a complex variable

Returns a matrix of values evaluated at complex variable s.

## is\_static\_gain()

True if and only if the system has no dynamics, that is, if A and B are zero.

#### isctime (strict=False)

Check to see if a system is a continuous-time system

#### **Parameters**

- **sys** (LTI system) System to be checked
- **strict** (bool, optional) If strict is True, make sure that timebase is not None. Default is False.

#### isdtime (strict=False)

Check to see if a system is a discrete-time system

**Parameters strict** (bool, optional) – If strict is True, make sure that timebase is not None. Default is False.

#### issiso()

Check to see if a system is single input, single output

## **lft** (other, nu=-1, ny=-1)

Return the Linear Fractional Transformation.

A definition of the LFT operator can be found in Appendix A.7, page 512 in the 2nd Edition, Multivariable Feedback Control by Sigurd Skogestad.

An alternative definition can be found here: https://www.mathworks.com/help/control/ref/lft.html

#### **Parameters**

- other (LTI) The lower LTI system
- ny (int, optional) Dimension of (plant) measurement output.
- nu (int, optional) Dimension of (plant) control input.

## **linearize** (x0, u0, t=0, $params={}$ }, eps=1e-06)

Linearize an input/output system at a given state and input.

Return the linearization of an input/output system at a given state and input value as a StateSpace system. See linearize() for complete documentation.

#### minreal(tol=0.0)

Calculate a minimal realization, removes unobservable and uncontrollable states

#### pole()

Compute the poles of a state space system.

## returnScipySignalLTI()

Return a list of a list of scipy.signal.lti objects.

For instance,

```
>>> out = ssobject.returnScipySignalLTI()
>>> out[3][5]
```

is a scipy.signal.lti object corresponding to the transfer function from the 6th input to the 4th output.

#### reverse (zflag)

Compute the states and input given the flat flag.

See control.flatsys.FlatSystem.reverse() for more info.

```
sample (Ts, method='zoh', alpha=None, prewarp_frequency=None)
```

Convert a continuous time system to discrete time

Creates a discrete-time system from a continuous-time system by sampling. Multiple methods of conversion are supported.

#### **Parameters**

- Ts (float) Sampling period
- method ({"gbt", "bilinear", "euler", "backward\_diff", "zoh"})
  - Which method to use:
  - gbt: generalized bilinear transformation
  - bilinear: Tustin's approximation ("gbt" with alpha=0.5)
  - euler: Euler (or forward differencing) method ("gbt" with alpha=0)
  - backward\_diff: Backwards differencing ("gbt" with alpha=1.0)
  - zoh: zero-order hold (default)
- alpha (float within [0, 1]) The generalized bilinear transformation weighting parameter, which should only be specified with method="gbt", and is ignored otherwise
- **prewarp\_frequency** (float within [0, infinity)) The frequency [rad/s] at which to match with the input continuous- time system's magnitude and phase (the gain=1 crossover frequency, for example). Should only be specified with method='bilinear' or 'gbt' with alpha=0.5 and ignored otherwise.

**Returns** sysd – Discrete time system, with sampling rate Ts

Return type StateSpace

#### **Notes**

Uses scipy.signal.cont2discrete()

## **Examples**

```
>>> sys = StateSpace(0, 1, 1, 0)
>>> sysd = sys.sample(0.5, method='bilinear')
```

## set\_inputs (inputs, prefix='u')

Set the number/names of the system inputs.

#### **Parameters**

- **inputs** (*int*, *list* of *str*, or *None*) Description of the system inputs. This can be given as an integer count or as a list of strings that name the individual signals. If an integer count is specified, the names of the signal will be of the form *u[i]* (where the prefix *u* can be changed using the optional prefix parameter).
- **prefix** (string, optional) If inputs is an integer, create the names of the states using the given prefix (default = 'u'). The names of the input will be of the form prefix[i].

```
set_outputs (outputs, prefix='y')
```

Set the number/names of the system outputs.

#### **Parameters**

- **outputs** (*int*, *list* of *str*, or *None*) Description of the system outputs. This can be given as an integer count or as a list of strings that name the individual signals. If an integer count is specified, the names of the signal will be of the form u[i] (where the prefix u can be changed using the optional prefix parameter).
- **prefix** (string, optional) If outputs is an integer, create the names of the states using the given prefix (default = 'y'). The names of the input will be of the form prefix[i].

```
set_states (states, prefix='x')
```

Set the number/names of the system states.

#### **Parameters**

- **states** (*int*, *list* of *str*, or *None*) Description of the system states. This can be given as an integer count or as a list of strings that name the individual signals. If an integer count is specified, the names of the signal will be of the form *u[i]* (where the prefix *u* can be changed using the optional prefix parameter).
- **prefix** (string, optional) If states is an integer, create the names of the states using the given prefix (default = 'x'). The names of the input will be of the form prefix[i].

#### zero()

Compute the zeros of a state space system.

## control.flatsys.PolyFamily

```
class control.flatsys.PolyFamily(N)
```

Polynomial basis functions.

This class represents the family of polynomials of the form

$$\phi_i(t) = t^i$$

```
___init___(N)
```

Create a polynomial basis of order N.

#### **Methods**

init(N)	Create a polynomial basis of order N.
eval_deriv(i, k, t)	Evaluate the kth derivative of the ith basis function
	at time t.

#### eval deriv(i, k, t)

Evaluate the kth derivative of the ith basis function at time t.

#### control.flatsys.SystemTrajectory

class control.flatsys.SystemTrajectory (sys, basis, coeffs=[], flaglen=[])

Class representing a system trajectory.

The *SystemTrajectory* class is used to represent the trajectory of a (differentially flat) system. Used by the point\_to\_point() function to return a trajectory.

\_\_init\_\_ (sys, basis, coeffs=[], flaglen=[])
Initilize a system trajectory object.

#### **Parameters**

- **sys** (FlatSystem) Flat system object associated with this trajectory.
- basis (BasisFamily) Family of basis vectors to use to represent the trajectory.
- **coeffs** (list of 1D arrays, optional) For each flat output, define the coefficients of the basis functions used to represent the trajectory. Defaults to an empty list.
- **flaglen** (list of ints, optional) For each flat output, the number of derivatives of the flat output used to define the trajectory. Defaults to an empty list.

#### Methods

init(sys, basis[, coeffs, flaglen])	Initilize a system trajectory object.
eval(tlist)	Return the state and input for a trajectory at a list of
	times.

#### eval (tlist)

Return the state and input for a trajectory at a list of times.

Evaluate the trajectory at a list of time points, returning the state and input vectors for the trajectory:

x, u = traj.eval(tlist)

**Parameters tlist** (1D array) – List of times to evaluate the trajectory.

#### Returns

- x (2D array) For each state, the values of the state at the given times.
- **u** (2D array) For each input, the values of the input at the given times.

# 6.4.2 Flat systems functions

<pre>point_to_point(sys, x0, u0, xf, uf, Tf[,])</pre>	Compute trajectory between an initial and final condi-
	tions.

**CHAPTER** 

SEVEN

### INPUT/OUTPUT SYSTEMS

The *iosys* module contains the InputOutputSystem class that represents (possibly nonlinear) input/output systems. The InputOutputSystem class is a general class that defines any continuous or discrete time dynamical system. Input/output systems can be simulated and also used to compute equilibrium points and linearizations.

## 7.1 Module usage

An input/output system is defined as a dynamical system that has a system state as well as inputs and outputs (either inputs or states can be empty). The dynamics of the system can be in continuous or discrete time. To simulate an input/output system, use the <code>input\_output\_response()</code> function:

```
t, y = input_output_response(io_sys, T, U, X0, params)
```

An input/output system can be linearized around an equilibrium point to obtain a <code>StateSpace</code> linear system. Use the <code>find\_eqpt()</code> function to obtain an equilibrium point and the <code>linearize()</code> function to linearize about that equilibrium point:

```
xeq, ueq = find_eqpt(io_sys, X0, U0)
ss_sys = linearize(io_sys, xeq, ueq)
```

Input/output systems can be created from state space LTI systems by using the LinearIOSystem class`:

```
io_sys = LinearIOSystem(ss_sys)
```

Nonlinear input/output systems can be created using the NonlinearIOSystem class, which requires the definition of an update function (for the right hand side of the differential or different equation) and and output function (computes the outputs from the state):

```
io_sys = NonlinearIOSystem(updfcn, outfcn, inputs=M, outputs=P, states=N)
```

More complex input/output systems can be constructed by using the InterconnectedSystem class, which allows a collection of input/output subsystems to be combined with internal connections between the subsystems and a set of overall system inputs and outputs that link to the subsystems:

```
steering = ct.InterconnectedSystem(
    (plant, controller), name='system',
    connections=(('controller.e', '-plant.y')),
    inplist=('controller.e'), inputs='r',
    outlist=('plant.y'), outputs='y')
```

Interconnected systems can also be created using block diagram manipulations such as the <code>series()</code>, <code>parallel()</code>, and <code>feedback()</code> functions. The <code>InputOutputSystem</code> class also supports various algebraic operations such as \*(series interconnection) and + (parallel interconnection).

## 7.2 Example

To illustrate the use of the input/output systems module, we create a model for a predator/prey system, following the notation and parameter values in FBS2e.

We begin by defining the dynamics of the system

```
import control
import numpy as np
import matplotlib.pyplot as plt
def predprey_rhs(t, x, u, params):
    # Parameter setup
    a = params.get('a', 3.2)
   b = params.get('b', 0.6)
   c = params.get('c', 50.)
   d = params.get('d', 0.56)
   k = params.get('k', 125)
   r = params.get('r', 1.6)
    # Map the states into local variable names
   H = x[0]
   L = x[1]
    # Compute the control action (only allow addition of food)
    u_0 = u \text{ if } u > 0 \text{ else } 0
    # Compute the discrete updates
    dH = (r + u_0) * H * (1 - H/k) - (a * H * L)/(c + H)
    dL = b * (a * H * L) / (c + H) - d * L
    return [dH, dL]
```

We now create an input/output system using these dynamics:

```
io_predprey = control.NonlinearIOSystem(
    predprey_rhs, None, inputs=('u'), outputs=('H', 'L'),
    states=('H', 'L'), name='predprey')
```

Note that since we have not specified an output function, the entire state will be used as the output of the system.

The *io\_predprey* system can now be simulated to obtain the open loop dynamics of the system:

```
X0 = [25, 20]  # Initial H, L
T = np.linspace(0, 70, 500)  # Simulation 70 years of time

# Simulate the system
t, y = control.input_output_response(io_predprey, T, 0, X0)

# Plot the response
plt.figure(1)
plt.plot(t, y[0])
plt.plot(t, y[1])
plt.legend(['Hare', 'Lynx'])
plt.show(block=False)
```

We can also create a feedback controller to stabilize a desired population of the system. We begin by finding the (unstable) equilibrium point for the system and computing the linearization about that point.

```
eqpt = control.find_eqpt(io_predprey, X0, 0)
xeq = eqpt[0]  # choose the nonzero equilibrium point
lin_predprey = control.linearize(io_predprey, xeq, 0)
```

We next compute a controller that stabilizes the equilibrium point using eigenvalue placement and computing the feedforward gain using the number of lynxes as the desired output (following FBS2e, Example 7.5):

```
K = control.place(lin_predprey.A, lin_predprey.B, [-0.1, -0.2])
A, B = lin_predprey.A, lin_predprey.B
C = np.array([[0, 1]]) # regulated output = number of lynxes
kf = -1/(C @ np.linalg.inv(A - B @ K) @ B)
```

To construct the control law, we build a simple input/output system that applies a corrective input based on deviations from the equilibrium point. This system has no dynamics, since it is a static (affine) map, and can constructed using the ~control.ios.NonlinearIOSystem class:

```
io_controller = control.NonlinearIOSystem(
None,
lambda t, x, u, params: -K @ (u[1:] - xeq) + kf * (u[0] - xeq[1]),
inputs=('Ld', 'u1', 'u2'), outputs=1, name='control')
```

The input to the controller is u, consisting of the vector of hare and lynx populations followed by the desired lynx population.

To connect the controller to the predatory-prey model, we create an *InterconnectedSystem*:

```
io_closed = control.InterconnectedSystem(
  (io_predprey, io_controller),  # systems
  connections=(
        ('predprey.u', 'control.y[0]'),
        ('control.u1', 'predprey.H'),
        ('control.u2', 'predprey.L')
    ),
    inplist=('control.Ld'),
    outlist=('predprey.H', 'predprey.L', 'control.y[0]')
)
```

Finally, we simulate the closed loop system:

```
# Simulate the system
t, y = control.input_output_response(io_closed, T, 30, [15, 20])

# Plot the response
plt.figure(2)
plt.subplot(2, 1, 1)
plt.plot(t, y[0])
plt.plot(t, y[1])
plt.legend(['Hare', 'Lynx'])
plt.subplot(2, 1, 2)
plt.plot(t, y[2])
plt.legend(['input'])
plt.show(block=False)
```

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# 7.3 Module classes and functions

## 7.3.1 Input/output system classes

<pre>InputOutputSystem([inputs, outputs, states,])</pre>	A class for representing input/output systems.
InterconnectedSystem(syslist[, connections,	Interconnection of a set of input/output systems.
])	
LinearIOSystem(linsys[, inputs, outputs,])	Input/output representation of a linear (state space) sys-
	tem.
NonlinearIOSystem(updfcn[, outfcn, inputs,])	Nonlinear I/O system.

# 7.3.2 Input/output system functions

<pre>find_eqpt(sys, x0[, u0, y0, t, params, iu,])</pre>	Find the equilibrium point for an input/output system.
<pre>linearize(sys, xeq[, ueq, t, params])</pre>	Linearize an input/output system at a given state and in-
	put.
input_output_response(sys, T[, U, X0,])	Compute the output response of a system to a given in-
	put.
ss2io(*args, **kw)	Create an I/O system from a state space linear system.
tf2io(*args, **kw)	Convert a transfer function into an I/O system

**CHAPTER** 

**EIGHT** 

### **EXAMPLES**

The source code for the examples below are available in the *examples/* subdirecory of the source code distribution. The can also be accessed online via the [python-control GitHub repository](https://github.com/python-control/python-control/tree/master/examples).

## 8.1 Python scripts

The following Python scripts document the use of a variety of methods in the Python Control Toolbox on examples drawn from standard control textbooks and other sources.

### 8.1.1 Secord order system (MATLAB module example)

This example computes time and frequency responses for a second-order system using the MATLAB compatibility module.

#### Code

```
# secord.py - demonstrate some standard MATLAB commands
   # RMM, 25 May 09
2
   import os
   import matplotlib.pyplot as plt # MATLAB plotting functions
   from control.matlab import * # MATLAB-like functions
   # Parameters defining the system
   m = 250.0 # system mass
   k = 40.0
                      # spring constant
11
   b = 60.0
                      # damping constant
12
   # System matrices
13
   A = [[0, 1.], [-k/m, -b/m]]
14
   B = [[0], [1/m]]
15
   C = [[1., 0]]
   sys = ss(A, B, C, 0)
17
18
   # Step response for the system
19
   plt.figure(1)
20
   yout, T = step(sys)
  plt.plot(T.T, yout.T)
  plt.show(block=False)
```

```
24
   # Bode plot for the system
25
   plt.figure(2)
26
   mag, phase, om = bode(sys, logspace(-2, 2), Plot=True)
27
   plt.show(block=False)
28
29
   # Nyquist plot for the system
30
   plt.figure(3)
31
   nyquist(sys, logspace(-2, 2))
32
   plt.show(block=False)
33
   # Root lcous plot for the system
   rlocus(sys)
37
   if 'PYCONTROL_TEST_EXAMPLES' not in os.environ:
38
       plt.show()
```

#### **Notes**

1. The environment variable PYCONTROL\_TEST\_EXAMPLES is used for testing to turn off plotting of the outputs.

## 8.1.2 Inner/outer control design for vertical takeoff and landing aircraft

This script demonstrates the use of the python-control package for analysis and design of a controller for a vectored thrust aircraft model that is used as a running example through the text Feedback Systems by Astrom and Murray. This example makes use of MATLAB compatible commands.

#### Code

```
# pvtol-nested.py - inner/outer design for vectored thrust aircraft
2
   # RMM, 5 Sep 09
   # This file works through a fairly complicated control design and
   # analysis, corresponding to the planar vertical takeoff and landing
   # (PVTOL) aircraft in Astrom and Murray, Chapter 11. It is intended
   # to demonstrate the basic functionality of the python-control
   # package.
10
   from __future__ import print_function
11
12
   import os
13
   import matplotlib.pyplot as plt # MATLAB plotting functions
14
   from control.matlab import * # MATLAB-like functions
15
   import numpy as np
   # System parameters
18
   m = 4
                     # mass of aircraft
19
   J = 0.0475
                       # inertia around pitch axis
20
                       # distance to center of force
   r = 0.25
21
   g = 9.8
                       # gravitational constant
22
   c = 0.05
                       # damping factor (estimated)
```

```
24
   # Transfer functions for dynamics
25
   Pi = tf([r], [J, 0, 0]) # inner loop (roll)
26
   Po = tf([1], [m, c, 0]) # outer loop (position)
27
28
29
   # Inner loop control design
30
31
   # This is the controller for the pitch dynamics. Goal is to have
32
   # fast response for the pitch dynamics so that we can use this as a
33
   # control for the lateral dynamics
34
37
   # Design a simple lead controller for the system
   k, a, b = 200, 2, 50
38
   Ci = k*tf([1, a], [1, b]) # lead compensator
39
   Li = Pi*Ci
40
41
   # Bode plot for the open loop process
42
43
   plt.figure(1)
   bode (Pi)
44
45
   # Bode plot for the loop transfer function, with margins
46
   plt.figure(2)
47
   bode(Li)
   # Compute out the gain and phase margins
50
   #! Not implemented
51
   # gm, pm, wcg, wcp = margin(Li)
52
53
   # Compute the sensitivity and complementary sensitivity functions
54
   Si = feedback(1, Li)
55
   Ti = Li * Si
56
57
   # Check to make sure that the specification is met
58
   plt.figure(3)
59
   gangof4(Pi, Ci)
60
   # Compute out the actual transfer function from u1 to v1 (see L8.2 notes)
63
   # Hi = Ci * (1-m*q*Pi)/(1+Ci*Pi)
   Hi = parallel(feedback(Ci, Pi), -m*g*feedback(Ci*Pi, 1))
64
65
   plt.figure(4)
66
   plt.clf()
67
   plt.subplot(221)
   bode (Hi)
69
70
   # Now design the lateral control system
71
   a, b, K = 0.02, 5, 2
72.
   Co = -K*tf([1, 0.3], [1, 10]) # another lead compensator
73
74
   Lo = -m*q*Po*Co
75
   plt.figure(5)
76
77
   bode(Lo) # margin(Lo)
78
   # Finally compute the real outer-loop loop gain + responses
79
   L = Co*Hi*Po
```

```
S = feedback(1, L)
81
   T = feedback(L, 1)
82
83
    # Compute stability margins
84
   gm, pm, wgc, wpc = margin(L)
85
   print("Gain margin: %g at %g" % (gm, wgc))
86
   print("Phase margin: %g at %g" % (pm, wpc))
87
88
   plt.figure(6)
89
   plt.clf()
   bode (L, np.logspace (-4, 3))
91
    # Add crossover line to the magnitude plot
94
    # Note: in matplotlib before v2.1, the following code worked:
95
96
       plt.subplot(211); hold(True);
97
        loglog([1e-4, 1e3], [1, 1], 'k-')
    # In later versions of matplotlib the call to plt.subplot will clear the
100
    # axes and so we have to extract the axes that we want to use by hand.
101
    # In addition, hold() is deprecated so we no longer require it.
102
103
   for ax in plt.gcf().axes:
104
        if ax.get_label() == 'control-bode-magnitude':
105
106
            break
   ax.semilogx([1e-4, 1e3], 20*np.log10([1, 1]), 'k-')
107
108
109
    # Replot phase starting at -90 degrees
110
111
112
    # Get the phase plot axes
   for ax in plt.gcf().axes:
113
        if ax.get_label() == 'control-bode-phase':
114
            break
115
116
    # Recreate the frequency response and shift the phase
117
   mag, phase, w = freqresp(L, np.logspace(-4, 3))
   phase = phase - 360
120
   # Replot the phase by hand
121
   ax.semilogx([1e-4, 1e3], [-180, -180], 'k-')
122
   ax.semilogx(w, np.squeeze(phase), 'b-')
123
   ax.axis([1e-4, 1e3, -360, 0])
124
   plt.xlabel('Frequency [deg]')
   plt.ylabel('Phase [deg]')
126
   # plt.set(gca, 'YTick', [-360, -270, -180, -90, 0])
127
   # plt.set(gca, 'XTick', [10^-4, 10^-2, 1, 100])
128
129
130
   # Nyquist plot for complete design
131
132
133
   plt.figure(7)
   plt.clf()
134
   nyquist(L, (0.0001, 1000))
135
   plt.axis([-700, 5300, -3000, 3000])
136
137
```

```
# Add a box in the region we are going to expand
138
   plt.plot([-400, -400, 200, 200, -400], [-100, 100, 100, -100, -100], 'r-')
139
140
   # Expanded region
141
   plt.figure(8)
142
   plt.clf()
143
   plt.subplot(231)
144
   nyquist(L)
145
   plt.axis([-10, 5, -20, 20])
146
147
   # set up the color
   color = 'b'
150
   # Add arrows to the plot
151
   \# H1 = L.evalfr(0.4); H2 = L.evalfr(0.41);
152
    # arrow([real(H1), imag(H1)], [real(H2), imag(H2)], AM_normal_arrowsize, \
153
    # 'EdgeColor', color, 'FaceColor', color);
154
155
    \# H1 = freqresp(L, 0.35); H2 = freqresp(L, 0.36);
156
    # arrow([real(H2), -imag(H2)], [real(H1), -imag(H1)], AM_normal_arrowsize, \
157
    # 'EdgeColor', color, 'FaceColor', color);
158
159
   plt.figure(9)
160
   Yvec, Tvec = step(T, np.linspace(0, 20))
   plt.plot(Tvec.T, Yvec.T)
162
163
   Yvec, Tvec = step(Co*S, np.linspace(0, 20))
164
   plt.plot(Tvec.T, Yvec.T)
165
166
   plt.figure(10)
167
   plt.clf()
   P, Z = pzmap(T, plot=True, grid=True)
169
   print ("Closed loop poles and zeros: ", P, Z)
170
171
   # Gang of Four
172
   plt.figure(11)
173
   plt.clf()
174
   gangof4(Hi*Po, Co)
177
   if 'PYCONTROL_TEST_EXAMPLES' not in os.environ:
        plt.show()
178
```

#### **Notes**

- 1. Importing *print\_function* from \_\_*future*\_\_ in line 11 is only required if using Python 2.7.
- 2. The environment variable PYCONTROL\_TEST\_EXAMPLES is used for testing to turn off plotting of the outputs.

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## 8.1.3 LQR control design for vertical takeoff and landing aircraft

This script demonstrates the use of the python-control package for analysis and design of a controller for a vectored thrust aircraft model that is used as a running example through the text Feedback Systems by Astrom and Murray. This example makes use of MATLAB compatible commands.

#### Code

```
# pvtol_lqr.m - LQR design for vectored thrust aircraft
   # RMM, 14 Jan 03
2
   # This file works through an LQR based design problem, using the
   # planar vertical takeoff and landing (PVTOL) aircraft example from
   # Astrom and Murray, Chapter 5. It is intended to demonstrate the
   # basic functionality of the python-control package.
   import os
10
   import numpy as np
11
   import matplotlib.pyplot as plt # MATLAB plotting functions
12
   from control.matlab import * # MATLAB-like functions
13
14
15
   # System dynamics
   # These are the dynamics for the PVTOL system, written in state space
18
   # form.
19
20
21
   # System parameters
22
   m = 4 # mass of aircraft
23
   J = 0.0475 # inertia around pitch axis
24
              # distance to center of force
   r = 0.25
25
   g = 9.8
               # gravitational constant
26
               # damping factor (estimated)
27
   c = 0.05
28
   # State space dynamics
   xe = [0, 0, 0, 0, 0, 0] # equilibrium point of interest
   ue = [0, m*q] # (note these are lists, not matrices)
31
32
   # TODO: The following objects need converting from np.matrix to np.array
33
   # This will involve re-working the subsequent equations as the shapes
34
   # See below.
35
36
   # Dynamics matrix (use matrix type so that * works for multiplication)
37
   A = np.matrix(
38
       [[0, 0, 0, 1, 0, 0],
39
        [0, 0, 0, 0, 1, 0],
40
41
        [0, 0, 0, 0, 0, 1],
        [0, 0, (-ue[0]*np.sin(xe[2]) - ue[1]*np.cos(xe[2]))/m, -c/m, 0, 0],
42
        [0, 0, (ue[0]*np.cos(xe[2]) - ue[1]*np.sin(xe[2]))/m, 0, -c/m, 0],
        [0, 0, 0, 0, 0, 0]]
44
45
46
   # Input matrix
47
   B = np.matrix(
```

```
[[0, 0], [0, 0], [0, 0],
49
         [np.cos(xe[2])/m, -np.sin(xe[2])/m],
50
         [np.sin(xe[2])/m, np.cos(xe[2])/m],
51
         [r/J, 0]]
52
53
54
    # Output matrix
55
   C = np.matrix([[1, 0, 0, 0, 0, 0], [0, 1, 0, 0, 0]])
56
   D = np.matrix([[0, 0], [0, 0]])
57
58
59
    # Construct inputs and outputs corresponding to steps in xy position
61
    # The vectors xd and yd correspond to the states that are the desired
62
    # equilibrium states for the system. The matrices Cx and Cy are the
63
    # corresponding outputs.
64
65
    # The way these vectors are used is to compute the closed loop system
66
    # dynamics as
67
68
       xdot = Ax + B u \Rightarrow xdot = (A-BK)x + K xd
69
          u = -K(x - xd) \qquad \qquad v = Cx
70
71
   # The closed loop dynamics can be simulated using the "step" command,
72
   \# with K*xd as the input vector (assumes that the "input" is unit size,
   # so that xd corresponds to the desired steady state.
75
76
   xd = np.matrix([[1], [0], [0], [0], [0], [0]])
77
   yd = np.matrix([[0], [1], [0], [0], [0], [0]])
78
79
80
    # Extract the relevant dynamics for use with SISO library
81
82
    # The current python-control library only supports SISO transfer
83
   # functions, so we have to modify some parts of the original MATLAB
84
   # code to extract out SISO systems. To do this, we define the 'lat' and
85
   # 'alt' index vectors to consist of the states that are are relevant
    # to the lateral (x) and vertical (y) dynamics.
88
89
   # Indices for the parts of the state that we want
90
   lat = (0, 2, 3, 5)
91
   alt = (1, 4)
92
    # Decoupled dynamics
94
   Ax = (A[lat, :])[:, lat] #! not sure why I have to do it this way
95
   Bx = B[lat, 0]
96
   Cx = C[0, lat]
97
   Dx = D[0, 0]
98
   Ay = (A[alt, :])[:, alt] # ! not sure why I have to do it this way
   By = B[alt, 1]
101
   Cv = C[1, alt]
102
   Dy = D[1, 1]
103
104
   # Label the plot
```

```
plt.clf()
106
   plt.suptitle("LQR controllers for vectored thrust aircraft (pvtol-lqr)")
107
108
109
    # LQR design
110
111
112
    # Start with a diagonal weighting
113
   Qx1 = np.diag([1, 1, 1, 1, 1, 1])
114
   Qu1a = np.diag([1, 1])
115
   K, X, E = lqr(A, B, Qx1, Qu1a)
   K1a = np.matrix(K)
118
   # Close the loop: xdot = Ax - B K (x-xd)
119
   # Note: python-control requires we do this 1 input at a time
120
   # H1a = ss(A-B*K1a, B*K1a*concatenate((xd, yd), axis=1), C, D);
121
   \# (T, Y) = step(H1a, T=np.linspace(0,10,100));
122
123
    # TODO: The following equations will need modifying when converting from np.matrix to.
124
    ⇔np.array
   # because the results and even intermediate calculations will be different with numpy.
125
    ⇔arrays
   # For example:
126
   #Bx = B[lat, 0]
127
   # Will need to be changed to:
   \# Bx = B[lat, 0].reshape(-1, 1)
   # (if we want it to have the same shape as before)
130
131
   # For reference, here is a list of the correct shapes of these objects:
132
   # A: (6, 6)
133
   # B: (6, 2)
134
135
   # C: (2, 6)
   # D: (2, 2)
136
   # xd: (6, 1)
137
   # yd: (6, 1)
138
   # Ax: (4, 4)
139
   # Bx: (4, 1)
   # Cx: (1, 4)
   # Dx: ()
143
   # Ay: (2, 2)
   # By: (2, 1)
144
   # Cy: (1, 2)
145
146
   # Step response for the first input
147
   H1ax = ss(Ax - Bx*K1a[0, lat], Bx*K1a[0, lat]*xd[lat, :], Cx, Dx)
148
   Yx, Tx = step(Hlax, T=np.linspace(0, 10, 100))
149
150
    # Step response for the second input
151
   H1ay = ss(Ay - By*K1a[1, alt], By*K1a[1, alt]*yd[alt, :], Cy, Dy)
152
   Yy, Ty = step(H1ay, T=np.linspace(0, 10, 100))
153
154
   plt.subplot(221)
155
   plt.title("Identity weights")
156
   # plt.plot(T, Y[:,1, 1], '-', T, Y[:,2, 2], '--')
157
   plt.plot(Tx.T, Yx.T, '-', Ty.T, Yy.T, '--')
158
   plt.plot([0, 10], [1, 1], 'k-')
159
160
```

```
plt.axis([0, 10, -0.1, 1.4])
161
    plt.ylabel('position')
162
    plt.legend(('x', 'y'), loc='lower right')
163
164
    # Look at different input weightings
    Qula = np.diag([1, 1])
166
    K1a, X, E = lqr(A, B, Qx1, Qu1a)
167
    H1ax = ss(Ax - Bx*K1a[0, lat], Bx*K1a[0, lat]*xd[lat, :], Cx, Dx)
168
169
170
    Qu1b = (40 ** 2)*np.diag([1, 1])
    K1b, X, E = lqr(A, B, Qx1, Qu1b)
171
    H1bx = ss(Ax - Bx*K1b[0, lat], Bx*K1b[0, lat]*xd[lat, :], Cx, Dx)
172
173
   Qu1c = (200 ** 2)*np.diag([1, 1])
174
   K1c, X, E = lqr(A, B, Qx1, Qu1c)
175
    H1cx = ss(Ax - Bx*K1c[0, lat], Bx*K1c[0, lat]*xd[lat, :], Cx, Dx)
176
177
    [Y1, T1] = step(H1ax, T=np.linspace(0, 10, 100))
178
    [Y2, T2] = step(H1bx, T=np.linspace(0, 10, 100))
179
    [Y3, T3] = step(H1cx, T=np.linspace(0, 10, 100))
180
181
   plt.subplot(222)
182
   plt.title("Effect of input weights")
183
   plt.plot(T1.T, Y1.T, 'b-')
   plt.plot(T2.T, Y2.T, 'b-')
   plt.plot(T3.T, Y3.T, 'b-')
187
   plt.plot([0, 10], [1, 1], 'k-')
188
   plt.axis([0, 10, -0.1, 1.4])
189
190
    # arcarrow([1.3, 0.8], [5, 0.45], -6)
191
192
    plt.text(5.3, 0.4, 'rho')
193
    # Output weighting - change Qx to use outputs
194
    Ox2 = C.T*C
195
    Qu2 = 0.1*np.diag([1, 1])
196
    K, X, E = lqr(A, B, Qx2, Qu2)
    K2 = np.matrix(K)
   H2x = ss(Ax - Bx*K2[0, lat], Bx*K2[0, lat]*xd[lat, :], Cx, Dx)
200
   H2y = ss(Ay - By*K2[1, alt], By*K2[1, alt]*yd[alt, :], Cy, Dy)
201
202
   plt.subplot(223)
203
   plt.title("Output weighting")
204
    [Y2x, T2x] = step(H2x, T=np.linspace(0, 10, 100))
    [Y2y, T2y] = step(H2y, T=np.linspace(0, 10, 100))
206
    plt.plot(T2x.T, Y2x.T, T2y.T, Y2y.T)
207
   plt.ylabel('position')
208
   plt.xlabel('time')
209
   plt.ylabel('position')
   plt.legend(('x', 'y'), loc='lower right')
212
213
214
    # Physically motivated weighting
215
    # Shoot for 1 cm error in x, 10 cm error in y. Try to keep the angle
216
    # less than 5 degrees in making the adjustments. Penalize side forces
```

```
# due to loss in efficiency.
218
219
220
    Qx3 = np.diag([100, 10, 2*np.pi/5, 0, 0, 0])
221
    Qu3 = 0.1*np.diag([1, 10])
222
    (K, X, E) = lqr(A, B, Qx3, Qu3)
223
   K3 = np.matrix(K)
224
225
   H3x = ss(Ax - Bx*K3[0, lat], Bx*K3[0, lat]*xd[lat, :], Cx, Dx)
226
   H3y = ss(Ay - By*K3[1, alt], By*K3[1, alt]*yd[alt, :], Cy, Dy)
227
   plt.subplot(224)
   # step(H3x, H3y, 10)
229
   [Y3x, T3x] = step(H3x, T=np.linspace(0, 10, 100))
   [Y3y, T3y] = step(H3y, T=np.linspace(0, 10, 100))
   plt.plot(T3x.T, Y3x.T, T3y.T, Y3y.T)
232
   plt.title("Physically motivated weights")
233
   plt.xlabel('time')
234
   plt.legend(('x', 'y'), loc='lower right')
235
   if 'PYCONTROL_TEST_EXAMPLES' not in os.environ:
237
        plt.show()
238
```

#### **Notes**

1. The environment variable *PYCONTROL\_TEST\_EXAMPLES* is used for testing to turn off plotting of the outputs.

## 8.1.4 Balanced model reduction examples

#### Code

```
#!/usr/bin/env python
   import os
   import numpy as np
   import control.modelsimp as msimp
   import control.matlab as mt
   from control.statesp import StateSpace
   import matplotlib.pyplot as plt
10
   plt.close('all')
11
12
   # controllable canonical realization computed in MATLAB for the
13
   # transfer function: num = [1 11 45 32], den = [1 15 60 200 60]
   A = np.array([
15
       [-15., -7.5, -6.25, -1.875],
       [8., 0., 0., 0.],
17
       [0., 4., 0., 0.],
18
       [0., 0., 1., 0.]
19
   ])
20
   B = np.array([
21
       [2.],
22
        [0.],
```

```
[0.],
24
       [0.]
25
26
   ])
   C = np.array([[0.5, 0.6875, 0.7031, 0.5]])
27
   D = np.array([[0.]])
28
29
   # The full system
30
   fsys = StateSpace(A, B, C, D)
31
32
   # The reduced system, truncating the order by 1
33
34
   rsys = msimp.balred(fsys, n, method='truncate')
   # Comparison of the step responses of the full and reduced systems
37
   plt.figure(1)
38
   y, t = mt.step(fsys)
39
   yr, tr = mt.step(rsys)
40
   plt.plot(t.T, y.T)
   plt.plot(tr.T, yr.T)
42
43
   # Repeat balanced reduction, now with 100-dimensional random state space
44
   sysrand = mt.rss(100, 1, 1)
45
   rsysrand = msimp.balred(sysrand, 10, method='truncate')
46
47
   \# Comparison of the impulse responses of the full and reduced random systems
   plt.figure(2)
   yrand, trand = mt.impulse(sysrand)
50
   yrandr, trandr = mt.impulse(rsysrand)
51
  plt.plot(trand.T, yrand.T, trandr.T, yrandr.T)
52
53
   if 'PYCONTROL_TEST_EXAMPLES' not in os.environ:
54
       plt.show()
```

#### **Notes**

1. The environment variable PYCONTROL\_TEST\_EXAMPLES is used for testing to turn off plotting of the outputs.

## 8.1.5 Phase plot examples

#### Code

```
# phaseplots.py - examples of phase portraits
# RMM, 24 July 2011
# This file contains examples of phase portraits pulled from "Feedback
# Systems" by Astrom and Murray (Princeton University Press, 2008).

import os

import numpy as np
import matplotlib.pyplot as plt
from control.phaseplot import phase_plot
from numpy import pi
```

(continues on next page)

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```
13
   # Clear out any figures that are present
14
   plt.close('all')
15
16
17
   # Inverted pendulum
18
19
20
   # Define the ODEs for a damped (inverted) pendulum
21
   def invpend_ode(x, t, m=1., l=1., b=0.2, g=1):
22
       return x[1], -b/m*x[1] + (g*1/m)*np.sin(x[0])
23
   # Set up the figure the way we want it to look
26
   plt.figure()
27
   plt.clf()
28
   plt.axis([-2*pi, 2*pi, -2.1, 2.1])
29
   plt.title('Inverted pendulum')
31
   # Outer trajectories
32
   phase_plot(
33
       invpend_ode,
34
       X0=[[-2*pi, 1.6], [-2*pi, 0.5], [-1.8, 2.1],
35
            [-1, 2.1], [4.2, 2.1], [5, 2.1],
36
            [2*pi, -1.6], [2*pi, -0.5], [1.8, -2.1],
37
            [1, -2.1], [-4.2, -2.1], [-5, -2.1]],
       T=np.linspace(0, 40, 200),
39
       logtime=(3, 0.7)
40
41
42
43
   # Separatrices
   phase_plot(invpend_ode, X0=[[-2.3056, 2.1], [2.3056, -2.1]], T=6, lingrid=0)
45
46
   # Systems of ODEs: damped oscillator example (simulation + phase portrait)
47
48
49
   def oscillator_ode(x, t, m=1., b=1, k=1):
51
       return x[1], -k/m*x[0] - b/m*x[1]
52
53
   # Generate a vector plot for the damped oscillator
54
   plt.figure()
55
   plt.clf()
56
   phase_plot(oscillator_ode, [-1, 1, 10], [-1, 1, 10], 0.15)
57
   #plt.plot([0], [0], '.')
58
   # a=gca; set(a, 'FontSize', 20); set(a, 'DataAspectRatio', [1,1,1])
59
   plt.xlabel('$x_1$')
60
   plt.ylabel('$x_2$')
61
   plt.title('Damped oscillator, vector field')
62
   # Generate a phase plot for the damped oscillator
   plt.figure()
65
   plt.clf()
66
   plt.axis([-1, 1, -1, 1]) # set(gca, 'DataAspectRatio', [1, 1, 1]);
67
68
   phase_plot(
       oscillator_ode,
```

(continues on next page)

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```
X0 = [
70
             [-1, 1], [-0.3, 1], [0, 1], [0.25, 1], [0.5, 1], [0.75, 1], [1, 1],
71
             [1, -1], [0.3, -1], [0, -1], [-0.25, -1], [-0.5, -1], [-0.75, -1], [-1, -1]
72
73
        T=np.linspace(0, 8, 80),
        timepts=[0.25, 0.8, 2, 3]
75
76
    plt.plot([0], [0], 'k.') # 'MarkerSize', AM_data_markersize*3)
77
    # set(gca, 'DataAspectRatio', [1,1,1])
78
    plt.xlabel('$x_1$')
    plt.ylabel('$x_2$')
    plt.title('Damped oscillator, vector field and stream lines')
83
    # Stability definitions
84
85
    # This set of plots illustrates the various types of equilibrium points.
86
87
88
89
    def saddle_ode(x, t):
90
        """Saddle point vector field"""
91
        return x[0] - 3*x[1], -3*x[0] + x[1]
92
93
95
    # Asy stable
   m = 1
96
   b = 1
   k = 1 # default values
98
   plt.figure()
   plt.clf()
100
101
    plt.axis([-1, 1, -1, 1]) # set(gca, 'DataAspectRatio', [1 1 1]);
   phase_plot(
102
        oscillator_ode,
103
        X0 = [
104
            [-1, 1], [-0.3, 1], [0, 1], [0.25, 1], [0.5, 1], [0.7, 1], [1, 1], [1.3, 1],
105
            [1, -1], [0.3, -1], [0, -1], [-0.25, -1], [-0.5, -1], [-0.7, -1], [-1, -1],
106
107
            [-1.3, -1]
        ],
        T=np.linspace(0, 10, 100),
109
        timepts=[0.3, 1, 2, 3],
110
        parms=(m, b, k)
111
112
    plt.plot([0], [0], 'k.') # 'MarkerSize', AM_data_markersize*3)
113
    # plt.set(gca, 'FontSize', 16)
    plt.xlabel('$x_1$')
115
   plt.ylabel('$x_2$')
116
   plt.title('Asymptotically stable point')
117
118
    # Saddle
119
   plt.figure()
120
   plt.clf()
   plt.axis([-1, 1, -1, 1]) # set(gca, 'DataAspectRatio', [1 1 1])
122
   phase_plot(
123
        saddle_ode,
124
125
        scale=2,
126
        timepts=[0.2, 0.5, 0.8],
```

(continues on next page)

159

```
X0 = [
127
            [-1, -1], [1, 1],
128
            [-1, -0.95], [-1, -0.9], [-1, -0.8], [-1, -0.6], [-1, -0.4], [-1, -0.2],
129
            [-0.95, -1], [-0.9, -1], [-0.8, -1], [-0.6, -1], [-0.4, -1], [-0.2, -1],
130
            [1, 0.95], [1, 0.9], [1, 0.8], [1, 0.6], [1, 0.4], [1, 0.2],
131
            [0.95, 1], [0.9, 1], [0.8, 1], [0.6, 1], [0.4, 1], [0.2, 1],
132
            [-0.5, -0.45], [-0.45, -0.5], [0.5, 0.45], [0.45, 0.5],
133
            [-0.04, 0.04], [0.04, -0.04]
134
        ],
135
        T=np.linspace(0, 2, 20)
136
137
   plt.plot([0], [0], 'k.') # 'MarkerSize', AM_data_markersize*3)
138
139
   # set(gca, 'FontSize', 16)
   plt.xlabel('$x_1$')
140
   plt.ylabel('$x_2$')
141
   plt.title('Saddle point')
142
143
   # Stable isL
144
   m = 1
145
   b = 0
146
   k = 1 # zero damping
147
   plt.figure()
148
   plt.clf()
149
   plt.axis([-1, 1, -1, 1]) # set(gca, 'DataAspectRatio', [1 1 1]);
150
   phase_plot(
151
152
        oscillator_ode,
        timepts=[pi/6, pi/3, pi/2, 2*pi/3, 5*pi/6, pi, 7*pi/6,
153
                 4*pi/3, 9*pi/6, 5*pi/3, 11*pi/6, 2*pi],
154
        X0 = [[0.2, 0], [0.4, 0], [0.6, 0], [0.8, 0], [1, 0], [1.2, 0], [1.4, 0]],
155
        T=np.linspace(0, 20, 200),
156
157
        parms=(m, b, k)
158
   plt.plot([0], [0], 'k.') # 'MarkerSize', AM_data_markersize*3)
159
   # plt.set(gca, 'FontSize', 16)
160
   plt.xlabel('$x_1$')
161
   plt.ylabel('$x_2$')
162
   plt.title('Undamped system\nLyapunov stable, not asympt. stable')
163
   if 'PYCONTROL_TEST_EXAMPLES' not in os.environ:
        plt.show()
166
```

#### **Notes**

1. The environment variable PYCONTROL\_TEST\_EXAMPLES is used for testing to turn off plotting of the outputs.

## 8.1.6 SISO robust control example (SP96, Example 2.1)

#### Code

```
"""robust_siso.py
2
   Demonstrate mixed-sensitivity H-infinity design for a SISO plant.
3
   Based on Example 2.11 from Multivariable Feedback Control, Skogestad
   and Postlethwaite, 1st Edition.
   import os
10
   import numpy as np
11
   import matplotlib.pyplot as plt
12
13
   from control import tf, mixsyn, feedback, step_response
14
15
   s = tf([1, 0], 1)
16
   # the plant
17
   g = 200/(10*s + 1) / (0.05*s + 1)**2
   # disturbance plant
   gd = 100/(10*s + 1)
21
   # first design
22
   # sensitivity weighting
23
   M = 1.5
24
   wb = 10
25
   A = 1e-4
   ws1 = (s/M + wb) / (s + wb*A)
27
   # KS weighting
28
   wu = tf(1, 1)
29
30
   k1, cl1, infol = mixsyn(g, ws1, wu)
31
32
33
   # sensitivity (S) and complementary sensitivity (T) functions for
   # design 1
34
   s1 = feedback(1, q*k1)
35
   t1 = feedback(g*k1, 1)
36
37
   # second design
38
   # this weighting differs from the text, where A**0.5 is used; if you use that,
   # the frequency response doesn't match the figure. The time responses
   # are similar, though.
41
   ws2 = (s/M ** 0.5 + wb) **2 / (s + wb*A) **2
42.
   # the KS weighting is the same as for the first design
43
44
45
   k2, c12, info2 = mixsyn(g, ws2, wu)
47
   # S and T for design 2
   s2 = feedback(1, q*k2)
48
   t2 = feedback(g*k2, 1)
49
50
51
   # frequency response
   omega = np.logspace(-2, 2, 101)
   ws1mag, _, _ = ws1.freqresp(omega)
```

(continues on next page)

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```
s1mag, _, _ = s1.freqresp(omega)
54
   ws2mag, _, _ = ws2.freqresp(omega)
55
   s2mag, _, _ = s2.freqresp(omega)
56
57
   plt.figure(1)
   # text uses log-scaled absolute, but dB are probably more familiar to most control,
59
    →engineers
   plt.semilogx(omega, 20*np.log10(s1mag.flat), label='$S_1$')
60
   plt.semilogx(omega, 20*np.log10(s2mag.flat), label='$S_2$')
61
   # -1 in logspace is inverse
62
   plt.semilogx(omega, -20*np.log10(ws1mag.flat), label='$1/w_{p1}$')
   plt.semilogx(omega, -20*np.log10(ws2mag.flat), label='$1/w_{P2}$')
   plt.ylim([-80, 10])
66
   plt.xlim([1e-2, 1e2])
67
   plt.xlabel('freq [rad/s]')
68
   plt.ylabel('mag [dB]')
   plt.legend()
   plt.title('Sensitivity and sensitivity weighting frequency responses')
71
72
   # time response
73
   time = np.linspace(0, 3, 201)
74
   _, y1 = step_response(t1, time)
75
   _, y2 = step_response(t2, time)
76
   # gd injects into the output (that is, g and gd are summed), and the
   # closed loop mapping from output disturbance->output is S.
79
   _, y1d = step_response(s1*gd, time)
80
   _, y2d = step_response(s2*gd, time)
81
82
83
   plt.figure(2)
84
   plt.subplot(1, 2, 1)
   plt.plot(time, y1, label='$y_1(t)')
85
   plt.plot(time, y2, label='$y_2(t)$')
86
87
   plt.ylim([-0.1, 1.5])
88
   plt.xlim([0, 3])
89
   plt.xlabel('time [s]')
   plt.ylabel('signal [1]')
   plt.legend()
92
   plt.title('Tracking response')
93
   plt.subplot(1, 2, 2)
95
   plt.plot(time, yld, label='$y_1(t)$')
   plt.plot(time, y2d, label='\$y_2(t)$')
   plt.ylim([-0.1, 1.5])
99
   plt.xlim([0, 3])
100
   plt.xlabel('time [s]')
101
   plt.ylabel('signal [1]')
102
   plt.legend()
   plt.title('Disturbance response')
105
   if 'PYCONTROL TEST EXAMPLES' not in os.environ:
106
107
       plt.show()
```

#### **Notes**

1. The environment variable PYCONTROL\_TEST\_EXAMPLES is used for testing to turn off plotting of the outputs.

### 8.1.7 MIMO robust control example (SP96, Example 3.8)

#### Code

```
"""robust_mimo.py
   Demonstrate mixed-sensitivity H-infinity design for a MIMO plant.
   Based on Example 3.8 from Multivariable Feedback Control, Skogestad and Postlethwaite,
   → 1st Edition.
6
   import os
   import numpy as np
10
   import matplotlib.pyplot as plt
11
12
   from control import tf, ss, mixsyn, step_response
13
14
15
   def weighting(wb, m, a):
16
        """weighting(wb, m, a) -> wf
17
       wb - design frequency (where |wf| is approximately 1)
18
       m - high frequency gain of 1/wf; should be > 1
19
       a - low frequency gain of 1/wf; should be < 1
20
21
       wf - SISO LTI object
       n n n
22
       s = tf([1, 0], [1])
23
       return (s/m + wb) / (s + wb*a)
24
25
26
   def plant():
27
       """plant() -> g
28
       g - LTI object; 2x2 plant with a RHP zero, at s=0.5.
29
30
       den = [0.2, 1.2, 1]
31
       gtf = tf([[[1], [1]],
32
                  [[2, 1], [2]]],
33
34
                 [[den, den],
                  [den, den]])
35
       return ss(gtf)
36
37
38
   # as of this writing (2017-07-01), python-control doesn't have an
39
   # equivalent to Matlab's sigma function, so use a trivial stand-in.
40
   def triv_sigma(g, w):
41
       """triv_sigma(g, w) \rightarrow s
42
       g - LTI object, order n
43
       w - frequencies, length m
44
       s - (m,n) array of singular values of g(1j*w)"""
45
       m, p, _ = g.freqresp(w)
46
47
       sjw = (m*np.exp(1j*p)).transpose(2, 0, 1)
```

(continues on next page)

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```
sv = np.linalg.svd(sjw, compute_uv=False)
48
        return sv
49
50
51
   def analysis():
52
        """Plot open-loop responses for various inputs"""
53
        g = plant()
54
55
        t = np.linspace(0, 10, 101)
56
        _, yu1 = step_response(g, t, input=0)
57
        _, yu2 = step_response(g, t, input=1)
58
60
        yu1 = yu1
        yu2 = yu2
61
62.
        # linear system, so scale and sum previous results to get the
63
        # [1,-1] response
64
        yuz = yu1 - yu2
65
66
        plt.figure(1)
67
        plt.subplot(1, 3, 1)
68
        plt.plot(t, yu1[0], label='$y_1$')
69
        plt.plot(t, yu1[1], label='$y_2$')
70
        plt.xlabel('time')
71
        plt.ylabel('output')
        plt.ylim([-1.1, 2.1])
74
        plt.legend()
        plt.title('o/l response\nto input [1,0]')
75
76
        plt.subplot(1, 3, 2)
77
78
        plt.plot(t, yu2[0], label='$y_1$')
79
        plt.plot(t, yu2[1], label='$y_2$')
        plt.xlabel('time')
80
        plt.ylabel('output')
81
        plt.ylim([-1.1, 2.1])
82
        plt.legend()
83
84
        plt.title('o/l response\nto input [0,1]')
86
        plt.subplot(1, 3, 3)
87
        plt.plot(t, yuz[0], label='$y_1$')
        plt.plot(t, yuz[1], label='$y_2$')
88
        plt.xlabel('time')
89
        plt.ylabel('output')
90
        plt.ylim([-1.1, 2.1])
91
92
        plt.legend()
        plt.title('o/l response\nto input [1,-1]')
93
94
95
   def synth(wb1, wb2):
96
        """synth(wb1, wb2) -> k, gamma
97
        wb1: S weighting frequency
        wb2: KS weighting frequency
        k: controller
100
        gamma: H-infinity norm of 'design', that is, of evaluation system
101
        with loop closed through design
102
        n n n
103
        g = plant()
```

```
wu = ss([], [], np.eye(2))
105
        wp1 = ss(weighting(wb=wb1, m=1.5, a=1e-4))
106
        wp2 = ss(weighting(wb=wb2, m=1.5, a=1e-4))
107
        wp = wp1.append(wp2)
108
        k, _, info = mixsyn(g, wp, wu)
109
        return k, info[0]
110
111
112
    def step_opposite(g, t):
113
        """reponse to step of [-1,1]"""
114
        _, yu1 = step_response(g, t, input=0)
115
        _, yu2 = step_response(g, t, input=1)
116
117
        return yu1 - yu2
118
119
    def design():
120
        """Show results of designs"""
121
        # equal weighting on each output
122
        k1, gam1 = synth(0.25, 0.25)
123
        # increase "bandwidth" of output 2 by moving crossover weighting frequency 100...
124
    →times higher
        k2, gam2 = synth(0.25, 25)
125
        # now weight output 1 more heavily
126
127
        # won't plot this one, just want gamma
        _{-}, gam3 = synth(25, 0.25)
128
129
        print('design 1 gamma {:.3g} (Skogestad: 2.80)'.format(gam1))
130
        print('design 2 gamma {:.3g} (Skogestad: 2.92)'.format(gam2))
131
        print('design 3 gamma {:.3g} (Skogestad: 6.73)'.format(gam3))
132
133
        # do the designs
134
135
        g = plant()
        w = np.logspace(-2, 2, 101)
136
        I = ss([], [], np.eye(2))
137
        s1 = I.feedback(g*k1)
138
        s2 = I.feedback(g*k2)
139
140
        # frequency response
141
        sv1 = triv_sigma(s1, w)
        sv2 = triv_sigma(s2, w)
143
144
        plt.figure(2)
145
146
        plt.subplot(1, 2, 1)
147
        plt.semilogx(w, 20*np.log10(sv1[:, 0]), label=r'$\sigma_1(S_1)$')
148
        plt.semilogx(w, 20*np.log10(sv1[:, 1]), label=r'$\sigma_2(S_1)$')
149
        plt.semilogx(w, 20*np.log10(sv2[:, 0]), label=r'$\sigma_1(S_2)$')
150
        plt.semilogx(w, 20*np.log10(sv2[:, 1]), label=r'$\sigma_2(S_2)$')
151
        plt.ylim([-60, 10])
152
        plt.ylabel('magnitude [dB]')
153
154
        plt.xlim([1e-2, 1e2])
        plt.xlabel('freq [rad/s]')
155
        plt.legend()
156
        plt.title('Singular values of S')
157
158
159
        # time response
```

```
# in design 1, both outputs have an inverse initial response; in
161
        # design 2, output 2 does not, and is very fast, while output 1
162
        # has a larger initial inverse response than in design 1
163
        time = np.linspace(0, 10, 301)
        t1 = (g*k1).feedback(I)
165
        t2 = (g*k2).feedback(I)
166
167
        y1 = step_opposite(t1, time)
168
        y2 = step\_opposite(t2, time)
169
170
        plt.subplot(1, 2, 2)
171
        plt.plot(time, y1[0], label='des. 1 $y_1(t))$')
172
173
        plt.plot(time, y1[1], label='des. 1 $y_2(t))$')
        plt.plot(time, y2[0], label='des. 2 $y_1(t))$')
174
        plt.plot(time, y2[1], label='des. 2 $y_2(t))$')
175
        plt.xlabel('time [s]')
176
        plt.ylabel('response [1]')
177
        plt.legend()
178
        plt.title('c/l response to reference [1,-1]')
179
180
181
    if __name__ == "__main__":
182
183
        analysis()
        design()
184
        if 'PYCONTROL_TEST_EXAMPLES' not in os.environ:
185
            plt.show()
```

#### **Notes**

1. The environment variable PYCONTROL\_TEST\_EXAMPLES is used for testing to turn off plotting of the outputs.

### 8.1.8 Cruise control design example (as a nonlinear I/O system)

#### Code

```
# cruise-control.py - Cruise control example from FBS
   # RMM, 16 May 2019
2
3
   # The cruise control system of a car is a common feedback system encountered
   # in everyday life. The system attempts to maintain a constant velocity in the
   # presence of disturbances primarily caused by changes in the slope of a
   # road. The controller compensates for these unknowns by measuring the speed
   # of the car and adjusting the throttle appropriately.
   # This file explore the dynamics and control of the cruise control system,
   # following the material presenting in Feedback Systems by Astrom and Murray.
   # A full nonlinear model of the vehicle dynamics is used, with both PI and
12
   # state space control laws. Different methods of constructing control systems
13
   # are show, all using the InputOutputSystem class (and subclasses).
14
  import numpy as np
16
  import matplotlib.pyplot as plt
17
   from math import pi
```

```
import control as ct
19
20
21
   # Section 4.1: Cruise control modeling and control
22
23
24
   # Vehicle model: vehicle()
25
26
   # To develop a mathematical model we start with a force balance for
27
   \# the car body. Let v be the speed of the car, m the total mass
28
   # (including passengers), F the force generated by the contact of the
   # wheels with the road, and Fd the disturbance force due to gravity,
31
   # friction, and aerodynamic drag.
32
   def vehicle_update(t, x, u, params={}):
33
        """Vehicle dynamics for cruise control system.
34
35
       Parameters
36
37
       x : array
38
             System state: car velocity in m/s
39
       u : array
40
            System input: [throttle, gear, road_slope], where throttle is
41
             a float between 0 and 1, gear is an integer between 1 and 5,
42
             and road_slope is in rad.
43
44
45
       Returns
46
       float
47
           Vehicle acceleration
48
49
50
       from math import copysign, sin
51
       sign = lambda x: copysign(1, x)
                                                  # define the sign() function
52
53
       # Set up the system parameters
54
       m = params.get('m', 1600.)
55
       g = params.get('g', 9.8)
57
       Cr = params.get('Cr', 0.01)
58
       Cd = params.get('Cd', 0.32)
       rho = params.get('rho', 1.3)
59
       A = params.get('A', 2.4)
60
       alpha = params.get(
61
            'alpha', [40, 25, 16, 12, 10])
                                               # gear ratio / wheel radius
62
63
        # Define variables for vehicle state and inputs
64
                                             # vehicle velocity
       v = x[0]
65
       throttle = np.clip(u[0], 0, 1)
                                             # vehicle throttle
66
                                             # vehicle gear
       gear = u[1]
67
       theta = u[2]
                                             # road slope
68
        # Force generated by the engine
71
       omega = alpha[int(gear)-1] * v
                                             # engine angular speed
72
       F = alpha[int(gear)-1] * motor_torque(omega, params) * throttle
73
74
        # Disturbance forces
```

```
76
        # The disturbance force Fd has three major components: Fq, the forces due
77
        # to gravity; Fr, the forces due to rolling friction; and Fa, the
78
        # aerodynamic drag.
79
80
        # Letting the slope of the road be \theta (theta), gravity gives the
81
        # force Fg = m g sin \theta.
82
83
        Fg = m * g * sin(theta)
84
85
        \# A simple model of rolling friction is Fr = m \ g \ Cr \ sgn(v), where Cr is
86
        # the coefficient of rolling friction and sgn(v) is the sign of v (+/- 1) or
87
88
        \# zero if v = 0.
89
        Fr = m * q * Cr * sign(v)
90
91
        # The aerodynamic drag is proportional to the square of the speed: Fa =
92
        \# 1/\rho Cd A |v| v, where \rho is the density of air, Cd is the
93
        # shape-dependent aerodynamic drag coefficient, and A is the frontal area
94
        # of the car.
95
96
        Fa = 1/2 * rho * Cd * A * abs(v) * v
97
98
        # Final acceleration on the car
99
        Fd = Fg + Fr + Fa
100
101
        dv = (F - Fd) / m
102
        return dv
103
104
    # Engine model: motor_torque
105
106
    # The force F is generated by the engine, whose torque is proportional to
107
    # the rate of fuel injection, which is itself proportional to a control
108
    \# signal 0 <= u <= 1 that controls the throttle position. The torque also
109
    # depends on engine speed omega.
110
111
   def motor_torque(omega, params={}):
112
113
        # Set up the system parameters
       Tm = params.get('Tm', 190.)
                                                   # engine torque constant
       omega_m = params.get('omega_m', 420.)
                                                   # peak engine angular speed
115
       beta = params.get('beta', 0.4)
                                                   # peak engine rolloff
116
117
        return np.clip(Tm * (1 - beta * (omega/omega_m - 1) **2), 0, None)
118
119
120
    # Define the input/output system for the vehicle
    vehicle = ct.NonlinearIOSystem(
121
        vehicle_update, None, name='vehicle',
122
        inputs = ('u', 'gear', 'theta'), outputs = ('v'), states=('v'))
123
124
   # Figure 1.11: A feedback system for controlling the speed of a vehicle. In
125
   # this example, the speed of the vehicle is measured and compared to the
   # desired speed. The controller is a PI controller represented as a transfer
   # function. In the textbook, the simulations are done for LTI systems, but
128
   # here we simulate the full nonlinear system.
129
130
   # Construct a PI controller with rolloff, as a transfer function
131
   Kp = 0.5
                                      # proportional gain
132
```

```
Ki = 0.1
133
                                       # integral gain
    control tf = ct.tf2io(
134
        ct.TransferFunction([Kp, Ki], [1, 0.01*Ki/Kp]),
135
        name='control', inputs='u', outputs='y')
136
137
    # Construct the closed loop control system
138
    # Inputs: vref, gear, theta
139
    # Outputs: v (vehicle velocity)
140
    cruise_tf = ct.InterconnectedSystem(
141
        (control_tf, vehicle), name='cruise',
142
        connections = (
143
            ('control.u', '-vehicle.v'),
145
             ('vehicle.u', 'control.y')),
        inplist = ('control.u', 'vehicle.gear', 'vehicle.theta'),
146
        inputs = ('vref', 'gear', 'theta'),
147
        outlist = ('vehicle.v', 'vehicle.u'),
148
        outputs = ('v', 'u'))
149
150
    # Define the time and input vectors
151
    T = np.linspace(0, 25, 101)
152
    vref = 20 * np.ones(T.shape)
153
    gear = 4 * np.ones(T.shape)
154
    theta0 = np.zeros(T.shape)
155
156
    # Now simulate the effect of a hill at t = 5 seconds
157
158
   plt.figure()
   plt.suptitle('Response to change in road slope')
159
    vel_axes = plt.subplot(2, 1, 1)
160
    inp_axes = plt.subplot(2, 1, 2)
161
    theta_hill = np.array([
162
        0 if t <= 5 else
163
164
        4./180. * pi * (t-5) if t <= 6 else
        4./180. * pi for t in T])
165
166
    for m in (1200, 1600, 2000):
167
        # Compute the equilibrium state for the system
168
169
        X0, U0 = ct.find_eqpt(
170
            cruise_tf, [0, vref[0]], [vref[0], gear[0], theta0[0]],
171
            iu=[1, 2], y0=[vref[0], 0], iy=[0], params={'m':m})
172
        t, y = ct.input_output_response(
173
            cruise_tf, T, [vref, gear, theta_hill], X0, params={'m':m})
174
175
        # Plot the velocity
176
177
        plt.sca(vel_axes)
        plt.plot(t, y[0])
178
179
        # Plot the input
180
        plt.sca(inp_axes)
181
182
        plt.plot(t, y[1])
    # Add labels to the plots
184
   plt.sca(vel_axes)
185
   plt.vlabel('Speed [m/s]')
186
   plt.legend(['m = 1000 kg', 'm = 2000 kg', 'm = 3000 kg'], frameon=False)
187
188
   plt.sca(inp_axes)
```

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169

```
plt.ylabel('Throttle')
190
    plt.xlabel('Time [s]')
191
192
    # Figure 4.2: Torque curves for a typical car engine. The graph on the
193
    # left shows the torque generated by the engine as a function of the
    # angular velocity of the engine, while the curve on the right shows
195
    # torque as a function of car speed for different gears.
196
197
   plt.figure()
198
    plt.suptitle('Torque curves for typical car engine')
199
200
    # Figure 4.2a - single torque curve as function of omega
    omega_range = np.linspace(0, 700, 701)
   plt.subplot(2, 2, 1)
203
   plt.plot(omega_range, [motor_torque(w) for w in omega_range])
204
   plt.xlabel('Angular velocity $\omega$ [rad/s]')
205
   plt.ylabel('Torque $T$ [Nm]')
206
   plt.grid(True, linestyle='dotted')
207
208
    # Figure 4.2b - torque curves in different gears, as function of velocity
209
    plt.subplot(2, 2, 2)
210
    v_range = np.linspace(0, 70, 71)
211
   alpha = [40, 25, 16, 12, 10]
212
   for gear in range(5):
213
        omega_range = alpha[gear] * v_range
214
215
        plt.plot(v_range, [motor_torque(w) for w in omega_range],
                 color='blue', linestyle='solid')
216
217
    # Set up the axes and style
218
   plt.axis([0, 70, 100, 200])
219
   plt.grid(True, linestyle='dotted')
220
    # Add labels
222
   plt.text(11.5, 120, '$n$=1')
223
   plt.text(24, 120, '$n$=2')
224
   plt.text(42.5, 120, '$n$=3')
225
   plt.text(58.5, 120, '$n$=4')
   plt.text(58.5, 185, '$n$=5')
   plt.xlabel('Velocity $v$ [m/s]')
229
   plt.ylabel('Torque $T$ [Nm]')
230
   plt.show(block=False)
231
232
    # Figure 4.3: Car with cruise control encountering a sloping road
233
    # PI controller model: control_pi()
235
236
    # We add to this model a feedback controller that attempts to regulate the
237
    # speed of the car in the presence of disturbances. We shall use a
238
    # proportional-integral controller
239
240
    def pi_update(t, x, u, params={}):
241
        # Get the controller parameters that we need
242
        ki = params.get('ki', 0.1)
243
        kaw = params.get('kaw', 2)
244
                                     # anti-windup gain
245
        # Assign variables for inputs and states (for readability)
```

```
v = u[0]
                                      # current velocity
247
        vref = u[1]
                                      # reference velocity
248
        z = x[0]
                                      # integrated error
249
250
        # Compute the nominal controller output (needed for anti-windup)
251
        u_a = pi_output(t, x, u, params)
252
253
        # Compute anti-windup compensation (scale by ki to account for structure)
254
        u_aw = kaw/ki * (np.clip(u_a, 0, 1) - u_a) if ki != 0 else 0
255
256
        # State is the integrated error, minus anti-windup compensation
257
        return (vref - v) + u_aw
258
259
    def pi_output(t, x, u, params={}):
260
        # Get the controller parameters that we need
261
        kp = params.get('kp', 0.5)
262
        ki = params.get('ki', 0.1)
263
264
        # Assign variables for inputs and states (for readability)
265
        v = u[0]
                                      # current velocity
266
        vref = u[1]
                                      # reference velocity
267
        z = x[0]
                                      # integrated error
268
269
        # PI controller
270
        return kp * (vref - v) + ki * z
271
272
    control_pi = ct.NonlinearIOSystem(
273
        pi_update, pi_output, name='control',
274
        inputs = ['v', 'vref'], outputs = ['u'], states = ['z'],
275
        params = \{'kp':0.5, 'ki':0.1\})
276
277
278
    # Create the closed loop system
    cruise_pi = ct.InterconnectedSystem(
279
        (vehicle, control_pi), name='cruise',
280
        connections=(
281
            ('vehicle.u', 'control.u'),
282
            ('control.v', 'vehicle.v')),
283
        inplist=('control.vref', 'vehicle.gear', 'vehicle.theta'),
284
        outlist=('control.u', 'vehicle.v'), outputs=['u', 'v'])
286
    # Figure 4.3b shows the response of the closed loop system. The figure shows
287
    # that even if the hill is so steep that the throttle changes from 0.17 to
288
    # almost full throttle, the largest speed error is less than 1 m/s, and the
289
    # desired velocity is recovered after 20 s.
290
    # Define a function for creating a "standard" cruise control plot
292
    def cruise_plot(sys, t, y, t_hill=5, vref=20, antiwindup=False,
293
                     linetype='b-', subplots=[None, None]):
294
        # Figure out the plot bounds and indices
295
        v_min = vref-1.2; v_max = vref+0.5; v_ind = sys.find_output('v')
296
        u_min = 0; u_max = 2 if antiwindup else 1; u_ind = sys.find_output('u')
297
        # Make sure the upper and lower bounds on v are OK
299
        while max(v[v ind]) > v max: v max += 1
300
        while min(y[v_ind]) < v_min: v_min -= 1
301
302
        # Create arrays for return values
```

(continues on next page)

171

```
subplot_axes = list(subplots)
304
305
        # Velocity profile
306
        if subplot_axes[0] is None:
307
            subplot_axes[0] = plt.subplot(2, 1, 1)
        else:
            plt.sca(subplots[0])
310
        plt.plot(t, y[v_ind], linetype)
311
        plt.plot(t, vref*np.ones(t.shape), 'k-')
312
        plt.plot([t_hill, t_hill], [v_min, v_max], 'k--')
313
        plt.axis([0, t[-1], v_min, v_max])
314
        plt.xlabel('Time $t$ [s]')
315
        plt.ylabel('Velocity $v$ [m/s]')
316
317
        # Commanded input profile
318
        if subplot_axes[1] is None:
319
            subplot_axes[1] = plt.subplot(2, 1, 2)
320
        else:
321
            plt.sca(subplots[1])
322
        plt.plot(t, y[u_ind], 'r--' if antiwindup else linetype)
323
        plt.plot([t_hill, t_hill], [u_min, u_max], 'k--')
324
        plt.axis([0, t[-1], u_min, u_max])
325
        plt.xlabel('Time $t$ [s]')
326
        plt.ylabel('Throttle $u$')
327
328
329
        # Applied input profile
        if antiwindup:
330
            # TODO: plot the actual signal from the process?
331
            plt.plot(t, np.clip(y[u_ind], 0, 1), linetype)
332
            plt.legend(['Commanded', 'Applied'], frameon=False)
333
334
335
        return subplot_axes
336
    # Define the time and input vectors
337
    T = np.linspace(0, 30, 101)
338
    vref = 20 * np.ones(T.shape)
339
    gear = 4 * np.ones(T.shape)
    theta0 = np.zeros(T.shape)
    \# Compute the equilibrium throttle setting for the desired speed (solve for x
343
    # and u given the gear, slope, and desired output velocity)
344
    X0, U0, Y0 = ct.find_eqpt(
345
        cruise_pi, [vref[0], 0], [vref[0], gear[0], theta0[0]],
346
        y0=[0, vref[0]], iu=[1, 2], iy=[1], return_y=True)
347
348
349
    # Now simulate the effect of a hill at t = 5 seconds
    plt.figure()
350
    plt.suptitle('Car with cruise control encountering sloping road')
351
    theta_hill = [
352
        0 if t <= 5 else
353
354
        4./180. * pi * (t-5) if t <= 6 else
        4./180. * pi for t in T]
    t, y = ct.input_output_response(cruise_pi, T, [vref, gear, theta_hill], X0)
356
    cruise_plot(cruise_pi, t, y)
357
358
359
    # Example 7.8: State space feedback with integral action
```

```
361
362
    # State space controller model: control_sf_ia()
363
364
    # Construct a state space controller with integral action, linearized around
    # an equilibrium point. The controller is constructed around the equilibrium
    # point (x_d, u_d) and includes both feedforward and feedback compensation.
367
368
    # Controller inputs: (x, y, r)
                                       system states, system output, reference
369
    # Controller state: z
                                        integrated error (y - r)
370
    # Controller output: u
                                        state feedback control
371
    # Note: to make the structure of the controller more clear, we implement this
    # as a "nonlinear" input/output module, even though the actual input/output
374
    # system is linear. This also allows the use of parameters to set the
375
    # operating point and gains for the controller.
376
377
    def sf_update(t, z, u, params={}):
378
        y, r = u[1], u[2]
379
        return y - r
380
381
    def sf_output(t, z, u, params={}):
382
        # Get the controller parameters that we need
383
        K = params.get('K', 0)
384
        ki = params.get('ki', 0)
386
        kf = params.get('kf', 0)
        xd = params.get('xd', 0)
387
        vd = params.get('vd', 0)
388
        ud = params.get('ud', 0)
389
390
        # Get the system state and reference input
391
        x, y, r = u[0], u[1], u[2]
393
        return ud - K * (x - xd) - ki * z + kf * (r - yd)
394
395
    # Create the input/output system for the controller
396
    control_sf = ct.NonlinearIOSystem(
397
        sf_update, sf_output, name='control',
        inputs=('x', 'y', 'r'),
        outputs=('u'),
400
        states=('z'))
401
402
    # Create the closed loop system for the state space controller
403
    cruise_sf = ct.InterconnectedSystem(
404
        (vehicle, control_sf), name='cruise',
        connections=(
406
            ('vehicle.u', 'control.u'),
407
            ('control.x', 'vehicle.v'),
408
            ('control.y', 'vehicle.v')),
409
        inplist=('control.r', 'vehicle.gear', 'vehicle.theta'),
410
        outlist=('control.u', 'vehicle.v'), outputs=['u', 'v'])
411
412
    # Compute the linearization of the dynamics around the equilibrium point
413
414
    # YO represents the steady state with PI control => we can use it to
415
    # identify the steady state velocity and required throttle setting.
416
   xd = Y0[1]
```

```
ud = Y0[0]
418
   yd = Y0[1]
419
420
    # Compute the linearized system at the eq pt
421
   cruise_linearized = ct.linearize(vehicle, xd, [ud, gear[0], 0])
422
    # Construct the gain matrices for the system
424
   A, B, C = cruise_linearized.A, cruise_linearized.B[0, 0], cruise_linearized.C
425
   K = 0.5
426
   kf = -1 / (C * np.linalg.inv(A - B * K) * B)
427
    # Response of the system with no integral feedback term
   plt.figure()
   plt.suptitle('Cruise control with proportional and PI control')
431
   theta hill = [
432
        0 if t <= 8 else
433
        4./180. * pi * (t-8) if t <= 9 else
434
        4./180. * pi for t in T]
435
    t, y = ct.input_output_response(
436
        cruise_sf, T, [vref, gear, theta_hill], [X0[0], 0],
437
        params={'K':K, 'kf':kf, 'ki':0.0, 'kf':kf, 'xd':xd, 'ud':ud, 'yd':yd})
438
   subplots = cruise_plot(cruise_sf, t, y, t_hill=8, linetype='b--')
439
440
    # Response of the system with state feedback + integral action
441
   t, y = ct.input_output_response(
442
        cruise_sf, T, [vref, gear, theta_hill], [X0[0], 0],
       params={'K':K, 'kf':kf, 'ki':0.1, 'kf':kf, 'xd':xd, 'ud':ud, 'yd':yd})
444
   cruise_plot(cruise_sf, t, y, t_hill=8, linetype='b-', subplots=subplots)
445
446
   # Add a legend
447
   plt.legend(['Proportional', 'PI control'], frameon=False)
448
    # Example 11.5: simulate the effect of a (steeper) hill at t = 5 seconds
450
451
    # The windup effect occurs when a car encounters a hill that is so steep (6
452
    # deg) that the throttle saturates when the cruise controller attempts to
453
   # maintain speed.
454
   plt.figure()
   plt.suptitle('Cruise control with integrator windup')
457
   T = np.linspace(0, 70, 101)
458
   vref = 20 * np.ones(T.shape)
459
   theta_hill = [
460
        0 if t <= 5 else
461
        6./180. * pi * (t-5) if t <= 6 else
        6./180. * pi for t in T]
463
    t, y = ct.input_output_response(
464
        cruise_pi, T, [vref, gear, theta_hill], X0,
465
        params={ 'kaw':0})
466
   cruise_plot(cruise_pi, t, y, antiwindup=True)
467
    # Example 11.6: add anti-windup compensation
470
   # Anti-windup can be applied to the system to improve the response. Because of
471
   # the feedback from the actuator model, the output of the integrator is
472
   # quickly reset to a value such that the controller output is at the
473
    # saturation limit.
```

```
475
   plt.figure()
476
   plt.suptitle('Cruise control with integrator anti-windup protection')
477
   t, y = ct.input_output_response(
478
        cruise_pi, T, [vref, gear, theta_hill], X0,
        params={ 'kaw':2.})
480
   cruise_plot(cruise_pi, t, y, antiwindup=True)
481
482
    # If running as a standalone program, show plots and wait before closing
483
   import os
484
   if __name__ == '__main__' and 'PYCONTROL_TEST_EXAMPLES' not in os.environ:
485
        plt.show()
   else:
        plt.show(block=False)
488
```

#### **Notes**

1. The environment variable PYCONTROL\_TEST\_EXAMPLES is used for testing to turn off plotting of the outputs.

### 8.1.9 Gain scheduled control for vehicle steering (I/O system)

#### Code

```
# steering-gainsched.py - gain scheduled control for vehicle steering
   # RMM, 8 May 2019
   # This file works through Example 1.1 in the "Optimization-Based Control"
   # course notes by Richard Murray (avaliable at http://fbsbook.org, in the
   # optimization-based control supplement). It is intended to demonstrate the
   # functionality for nonlinear input/output systems in the python-control
   # package.
   import numpy as np
10
   import control as ct
11
   from cmath import sgrt
12
   import matplotlib.pyplot as mpl
13
14
15
   # Vehicle steering dynamics
16
17
   # The vehicle dynamics are given by a simple bicycle model. We take the state
18
   \# of the system as (x, y, theta) where (x, y) is the position of the vehicle
19
   # in the plane and theta is the angle of the vehicle with respect to
20
   # horizontal. The vehicle input is given by (v, phi) where v is the forward
21
   # velocity of the vehicle and phi is the angle of the steering wheel. The
   # model includes saturation of the vehicle steering angle.
24
   # System state: x, y, theta
25
   # System input: v, phi
   # System output: x, y
27
   # System parameters: wheelbase, maxsteer
28
   def vehicle_update(t, x, u, params):
```

(continues on next page)

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```
# Get the parameters for the model
31
       1 = params.get('wheelbase', 3.)
                                                 # vehicle wheelbase
32
       phimax = params.get('maxsteer', 0.5)
                                                # max steering angle (rad)
33
34
       # Saturate the steering input
35
       phi = np.clip(u[1], -phimax, phimax)
36
37
       # Return the derivative of the state
38
       return np.array([
39
                                             # xdot = cos(theta) v
           np.cos(x[2]) * u[0],
40
                                             # ydot = sin(theta) v
41
           np.sin(x[2]) * u[0],
           (u[0] / 1) * np.tan(phi)
                                             # thdot = v/l tan(phi)
42
43
44
   def vehicle_output(t, x, u, params):
45
                                             # return x, y, theta (full state)
       return x
46
47
   # Define the vehicle steering dynamics as an input/output system
48
   vehicle = ct.NonlinearIOSystem(
49
       vehicle_update, vehicle_output, states=3, name='vehicle',
50
       inputs=('v', 'phi'),
51
       outputs=('x', 'y', 'theta'))
52
53
54
   # Gain scheduled controller
   # For this system we use a simple schedule on the forward vehicle velocity and
57
   # place the poles of the system at fixed values. The controller takes the
58
   # current vehicle position and orientation plus the velocity velocity as
59
   # inputs, and returns the velocity and steering commands.
60
61
62
   # System state: none
   # System input: ex, ey, etheta, vd, phid
63
   # System output: v, phi
64
   # System parameters: longpole, latpole1, latpole2
65
66
67
   def control_output(t, x, u, params):
68
       # Get the controller parameters
       longpole = params.get('longpole', -2.)
       latpole1 = params.get('latpole1', -1/2 + sgrt(-7)/2)
70
       latpole2 = params.get('latpole2', -1/2 - sqrt(-7)/2)
71
       1 = params.get('wheelbase', 3)
72
73
       # Extract the system inputs
74
75
       ex, ey, etheta, vd, phid = u
76
       # Determine the controller gains
77
       alpha1 = -np.real(latpole1 + latpole2)
78
       alpha2 = np.real(latpole1 * latpole2)
79
80
81
       # Compute and return the control law
       v = -longpole * ex # Note: no feedfwd (to make plot interesting)
82
       if vd != 0:
83
           phi = phid + (alpha1 * 1) / vd * ey + (alpha2 * 1) / vd * etheta
84
85
       else:
            # We aren't moving, so don't turn the steering wheel
86
           phi = phid
```

```
88
        return np.array([v, phi])
89
90
    # Define the controller as an input/output system
91
    controller = ct.NonlinearIOSystem(
92
        None, control_output, name='controller',
                                                     # static system
93
        inputs=('ex', 'ey', 'etheta', 'vd', 'phid'),
                                                          # system inputs
94
        outputs=('v', 'phi')
                                                           # system outputs
95
96
97
98
    # Reference trajectory subsystem
100
    # The reference trajectory block generates a simple trajectory for the system
101
    # given the desired speed (vref) and lateral position (yref). The trajectory
102
    # consists of a straight line of the form (vref * t, yref, 0) with nominal
103
    # input (vref, 0).
104
    # System state: none
106
    # System input: vref, yref
107
    # System output: xd, yd, thetad, vd, phid
108
    # System parameters: none
109
110
    def trajgen_output(t, x, u, params):
111
        vref, yref = u
112
113
        return np.array([vref * t, yref, 0, vref, 0])
114
    # Define the trajectory generator as an input/output system
115
    trajgen = ct.NonlinearIOSystem(
116
        None, trajgen_output, name='trajgen',
117
        inputs=('vref', 'yref'),
118
        outputs=('xd', 'yd', 'thetad', 'vd', 'phid'))
119
120
121
    # System construction
122
123
    # The input to the full closed loop system is the desired lateral position and
124
    # the desired forward velocity. The output for the system is taken as the
    # full vehicle state plus the velocity of the vehicle. The following diagram
127
    # summarizes the interconnections:
128
129
130
131
132
    # [ ] ---> trajgen -+-+-> controller -+-> vehicle -+-> [x, y, theta]
133
134
135
136
    # We construct the system using the InterconnectedSystem constructor and using
137
    # signal labels to keep track of everything.
    steering = ct.InterconnectedSystem(
140
141
        # List of subsystems
        (trajgen, controller, vehicle), name='steering',
142
143
        # Interconnections between subsystems
144
```

```
connections=(
145
            ('controller.ex', 'trajgen.xd', '-vehicle.x'),
146
            ('controller.ey', 'trajgen.yd', '-vehicle.y'),
147
            ('controller.etheta', 'trajgen.thetad', '-vehicle.theta'),
148
            ('controller.vd', 'trajgen.vd'),
149
            ('controller.phid', 'trajgen.phid'),
150
            ('vehicle.v', 'controller.v'),
151
            ('vehicle.phi', 'controller.phi')
152
        ),
153
154
        # System inputs
155
        inplist=['trajgen.vref', 'trajgen.yref'],
156
157
        inputs=['yref', 'vref'],
158
        # System outputs
159
        outlist=['vehicle.x', 'vehicle.y', 'vehicle.theta', 'controller.v',
160
                  'controller.phi'],
161
        outputs=['x', 'y', 'theta', 'v', 'phi']
162
163
164
    # Set up the simulation conditions
165
   yref = 1
166
   T = np.linspace(0, 5, 100)
167
   # Set up a figure for plotting the results
170
   mpl.figure();
171
   # Plot the reference trajectory for the y position
172
   mpl.plot([0, 5], [yref, yref], 'k--')
173
174
   # Find the signals we want to plot
175
   y_index = steering.find_output('y')
176
   v_index = steering.find_output('v')
177
178
    # Do an iteration through different speeds
179
   for vref in [8, 10, 12]:
180
181
        # Simulate the closed loop controller response
        tout, yout = ct.input_output_response(
182
            steering, T, [vref * np.ones(len(T)), yref * np.ones(len(T))])
184
        # Plot the reference speed
185
        mpl.plot([0, 5], [vref, vref], 'k--')
186
187
        # Plot the system output
188
        y_line, = mpl.plot(tout, yout[y_index, :], 'r') # lateral position
189
        v_line, = mpl.plot(tout, yout[v_index, :], 'b') # vehicle velocity
190
191
    # Add axis labels
192
   mpl.xlabel('Time (s)')
193
   mpl.ylabel('x vel (m/s), y pos (m)')
   mpl.legend((v_line, y_line), ('v', 'y'), loc='center right', frameon=False)
```

#### **Notes**

# 8.1.10 Differentially flat system - kinematic car

This example demonstrates the use of the *flatsys* module for generating trajectories for differentially flat systems. The example is drawn from Chapter 8 of FBS2e.

#### Code

```
# kincar-flatsys.py - differentially flat systems example
   # RMM, 3 Jul 2019
2
3
   # This example demonstrates the use of the `flatsys` module for generating
4
   # trajectories for differnetially flat systems by computing a trajectory for a
   # kinematic (bicycle) model of a car changing lanes.
   import os
   import numpy as np
   import matplotlib.pyplot as plt
10
   import control as ct
11
   import control.flatsys as fs
12
13
14
   # Function to take states, inputs and return the flat flag
15
   def vehicle_flat_forward(x, u, params={}):
16
       # Get the parameter values
17
       b = params.get('wheelbase', 3.)
18
19
       # Create a list of arrays to store the flat output and its derivatives
20
       zflag = [np.zeros(3), np.zeros(3)]
21
22
       # Flat output is the x, y position of the rear wheels
23
       zflag[0][0] = x[0]
24
       zflag[1][0] = x[1]
25
26
       # First derivatives of the flat output
27
       zflag[0][1] = u[0] * np.cos(x[2]) # dx/dt
28
       zflag[1][1] = u[0] * np.sin(x[2]) # dy/dt
29
30
       # First derivative of the angle
31
       thdot = (u[0]/b) * np.tan(u[1])
32
33
       # Second derivatives of the flat output (setting vdot = 0)
34
35
       zflag[0][2] = -u[0] * thdot * np.sin(x[2])
       zflag[1][2] = u[0] * thdot * np.cos(x[2])
36
37
       return zflag
38
39
40
   # Function to take the flat flag and return states, inputs
41
   def vehicle_flat_reverse(zflag, params={}):
42
       # Get the parameter values
43
       b = params.get('wheelbase', 3.)
44
45
       # Create a vector to store the state and inputs
46
       x = np.zeros(3)
```

(continues on next page)

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```
u = np.zeros(2)
48
49
        # Given the flat variables, solve for the state
50
        x[0] = zflag[0][0] # x position
51
        x[1] = zflag[1][0] # y position
52
        x[2] = np.arctan2(zflag[1][1], zflag[0][1]) # tan(theta) = ydot/xdot
53
54
        # And next solve for the inputs
55
        u[0] = zflag[0][1] * np.cos(x[2]) + zflag[1][1] * np.sin(x[2])
56
        thdot_v = zflag[1][2] * np.cos(x[2]) - zflag[0][2] * np.sin(x[2])
57
        u[1] = np.arctan2(thdot_v, u[0]**2 / b)
58
60
        return x, u
61
62
    # Function to compute the RHS of the system dynamics
63
   def vehicle_update(t, x, u, params):
64
        b = params.get('wheelbase', 3.)
                                                     # get parameter values
65
        dx = np.array([
66
            np.cos(x[2]) * u[0],
67
            np.sin(x[2]) * u[0],
68
            (u[0]/b) * np.tan(u[1])
69
70
        1)
71
        return dx
72
74
   # Create differentially flat input/output system
   vehicle flat = fs.FlatSystem(
75
        vehicle_flat_forward, vehicle_flat_reverse, vehicle_update,
76
        inputs=('v', 'delta'), outputs=('x', 'y', 'theta'),
77
        states=('x', 'y', 'theta'))
78
79
    # Define the endpoints of the trajectory
80
   x0 = [0., -2., 0.]; u0 = [10., 0.]
81
   xf = [40., 2., 0.]; uf = [10., 0.]
82
83
84
   # Define a set of basis functions to use for the trajectories
   poly = fs.PolyFamily(6)
87
   # Find a trajectory between the initial condition and the final condition
88
   traj = fs.point_to_point(vehicle_flat, x0, u0, xf, uf, Tf, basis=poly)
89
   # Create the desired trajectory between the initial and final condition
91
   T = np.linspace(0, Tf, 500)
92
   xd, ud = traj.eval(T)
93
   # Simulation the open system dynamics with the full input
95
   t, y, x = ct.input_output_response(
96
        vehicle_flat, T, ud, x0, return_x=True)
97
   # Plot the open loop system dynamics
   plt.figure()
100
   plt.suptitle("Open loop trajectory for kinematic car lane change")
101
102
   # Plot the trajectory in xy coordinates
103
   plt.subplot(4, 1, 2)
```

```
plt.plot(x[0], x[1])
105
   plt.xlabel('x [m]')
106
   plt.ylabel('y [m]')
107
   plt.axis([x0[0], xf[0], x0[1]-1, xf[1]+1])
    # Time traces of the state and input
   plt.subplot(2, 4, 5)
111
   plt.plot(t, x[1])
112
   plt.ylabel('y [m]')
113
   plt.subplot(2, 4, 6)
   plt.plot(t, x[2])
   plt.ylabel('theta [rad]')
118
   plt.subplot(2, 4, 7)
119
   plt.plot(t, ud[0])
120
   plt.xlabel('Time t [sec]')
121
   plt.ylabel('v [m/s]')
   plt.axis([0, Tf, u0[0] - 1, uf[0] + 1])
123
124
   plt.subplot(2, 4, 8)
125
   plt.plot(t, ud[1])
126
   plt.xlabel('Ttime t [sec]')
127
   plt.ylabel('$\delta$ [rad]')
   plt.tight_layout()
131
   # Show the results unless we are running in batch mode
   if 'PYCONTROL TEST EXAMPLES' not in os.environ:
132
        plt.show()
133
```

#### **Notes**

1. The environment variable *PYCONTROL\_TEST\_EXAMPLES* is used for testing to turn off plotting of the outputs.

# 8.2 Jupyter notebooks

The examples below use *python-control* in a Jupyter notebook environment. These notebooks demonstrate the use of modeling, analysis, and design tools using running examples in FBS2e.

## 8.2.1 Cruise control

Richard M. Murray and Karl J. Åström 17 Jun 2019

The cruise control system of a car is a common feedback system encountered in everyday life. The system attempts to maintain a constant velocity in the presence of disturbances primarily caused by changes in the slope of a road. The controller compensates for these unknowns by measuring the speed of the car and adjusting the throttle appropriately.

This notebook explores the dynamics and control of the cruise control system, following the material presenting in Feedback Systems by Astrom and Murray. A nonlinear model of the vehicle dynamics is used, with both state space and frequency domain control laws. The process model is presented in Section 1, and a controller based on state

feedback is discussed in Section 2, where we also add integral action to the controller. In Section 3 we explore the behavior with PI control including the effect of actuator saturation and how it is avoided by windup protection. Different methods of constructing control systems are shown, all using the InputOutputSystem class (and subclasses).

```
[1]: import numpy as np
import matplotlib.pyplot as plt
from math import pi
import control as ct
```

#### **Process Model**

#### **Vehicle Dynamics**

To develop a mathematical model we start with a force balance for the car body. Let v be the speed of the car, m the total mass (including passengers), F the force generated by the contact of the wheels with the road, and  $F_d$  the disturbance force due to gravity, friction, and aerodynamic drag.

```
[2]: def vehicle_update(t, x, u, params={}):
        """Vehicle dynamics for cruise control system.
        Parameters
        x : arrav
            System state: car velocity in m/s
        u : array
            System input: [throttle, gear, road_slope], where throttle is
             a float between 0 and 1, gear is an integer between 1 and 5,
            and road_slope is in rad.
        Returns
        float
           Vehicle acceleration
        from math import copysign, sin
        sign = lambda x: copysign(1, x)
                                               # define the sign() function
        # Set up the system parameters
       m = params.get('m', 1600.) # vehicle mass, kg
        g = params.get('g', 9.8)
                                             # gravitational constant, m/s^2
        Cr = params.get('Cr', 0.01)
                                             # coefficient of rolling friction
       Cd = params.get('Cd', 0.32)
                                             # drag coefficient
        rho = params.get('rho', 1.3)
                                             # density of air, kg/m^3
        A = params.get('A', 2.4)
                                              # car area, m^2
        alpha = params.get(
            'alpha', [40, 25, 16, 12, 10]) # gear ratio / wheel radius
        # Define variables for vehicle state and inputs
                                       # vehicle velocity
        v = x[0]
                                        # vehicle throttle
        throttle = np.clip(u[0], 0, 1)
                                        # vehicle gear
        gear = u[1]
                                         # road slope
        theta = u[2]
        # Force generated by the engine
```

```
omega = alpha[int(gear)-1] * v # engine angular speed
F = alpha[int(gear)-1] * motor_torque(omega, params) * throttle
# Disturbance forces
# The disturbance force Fd has three major components: Fg, the forces due
# to gravity; Fr, the forces due to rolling friction; and Fa, the
# aerodynamic drag.
# Letting the slope of the road be \theta (theta), gravity gives the
# force Fg = m g sin \theta.
Fg = m * g * sin(theta)
# A simple model of rolling friction is Fr = m \ q \ Cr \ sqn(v), where Cr \ is
# the coefficient of rolling friction and sgn(v) is the sign of v (\pm 1) or
\# zero if v = 0.
Fr = m * g * Cr * sign(v)
# The aerodynamic drag is proportional to the square of the speed: Fa =
# 1/2 \rho Cd A |v| v, where \rho is the density of air, Cd is the
# shape-dependent aerodynamic drag coefficient, and A is the frontal area
# of the car.
Fa = 1/2 * rho * Cd * A * abs(v) * v
# Final acceleration on the car
Fd = Fq + Fr + Fa
dv = (F - Fd) / m
return dv
```

#### **Engine model**

The force F is generated by the engine, whose torque is proportional to the rate of fuel injection, which is itself proportional to a control signal  $0 \le u \le 1$  that controls the throttle position. The torque also depends on engine speed omega.

```
[3]: def motor_torque(omega, params={}):
    # Set up the system parameters
    Tm = params.get('Tm', 190.) # engine torque constant
    omega_m = params.get('omega_m', 420.) # peak engine angular speed
    beta = params.get('beta', 0.4) # peak engine rolloff

return np.clip(Tm * (1 - beta * (omega/omega_m - 1)**2), 0, None)
```

Torque curves for a typical car engine. The graph on the left shows the torque generated by the engine as a function of the angular velocity of the engine, while the curve on the right shows torque as a function of car speed for different gears.

```
[4]: # Figure 4.2a - single torque curve as function of omega
omega_range = np.linspace(0, 700, 701)
plt.subplot(2, 2, 1)
(continues on next page)
```

```
plt.plot(omega_range, [motor_torque(w) for w in omega_range])
plt.xlabel('Angular velocity $\omega$ [rad/s]')
plt.ylabel('Torque $T$ [Nm]')
plt.grid(True, linestyle='dotted')
# Figure 4.2b - torque curves in different gears, as function of velocity
plt.subplot(2, 2, 2)
v_range = np.linspace(0, 70, 71)
alpha = [40, 25, 16, 12, 10]
for gear in range(5):
    omega_range = alpha[gear] * v_range
    plt.plot(v_range, [motor_torque(w) for w in omega_range],
             color='blue', linestyle='solid')
# Set up the axes and style
plt.axis([0, 70, 100, 200])
plt.grid(True, linestyle='dotted')
# Add labels
plt.text(11.5, 120, '$n$=1')
plt.text(24, 120, '$n$=2')
plt.text(42.5, 120, '$n$=3')
plt.text(58.5, 120, '$n$=4')
plt.text(58.5, 185, '$n$=5')
plt.xlabel('Velocity $v$ [m/s]')
plt.ylabel('Torque $T$ [Nm]')
plt.tight_layout()
plt.suptitle('Torque curves for typical car engine');
               Torque curves for typical car engine
                                                          n=5
   180
                                   180
Torque T [Nm]
   160
                                   160
                                   140
   140
                                   120
   120
                                   100
             200
                   400
                         600
                                             20
                                                   40
         Angular velocity ω [rad/s]
                                            Velocity v [m/s]
```

#### Input/ouput model for the vehicle system

We now create an input/output model for the vehicle system that takes the throttle input u, the gear and the angle of the road  $\theta$  as input. The output of this model is the current vehicle velocity v.

```
v_min = vref-1.2; v_max = vref+0.5; v_ind = sys.find_output('v')
u_min = 0; u_max = 2 if antiwindup else 1; u_ind = sys.find_output('u')
# Make sure the upper and lower bounds on v are OK
while max(y[v_ind]) > v_max: v_max += 1
while min(y[v_ind]) < v_min: v_min -= 1</pre>
# Create arrays for return values
subplot_axes = subplots.copy()
# Velocity profile
if subplot_axes[0] is None:
    subplot_axes[0] = plt.subplot(2, 1, 1)
else:
    plt.sca(subplots[0])
plt.plot(t, y[v_ind], linetype)
plt.plot(t, vref*np.ones(t.shape), 'k-')
plt.plot([t_hill, t_hill], [v_min, v_max], 'k--')
plt.axis([0, t[-1], v_min, v_max])
plt.xlabel('Time $t$ [s]')
plt.ylabel('Velocity $v$ [m/s]')
# Commanded input profile
if subplot_axes[1] is None:
    subplot_axes[1] = plt.subplot(2, 1, 2)
else:
    plt.sca(subplots[1])
plt.plot(t, y[u_ind], 'r--' if antiwindup else linetype)
plt.plot([t_hill, t_hill], [u_min, u_max], 'k--')
plt.axis([0, t[-1], u_min, u_max])
plt.xlabel('Time $t$ [s]')
plt.ylabel('Throttle $u$')
# Applied input profile
if antiwindup:
    plt.plot(t, np.clip(y[u_ind], 0, 1), linetype)
    plt.legend(['Commanded', 'Applied'], frameon=False)
return subplot_axes
```

#### State space controller

Construct a state space controller with integral action, linearized around an equilibrium point. The controller is constructed around the equilibrium point  $(x_d, u_d)$  and includes both feedforward and feedback compensation.

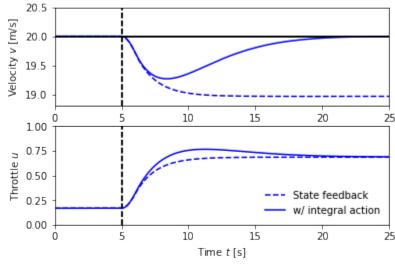
- Controller inputs (x, y, r): system states, system output, reference
- Controller state z: integrated error (y-r)
- Controller output u: state feedback control

Note: to make the structure of the controller more clear, we implement this as a "nonlinear" input/output module, even though the actual input/output system is linear. This also allows the use of parameters to set the operating point and gains for the controller.

```
[6]: def sf_update(t, z, u, params={}):
        y, r = u[1], u[2]
        return y - r
    def sf_output(t, z, u, params={}):
        # Get the controller parameters that we need
        K = params.get('K', 0)
        ki = params.get('ki', 0)
        kf = params.get('kf', 0)
        xd = params.get('xd', 0)
        yd = params.get('yd', 0)
        ud = params.get('ud', 0)
        # Get the system state and reference input
        x, y, r = u[0], u[1], u[2]
        return ud - K * (x - xd) - ki * z + kf * (r - yd)
    # Create the input/output system for the controller
    control_sf = ct.NonlinearIOSystem(
        sf_update, sf_output, name='control',
        inputs=('x', 'y', 'r'),
        outputs=('u'),
        states=('z'))
    # Create the closed loop system for the state space controller
    cruise_sf = ct.InterconnectedSystem(
        (vehicle, control_sf), name='cruise',
        connections=(
            ('vehicle.u', 'control.u'),
             ('control.x', 'vehicle.v'),
            ('control.y', 'vehicle.v')),
        inplist=('control.r', 'vehicle.gear', 'vehicle.theta'),
        outlist=('control.u', 'vehicle.v'), outputs=['u', 'v'])
    # Define the time and input vectors
    T = np.linspace(0, 25, 501)
    vref = 20 * np.ones(T.shape)
    gear = 4 * np.ones(T.shape)
    theta0 = np.zeros(T.shape)
    # Find the equilibrium point for the system
    Xeq, Ueq = ct.find_eqpt(
        vehicle, [vref[0]], [0, gear[0], theta0[0]], y0=[vref[0]], iu=[1, 2])
    print("Xeq = ", Xeq)
    print("Ueq = ", Ueq)
    # Compute the linearized system at the eq pt
    cruise_linearized = ct.linearize(vehicle, Xeq, [Ueq[0], gear[0], 0])
    Xeq = [20.]
    Ueq = [0.16874874 4.
```

```
[7]: # Construct the gain matrices for the system
A, B, C = cruise_linearized.A, cruise_linearized.B[0, 0], cruise_linearized.C
K = 0.5
kf = -1 / (C * np.linalg.inv(A - B * K) * B)
```

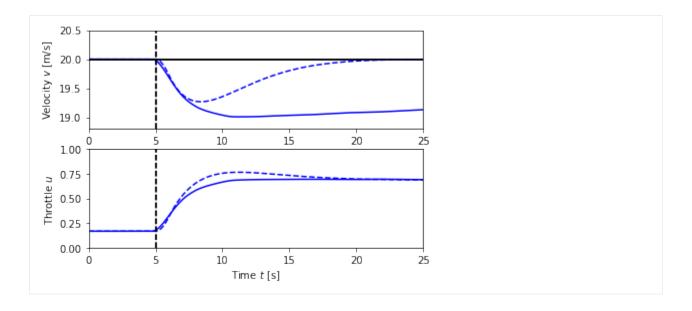
```
# Compute the steady state velocity and throttle setting
xd = Xeq[0]
ud = Ueq[0]
yd = vref[-1]
# Response of the system with no integral feedback term
plt.figure()
theta_hill = [
    0 if t <= 5 else
    4./180. * pi * (t-5) if t <= 6 else
   4./180. * pi for t in T]
t, y_sfb = ct.input_output_response(
   cruise_sf, T, [vref, gear, theta_hill], [Xeq[0], 0],
   params={'K':K, 'ki':0.0, 'kf':kf, 'xd':xd, 'ud':ud, 'yd':yd})
subplots = cruise_plot(cruise_sf, t, y_sfb, t_hill=5, linetype='b--')
# Response of the system with state feedback + integral action
t, y_sfb_int = ct.input_output_response(
    cruise_sf, T, [vref, gear, theta_hill], [Xeq[0], 0],
    params={'K':K, 'ki':0.1, 'kf':kf, 'xd':xd, 'ud':ud, 'yd':yd})
cruise_plot(cruise_sf, t, y_sfb_int, t_hill=5, linetype='b-', subplots=subplots)
# Add title and legend
plt.suptitle('Cruise control with state feedback, integral action')
import matplotlib.lines as mlines
p_line = mlines.Line2D([], [], color='blue', linestyle='--', label='State feedback')
pi_line = mlines.Line2D([], [], color='blue', linestyle='-', label='w/ integral action
plt.legend(handles=[p_line, pi_line], frameon=False, loc='lower right');
         Cruise control with state feedback, integral action
```



#### Pole/zero cancellation

The transfer function for the linearized dynamics of the cruise control system is given by P(s) = b/(s+a). A simple (but not necessarily good) way to design a PI controller is to choose the parameters of the PI controller as  $k_i = ak_p$ . The controller transfer function is then  $C(s) = k_p + k_i/s = k_i(s+a)/s$ . It has a zero at  $s = -k_i/k_p = -a$  that cancels the process pole at s = -a. We have  $P(s)C(s) = k_i/s$  giving the transfer function from reference to vehicle velocity as  $G_{yr}(s) = bk_p/(s+bk_p)$ , and control design is then simply a matter of choosing the gain  $k_p$ . The closed loop system dynamics are of first order with the time constant  $1/(bk_p)$ .

```
[8]: # Get the transfer function from throttle input + hill to vehicle speed
    P = ct.ss2tf(cruise_linearized[0, 0])
    # Construction a controller that cancels the pole
    kp = 0.5
    a = -P.pole()[0]
    b = np.real(P(0)) * a
    ki = a * kp
    C = ct.tf2ss(ct.TransferFunction([kp, ki], [1, 0]))
    control_pz = ct.LinearIOSystem(C, name='control', inputs='u', outputs='y')
    print("system: a = ", a, ", b = ", b)
    print("pzcancel: kp =", kp, ", ki =", ki, ", 1/(kp \ b) = ", 1/(kp \ * b))
    print("sfb_int: K = ", K, ", ki = 0.1")
    # Construct the closed loop system and plot the response
    # Create the closed loop system for the state space controller
    cruise_pz = ct.InterconnectedSystem(
        (vehicle, control_pz), name='cruise_pz',
        connections = (
            ('control.u', '-vehicle.v'),
             ('vehicle.u', 'control.y')),
        inplist = ('control.u', 'vehicle.gear', 'vehicle.theta'),
        inputs = ('vref', 'gear', 'theta'),
        outlist = ('vehicle.v', 'vehicle.u'),
        outputs = ('v', 'u'))
    # Find the equilibrium point
    X0, U0 = ct.find_eqpt(
        cruise_pz, [vref[0], 0], [vref[0], gear[0], theta0[0]],
        iu=[1, 2], y0=[vref[0], 0], iy=[0])
    # Response of the system with PI controller canceling process pole
    t, y_pzcancel = ct.input_output_response(
        cruise_pz, T, [vref, gear, theta_hill], X0)
    subplots = cruise_plot(cruise_pz, t, y_pzcancel, t_hill=5, linetype='b-')
    cruise_plot(cruise_sf, t, y_sfb_int, t_hill=5, linetype='b--', subplots=subplots);
    system: a = 0.010124405669387215, b = 1.3203061238159202
    pzcancel: kp = 0.5, ki = 0.005062202834693608, 1/(kp b) = 1.5148002148317266
    sfb_{int}: K = 0.5, ki = 0.1
```



## **PI Controller**

In this example, the speed of the vehicle is measured and compared to the desired speed. The controller is a PI controller represented as a transfer function. In the textbook, the simulations are done for LTI systems, but here we simulate the full nonlinear system.

#### Parameter design through pole placement

To illustrate the design of a PI controller, we choose the gains  $k_p$  and  $k_i$  so that the characteristic polynomial has the form

$$s^2 + 2\zeta\omega_0 s + \omega_0^2$$

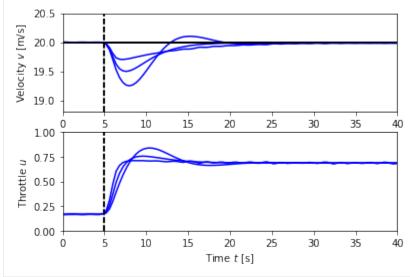
```
[9]: # Values of the first order transfer function P(s) = b/(s + a) are set above
    # Define the input that we want to track
    T = np.linspace(0, 40, 101)
    vref = 20 * np.ones(T.shape)
    gear = 4 * np.ones(T.shape)
    theta_hill = np.array([
        0 if t <= 5 else
        4./180. * pi * (t-5) if t <= 6 else
        4./180. * pi for t in T])
    # Fix \omega_0 and vary \zeta
    w0 = 0.5
    subplots = [None, None]
    for zeta in [0.5, 1, 2]:
         # Create the controller transfer function (as an I/O system)
        kp = (2*zeta*w0 - a)/b
        ki = w0 * *2 / b
        control_tf = ct.tf2io(
            ct.TransferFunction([kp, ki], [1, 0.01*ki/kp]),
```

```
name='control', inputs='u', outputs='y')

# Construct the closed loop system by interconnecting process and controller
cruise_tf = ct.InterconnectedSystem(
  (vehicle, control_tf), name='cruise',
  connections = [('control.u', '-vehicle.v'), ('vehicle.u', 'control.y')],
  inplist = ('control.u', 'vehicle.gear', 'vehicle.theta'),
    inputs = ('vref', 'gear', 'theta'),
  outlist = ('vehicle.v', 'vehicle.u'), outputs = ('v', 'u'))

# Plot the velocity response
X0, U0 = ct.find_eqpt(
    cruise_tf, [vref[0], 0], [vref[0], gear[0], theta_hill[0]],
    iu=[1, 2], y0=[vref[0], 0], iy=[0])

t, y = ct.input_output_response(cruise_tf, T, [vref, gear, theta_hill], X0)
  subplots = cruise_plot(cruise_tf, t, y, t_hill=5, subplots=subplots)
```



```
[10]: # Fix \zeta and vary \omega_0
     zeta = 1
     subplots = [None, None]
     for w0 in [0.2, 0.5, 1]:
          # Create the controller transfer function (as an I/O system)
         kp = (2*zeta*w0 - a)/b
         ki = w0**2 / b
         control_tf = ct.tf2io(
             ct.TransferFunction([kp, ki], [1, 0.01*ki/kp]),
             name='control', inputs='u', outputs='y')
         # Construct the closed loop system by interconnecting process and controller
         cruise_tf = ct.InterconnectedSystem(
         (vehicle, control_tf), name='cruise',
         connections = [('control.u', '-vehicle.v'), ('vehicle.u', 'control.y')],
         inplist = ('control.u', 'vehicle.gear', 'vehicle.theta'),
             inputs = ('vref', 'gear', 'theta'),
         outlist = ('vehicle.v', 'vehicle.u'), outputs = ('v', 'u'))
```

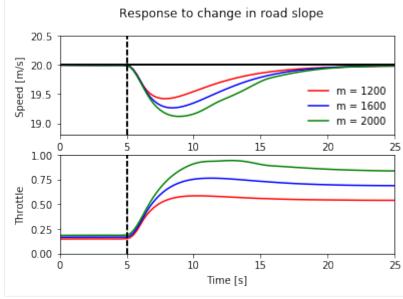
```
# Plot the velocity response
   X0, U0 = ct.find_eqpt(
        cruise_tf, [vref[0], 0], [vref[0], gear[0], theta_hill[0]],
        iu=[1, 2], y0=[vref[0], 0], iy=[0])
   t, y = ct.input_output_response(cruise_tf, T, [vref, gear, theta_hill], X0)
   subplots = cruise_plot(cruise_tf, t, y, t_hill=5, subplots=subplots)
   20.5
Velocity v [m/s]
  20.0
  19.5
  19.0
              5
                    10
                           15
                                  20
                                         25
                                               30
                                                      35
                                                             40
   1.00
  0.75
Throttle u
  0.50
  0.25
   0.00
                                         25
                    10
                           15
                                  20
                                               30
                                                      35
                                                             40
                               Time t [s]
```

#### Robustness to change in mass

```
[12]: # Define the time and input vectors
   T = np.linspace(0, 25, 101)
   vref = 20 * np.ones(T.shape)
   gear = 4 * np.ones(T.shape)
   theta0 = np.zeros(T.shape)

# Now simulate the effect of a hill at t = 5 seconds
   plt.figure()
   plt.suptitle('Response to change in road slope')
   theta_hill = np.array([
```

```
0 if t <= 5 else
    4./180. * pi * (t-5) if t <= 6 else
    4./180. * pi for t in T])
subplots = [None, None]
linecolor = ['red', 'blue', 'green']
handles = []
for i, m in enumerate([1200, 1600, 2000]):
    # Compute the equilibrium state for the system
   X0, U0 = ct.find_eqpt(
        cruise_tf, [vref[0], 0], [vref[0], gear[0], theta0[0]],
        iu=[1, 2], y0=[vref[0], 0], iy=[0], params={'m':m})
    t, y = ct.input_output_response(
        cruise_tf, T, [vref, gear, theta_hill], X0, params={'m':m})
    subplots = cruise_plot(cruise_tf, t, y, t_hill=5, subplots=subplots,
                           linetype=linecolor[i][0] + '-')
    handles.append(mlines.Line2D([], [], color=linecolor[i], linestyle='-',
                                 label="m = %d" % m))
# Add labels to the plots
plt.sca(subplots[0])
plt.ylabel('Speed [m/s]')
plt.legend(handles=handles, frameon=False, loc='lower right');
plt.sca(subplots[1])
plt.ylabel('Throttle')
plt.xlabel('Time [s]');
```



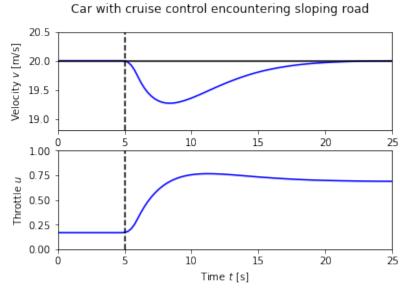
## PI controller with antiwindup protection

We now create a more complicated feedback controller that includes anti-windup protection.

```
[13]: def pi_update(t, x, u, params={}):
         # Get the controller parameters that we need
         ki = params.get('ki', 0.1)
         kaw = params.get('kaw', 2) # anti-windup gain
         # Assign variables for inputs and states (for readability)
                           # current velocity
         v = u[0]
         vref = u[1]
                                     # reference velocity
         z = x[0]
                                     # integrated error
         # Compute the nominal controller output (needed for anti-windup)
         u_a = pi_output(t, x, u, params)
         # Compute anti-windup compensation (scale by ki to account for structure)
         u_aw = kaw/ki * (np.clip(u_a, 0, 1) - u_a) if ki != 0 else 0
         # State is the integrated error, minus anti-windup compensation
         return (vref - v) + u_aw
     def pi_output(t, x, u, params={}):
         # Get the controller parameters that we need
         kp = params.get('kp', 0.5)
         ki = params.get('ki', 0.1)
         # Assign variables for inputs and states (for readability)
                               # current velocity
         v = u[0]
         vref = u[1]
                                    # reference velocity
         z = x[0]
                                    # integrated error
         # PI controller
         return kp * (vref - v) + ki * z
     control_pi = ct.NonlinearIOSystem(
         pi_update, pi_output, name='control',
         inputs = ['v', 'vref'], outputs = ['u'], states = ['z'],
         params = \{'kp':0.5, 'ki':0.1\})
      # Create the closed loop system
     cruise_pi = ct.InterconnectedSystem(
         (vehicle, control_pi), name='cruise',
         connections=(
             ('vehicle.u', 'control.u'),
             ('control.v', 'vehicle.v')),
         inplist=('control.vref', 'vehicle.gear', 'vehicle.theta'),
         outlist=('control.u', 'vehicle.v'), outputs=['u', 'v'])
```

## Response to a small hill

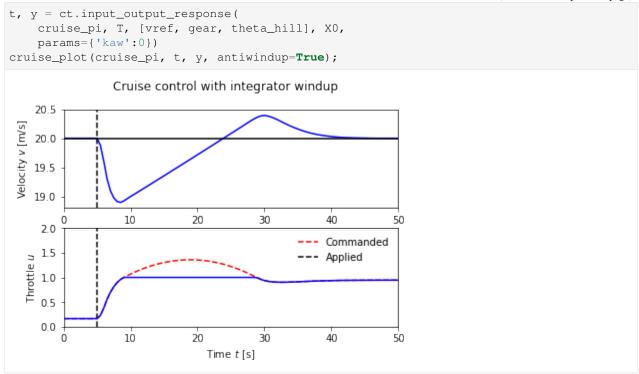
Figure 4.3b shows the response of the closed loop system. The figure shows that even if the hill is so steep that the throttle changes from 0.17 to almost full throttle, the largest speed error is less than 1 m/s, and the desired velocity is recovered after 20 s.



## **Effect of Windup**

The windup effect occurs when a car encounters a hill that is so steep  $(6^{\circ})$  that the throttle saturates when the cruise controller attempts to maintain speed.

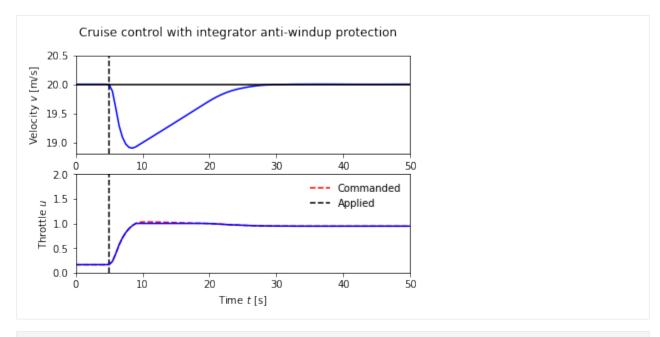
Chapter 8. Examples



## PI controller with anti-windup compensation

Anti-windup can be applied to the system to improve the response. Because of the feedback from the actuator model, the output of the integrator is quickly reset to a value such that the controller output is at the saturation limit.

```
[16]: plt.figure()
  plt.suptitle('Cruise control with integrator anti-windup protection')
  t, y = ct.input_output_response(
        cruise_pi, T, [vref, gear, theta_hill], X0,
        params={'kaw':2.})
  cruise_plot(cruise_pi, t, y, antiwindup=True);
```



[ ]:

# 8.2.2 Vehicle steering

Karl J. Astrom and Richard M. Murray 23 Jul 2019

This notebook contains the computations for the vehicle steering running example in Feedback Systems.

RMM comments to Karl, 27 Jun 2019 \* I'm using this notebook to walk through all of the vehicle steering examples and make sure that all of the parameters, conditions, and maximum steering angles are consitent and reasonable. \* Please feel free to send me comments on the contents as well as the bulletted notes, in whatever form is most convenient. \* Once we have sorted out all of the settings we want to use, I'll copy over the changes into the MATLAB files that we use for creating the figures in the book. \* These notes will be removed from the notebook once we have finalized everything.

```
[1]: import numpy as np
  import matplotlib.pyplot as plt
  import control as ct
  ct.use_fbs_defaults()
  ct.use_numpy_matrix(False)
```

## Vehicle steering dynamics (Example 3.11)

• System state: x, y, theta

The vehicle dynamics are given by a simple bicycle model. We take the state of the system as  $(x, y, \theta)$  where (x, y) is the position of the reference point of the vehicle in the plane and  $\theta$  is the angle of the vehicle with respect to horizontal. The vehicle input is given by  $(v, \delta)$  where v is the forward velocity of the vehicle and  $\delta$  is the angle of the steering wheel. We take as parameters the wheelbase b and the offset a between the rear wheels and the reference point. The model includes saturation of the vehicle steering angle (maxsteer).

```
System input: v, deltaSystem output: x, y
```

• System parameters: wheelbase, refoffset, maxsteer

Assuming no slipping of the wheels, the motion of the vehicle is given by a rotation around a point O that depends on the steering angle  $\delta$ . To compute the angle  $\alpha$  of the velocity of the reference point with respect to the axis of the vehicle, we let the distance from the center of rotation O to the contact point of the rear wheel be  $r_r$  and it the follows from Figure 3.17 in FBS that  $b = r_r \tan \delta$  and  $a = r_r \tan \alpha$ , which implies that  $\tan \alpha = (a/b) \tan \delta$ .

Reasonable limits for the steering angle depend on the speed. The physical limit is given in our model as 0.5 radians (about 30 degrees). However, this limit is rarely possible when the car is driving since it would cause the tires to slide on the pavement. We us a limit of 0.1 radians (about 6 degrees) at 10 m/s ( $\approx$  35 kph) and 0.05 radians (about 3 degrees) at 30 m/s ( $\approx$  110 kph). Note that a steering angle of 0.05 rad gives a cross acceleration of  $(v^2/b) \tan \delta \approx (100/3)0.05 = 1.7$  m/s<sup>2</sup> at 10 m/s and 15 m/s<sup>2</sup> at 30 m/s ( $\approx$  1.5 times the force of gravity).

```
[2]: def vehicle_update(t, x, u, params):
         # Get the parameters for the model
        a = params.get('refoffset', 1.5)  # offset to vehicle reference point
b = params.get('wheelbase', 3.)  # vehicle wheelbase
        b = params.get('wheelbase', 3.)
        maxsteer = params.get('maxsteer', 0.5) # max steering angle (rad)
         # Saturate the steering input
        delta = np.clip(u[1], -maxsteer, maxsteer)
        alpha = np.arctan2(a * np.tan(delta), b)
         # Return the derivative of the state
        return np.array([
             u[0] * np.cos(x[2] + alpha), # xdot = cos(theta + alpha) v
                                            # ydot = sin(theta + alpha) v
             u[0] * np.sin(x[2] + alpha),
                                             # thdot = v/l tan(phi)
             (u[0] / b) * np.tan(delta)
         ])
    def vehicle_output(t, x, u, params):
        return x[0:2]
    # Default vehicle parameters (including nominal velocity)
    vehicle_params={'refoffset': 1.5, 'wheelbase': 3, 'velocity': 15,
                     'maxsteer': 0.5}
    # Define the vehicle steering dynamics as an input/output system
    vehicle = ct.NonlinearIOSystem(
         vehicle_update, vehicle_output, states=3, name='vehicle',
         inputs=('v', 'delta'), outputs=('x', 'y'), params=vehicle_params)
```

## Vehicle driving on a curvy road (Figure 8.6a)

To illustrate the dynamics of the system, we create an input that correspond to driving down a curvy road. This trajectory will be used in future simulations as a reference trajectory for estimation and control.

RMM notes, 27 Jun 2019: \* The figure below appears in Chapter 8 (output feedback) as Example 8.3, but I've put it here in the notebook since it is a good way to demonstrate the dynamics of the vehicle. \* In the book, this figure is created for the linear model and in a manner that I can't quite understand, since the linear model that is used is only for the lateral dynamics. The original file is OutputFeedback/figures/steering\_obs.m. \* To create the figure here, I set the initial vehicle angle to be  $\theta(0) = 0.75$  rad and then used an input that gives a figure approximating Example 8.3 To create the lateral offset, I think subtracted the trajectory from the averaged straight line trajectory, shown as a dashed line in the xy figure below. \* I find the approach that we used in the MATLAB version to be confusing, but I also think the method of creating the lateral error here is a hart to follow. We might instead consider choosing a trajectory that goes mainly vertically, with the 2D dynamics being the x,  $\theta$  dynamics instead of the y,  $\theta$  dynamics.

KJA comments, 1 Jul 2019:

- 0. I think we should point out that the reference point is typically the projection of the center of mass of the whole vehicle.
- 1. The heading angle  $\theta$  must be marked in Figure 3.17b.
- 2. I think it is useful to start with a curvy road that you have done here but then to specialized to a trajectory that is essentially horizontal, where y is the deviation from the nominal horizontal x axis. Assuming that  $\alpha$  and  $\theta$  are small we get the natural linearization of (3.26)  $\dot{x} = v$  and  $\dot{y} = v(\alpha + \theta)$

RMM response, 16 Jul 2019: \* I've changed the trajectory to be about the horizontal axis, but I am ploting things vertically for better figure layout. This corresponds to what is done in Example 9.10 in the text, which I think looks OK.

KJA response, 20 Jul 2019: Fig 8.6a is fine

```
[3]: # System parameters
    wheelbase = vehicle_params['wheelbase']
    v0 = vehicle_params['velocity']
    # Control inputs
    T_{curvy} = np.linspace(0, 7, 500)
    v_curvy = v0*np.ones(T_curvy.shape)
    delta_curvy = 0.1*np.sin(T_curvy)*np.cos(4*T_curvy) + 0.0025*np.sin(T_curvy*np.pi/7)
    u_curvy = [v_curvy, delta_curvy]
    X0_{curvy} = [0, 0.8, 0]
    # Simulate the system + estimator
    t_curvy, y_curvy, x_curvy = ct.input_output_response(
        vehicle, T_curvy, u_curvy, X0_curvy, params=vehicle_params, return_x=True)
    # Configure matplotlib plots to be a bit bigger and optimize layout
    plt.figure(figsize=[9, 4.5])
    # Plot the resulting trajectory (and some road boundaries)
    plt.subplot(1, 4, 2)
    plt.plot(y_curvy[1], y_curvy[0])
    plt.plot(y_curvy[1] - 9/np.cos(x_curvy[2]), y_curvy[0], 'k-', linewidth=1)
    plt.plot(y_curvy[1] - 3/np.cos(x_curvy[2]), y_curvy[0], 'k--', linewidth=1)
    plt.plot(y_curvy[1] + 3/np.cos(x_curvy[2]), y_curvy[0], 'k-', linewidth=1)
    plt.xlabel('y [m]')
```

```
plt.ylabel('x [m]');
plt.axis('Equal')
# Plot the lateral position
plt.subplot(2, 2, 2)
plt.plot(t_curvy, y_curvy[1])
plt.ylabel('Lateral position $y$ [m]')
# Plot the steering angle
plt.subplot(2, 2, 4)
plt.plot(t_curvy, delta_curvy)
plt.ylabel('Steering angle $\\delta$ [rad]')
plt.xlabel('Time t [sec]')
plt.tight_layout()
                                    Lateral position y [m]
   100
                                         0.5
                                         0.0
    80
                                       -0.5
    60
x [m]
                                                         ż
                                              0
                                                                              6
                                        0.10
                                   Steering angle 6 [rad]
    40
                                        0.05
    20
                                        0.00
                                      -0.05
      0
                                      -0.10
                  0
                                                         2
        -20
                                              0
                                                                              6
              y [m]
                                                            Time t [sec]
```

## Linearization of lateral steering dynamics (Example 6.13)

We are interested in the motion of the vehicle about a straight-line path  $(\theta = \theta_0)$  with constant velocity  $v_0 \neq 0$ . To find the relevant equilibrium point, we first set  $\dot{\theta} = 0$  and we see that we must have  $\delta = 0$ , corresponding to the steering wheel being straight. The motion in the xy plane is by definition not at equilibrium and so we focus on lateral deviation of the vehicle from a straight line. For simplicity, we let  $\theta_e = 0$ , which corresponds to driving along the x axis. We can then focus on the equations of motion in the y and  $\theta$  directions with input  $u = \delta$ .

```
[4]: # Define the lateral dynamics as a subset of the full vehicle steering dynamics
lateral = ct.NonlinearIOSystem(
    lambda t, x, u, params: vehicle_update(
        t, [0., x[0], x[1]], [params.get('velocity', 1), u[0]], params)[1:],
    lambda t, x, u, params: vehicle_output(
        t, [0., x[0], x[1]], [params.get('velocity', 1), u[0]], params)[1:],
    states=2, name='lateral', inputs=('phi'), outputs=('y')
)

# Compute the linearization at velocity v0 = 15 m/sec
```

```
lateral_linearized = ct.linearize(lateral, [0, 0], [0], params=vehicle_params)
\# Normalize dynamics using state [x1/b, x2] and timescale v0 t / b
b = vehicle_params['wheelbase']
v0 = vehicle_params['velocity']
lateral_transformed = ct.similarity_transform(
    lateral_linearized, [[1/b, 0], [0, 1]], timescale=v0/b)
# Set the output to be the normalized state x1/b
lateral_normalized = lateral_transformed * (1/b)
print("Linearized system dynamics:\n")
print(lateral_normalized)
# Save the system matrices for later use
A = lateral normalized.A
B = lateral_normalized.B
C = lateral_normalized.C
Linearized system dynamics:
A = [[0. 1.]]
[0. 0.]]
B = [[0.5]]
[1.]
C = [[1. 0.]]
D = [[0.]]
```

#### Eigenvalue placement controller design (Example 7.4)

We want to design a controller that stabilizes the dynamics of the vehicle and tracks a given reference value r of the lateral position of the vehicle. We use feedback to design the dynamics of the system to have the characteristic polynomial  $p(s) = s^2 + 2\zeta_c\omega_c + \omega_c^2$ .

To find reasonable values of  $\omega_c$  we observe that the initial response of the steering angle to a unit step change in the steering command is  $\omega_c^2 r$ , where r is the commanded lateral transition. Recall that the model is normalized so that the length unit is the wheelbase b and the time unit is the time  $b/v_0$  to travel one wheelbase. A typical car has a wheelbase of about 3 m and, assuming a speed of 30 m/s, a normalized time unit corresponds to 0.1 s. To determine a reasonable steering angle when making a gentle lane change, we assume that the turning radius is R = 600 m. For a wheelbase of 3 m this corresponds to a steering angle  $\delta \approx 3/600 = 0.005$  rad and a lateral acceleration of  $v^2/R = 302/600 = 1.5$  m/s<sup>2</sup>. Assuming that a lane change corresponds to a translation of one wheelbase we find  $\omega_c = \sqrt{0.005} = 0.07$  rad/s.

The unit step responses for the closed loop system for different values of the design parameters are shown below. The effect of  $\omega_c$  is shown on the left, which shows that the response speed increases with increasing  $\omega_c$ . All responses have overshoot less than 5% (15 cm), as indicated by the dashed lines. The settling times range from 30 to 60 normalized time units, which corresponds to about 3–6 s, and are limited by the acceptable lateral acceleration of the vehicle. The effect of  $\zeta_c$  is shown on the right. The response speed and the overshoot increase with decreasing damping. Using these plots, we conclude that a reasonable design choice is  $\omega_c = 0.07$  and  $\zeta_c = 0.7$ .

RMM note, 27 Jun 2019: \* The design guidelines are for  $v_0 = 30$  m/s (highway speeds) but most of the examples below are done at lower speed (typically 10 m/s). Also, the eigenvalue locations above are not the same ones that we use in the output feedback example below. We should probably make things more consistent.

KJA comment, 1 Jul 2019: \* I am all for maiking it consist and choosing e.g. v0 = 30 m/s

RMM comment, 17 Jul 2019: \* I've updated the examples below to use v0 = 30 m/s for everything except the forward/reverse example. This corresponds to ~105 kph (freeway speeds) and a reasonable bound for the steering angle to avoid slipping is 0.05 rad.

```
[5]: # Utility function to place poles for the normalized vehicle steering system
    def normalized_place(wc, zc):
        # Get the dynamics and input matrices, for later use
        A, B = lateral_normalized.A, lateral_normalized.B
        # Compute the eigenvalues from the characteristic polynomial
        eigs = np.roots([1, 2*zc*wc, wc**2])
        # Compute the feedback gain using eigenvalue placement
        K = ct.place_varga(A, B, eigs)
        # Create a new system representing the closed loop response
        clsys = ct.StateSpace(A - B @ K, B, lateral_normalized.C, 0)
        # Compute the feedforward gain based on the zero frequency gain of the closed loop
        kf = np.real(1/clsys.evalfr(0))
        # Scale the input by the feedforward gain
        clsys *= kf
        # Return gains and closed loop system dynamics
        return K, kf, clsys
    # Utility function to plot simulation results for normalized vehicle steering system
    def normalized_plot(t, y, u, inpfig, outfig):
        plt.sca(outfig)
        plt.plot(t, y)
        plt.sca(inpfig)
        plt.plot(t, u[0])
    # Utility function to label plots of normalized vehicle steering system
    def normalized_label(inpfig, outfig):
        plt.sca(inpfig)
        plt.xlabel('Normalized time $v_0 t / b$')
        plt.ylabel('Steering angle $\delta$ [rad]')
        plt.sca(outfig)
        plt.ylabel('Lateral position $y/b$')
        plt.plot([0, 20], [0.95, 0.95], 'k--')
        plt.plot([0, 20], [1.05, 1.05], 'k--')
    # Configure matplotlib plots to be a bit bigger and optimize layout
    plt.figure(figsize=[9, 4.5])
    # Explore range of values for omega_c, with zeta_c = 0.7
    outfig = plt.subplot(2, 2, 1)
    inpfig = plt.subplot(2, 2, 3)
    zc = 0.7
    for wc in [0.5, 0.7, 1]:
        # Place the poles of the system
        K, kf, clsys = normalized_place(wc, zc)
```

(continued from previous page) # Compute the step response t, y, x = ct.step\_response(clsys, np.linspace(0, 20, 100), return\_x=True) # Compute the input used to generate the control response u = -K @ x + kf \* 1# Plot the results normalized\_plot(t, y, u, inpfig, outfig) # Add labels to the figure normalized\_label(inpfig, outfig)  $plt.legend(('$\oega_c = 0.5$', '$\oega_c = 0.7$', '$\oega_c = 0.1$'))$ # Explore range of values for zeta\_c, with omega\_c = 0.07outfig = plt.subplot(2, 2, 2)inpfig = plt.subplot(2, 2, 4) wc = 0.7for zc in [0.5, 0.7, 1]: # Place the poles of the system K, kf, clsys = normalized\_place(wc, zc) # Compute the step response t, y,  $x = ct.step_response(clsys, np.linspace(0, 20, 100), return_x=True)$ # Compute the input used to generate the control response u = -K @ x + kf \* 1# Plot the results normalized\_plot(t, y, u, inpfig, outfig) # Add labels to the figure normalized\_label(inpfig, outfig) plt.legend(('\$\zeta\_c = 0.5\$', '\$\zeta\_c = 0.7\$', '\$\zeta\_c = 1\$')) plt.tight\_layout() 1.00 ateral position y/b Lateral position y/b 1.0 0.75 0.50  $\omega_c = 0.5$  $\zeta_c = 0.5$ 0.5  $\omega_c = 0.7$  $\zeta_c = 0.7$ 0.25  $\omega_c = 0.1$  $\zeta_c = 1$ 0.00 0.0 5 5 10 10 15 20 15 20 1.0 Steering angle 6 [rad] Steering angle 6 [rad] 0.4 0.5 0.2 0.0

RMM notes, 17 Jul 2019 \* These step responses are very slow. Note that the steering wheel angles are about 10X

0

20

0.0

0

10

Normalized time  $v_0t/b$ 

15

15

20

10

Normalized time  $v_0t/b$ 

less than a resonable bound (0.05 rad at 30 m/s). A consequence of these low gains is that the tracking controller in Example 8.4 has to use a different set of gains. We could update, but the gains listed here have a rationale that we would have to update as well. \* Based on the discussion below, I think we should make  $\omega_c$  range from 0.5 to 1 (10X faster).

KJA response, 20 Jul 2019: Makes a lot of sense to make  $\omega_c$  range from 0.5 to 1 (10X faster). The plots were still in the range 0.05 to 0.1 in the note you sent me.

RMM response: 23 Jul 2019: Updated  $\omega_c$  to 10X faster. Note that this makes size of the inputs for the step response quite large, but that is in part because a unit step in the desired position produces an (instantaneous) error of b=3 m  $\Longrightarrow$  quite a large error. A lateral error of 10 cm with  $\omega_c=0.7$  would produce an (initial) input of 0.015 rad.

## Eigenvalue placement observer design (Example 8.3)

We construct an estimator for the (normalized) lateral dynamics by assigning the eigenvalues of the estimator dynamics to desired value, specifified in terms of the second order characteristic equation for the estimator dynamics.

#### Linear observer applied to nonlinear system output

A simulation of the observer for a vehicle driving on a curvy road is shown below. The first figure shows the trajectory of the vehicle on the road, as viewed from above. The response of the observer is shown on the right, where time is normalized to the vehicle length. We see that the observer error settles in about 4 vehicle lengths.

RMM note, 27 Jun 2019: \* As an alternative, we can attempt to estimate the state of the full nonlinear system using a linear estimator. This system does not necessarily converge to zero since there will be errors in the nominal dynamics of the system for the linear estimator. \* The limits on the x axis for the time plots are different to show the error over the entire trajectory. \* We should decide whether we want to keep the figure above or the one below for the text.

KJA comment, 1 Jul 2019: \* I very much like your observation about the nonlinear system. I think it is a very good idea to use your new simulation

RMM comment, 17 Jul 2019: plan to use this version in the text.

KJA comment, 20 Jul 2019: I think this is a big improvement we show that an observer based on a linearized model works on a nonlinear simulation, If possible we could add a line telling why the linear model works and that this is standard procedure in control engineering.

```
[7]: # Convert the curvy trajectory into normalized coordinates
x_ref = x_curvy[0] / wheelbase
y_ref = x_curvy[1] / wheelbase
(continues on next page)
```

```
theta_ref = x_curvy[2]
tau = v0 * T_curvy / b
# Simulate the estimator, with a small initial error in y position
t, y_est, x_est = ct.forced_response(est, tau, [delta_curvy, y_ref], [0.5, 0])
# Configure matplotlib plots to be a bit bigger and optimize layout
plt.figure(figsize=[9, 4.5])
# Plot the actual and estimated states
ax = plt.subplot(2, 2, 1)
plt.plot(t, y_ref)
plt.plot(t, x_est[0])
ax.set(xlim=[0, 10])
plt.legend(['actual', 'estimated'])
plt.ylabel('Lateral position $y/b$')
ax = plt.subplot(2, 2, 2)
plt.plot(t, x_est[0] - y_ref)
ax.set(xlim=[0, 10])
plt.ylabel('Lateral error')
ax = plt.subplot(2, 2, 3)
plt.plot(t, theta_ref)
plt.plot(t, x_est[1])
ax.set(xlim=[0, 10])
plt.xlabel('Normalized time $v_0 t / b$')
plt.ylabel('Vehicle angle $\\theta$')
ax = plt.subplot(2, 2, 4)
plt.plot(t, x_est[1] - theta_ref)
ax.set(xlim=[0, 10])
plt.xlabel('Normalized time $v_0 t / b$')
plt.ylabel('Angle error')
plt.tight_layout()
                                      actual
 Lateral position y/b
                                                       0.2
     0.4
                                      estimated
                                                    Lateral error
     0.2
                                                       0.1
     0.0
                                                       0.0
    -0.2
                2
                       4
                                       8
                                              10
                                                                                        8
                                                                                                10
                                                     0.000
     0.1
 Vehicle angle \theta
                                                    -0.025
                                                 Angle error
     0.0
                                                    -0.050
                                                   -0.075
    -0.1
                                                    -0.100
                2
        0
                                       8
                                              10
                                                          0
                                                                                                10
                   Normalized time v_0t/b
                                                                     Normalized time vot/b
```

## **Output Feedback Controller (Example 8.4)**

RMM note, 27 Jun 2019 \* The feedback gains for the controller below are different that those computed in the eigenvalue placement example (from Ch 7), where an argument was given for the choice of the closed loop eigenvalues. Should we choose a single, consistent set of gains in both places? \* This plot does not quite match Example 8.4 because a different reference is being used for the laterial position. \* The transient in  $\delta$  is quiet large. This appears to be due to the error in  $\theta(0)$ , which is initialized to zero intead of to theta curvy.

KJA comment, 1 Jul 2019: 1. The large initial errors dominate the plots.

2. There is somehing funny happening at the end of the simulation, may be due to the small curvature at the end of the path?

RMM comment, 17 Jul 2019: \* Updated to use the new trajectory \* We will have the issue that the gains here are different than the gains that we used in Chapter 7. I think that what we need to do is update the gains in Ch 7 (they are too sluggish, as noted above). \* Note that unlike the original example in the book, the errors do not converge to zero. This is because we are using pure state feedback (no feedforward) => the controller doesn't apply any input until there is an error.

KJA comment, 20 Jul 2019: We may add that state feedback is a proportional controller which does not guarantee that the error goes to zero for example by changing the line "The tracking error ..." to "The tracking error can be improved by adding integral action (Section 7.4), later in this chapter "Disturbance Modeling" or feedforward (Section 8,5). Should we do an exercises?

```
[8]: # Compute the feedback gains
    # K, kf, clsys = normalized_place(1, 0.707)
                                                    # Gains from MATLAB
    # K, kf, clsys = normalized_place(0.07, 0.707) # Original gains
    K, kf, clsys = normalized_place(0.7, 0.707)
                                                     # Final gains
    # Print out the gains
    print("K = ", K)
    print("kf = ", kf)
    # Construct an output-based controller for the system
    clsys = ct.StateSpace(
        np.block([[A, -B@K], [L@C, A - B@K - L@C]]),
        np.block([[B], [B]]) * kf,
        np.block([[C, np.zeros(C.shape)], [np.zeros(C.shape), C]]),
        np.zeros((2,1)))
    # Simulate the system
    t, y, x = \text{ct.forced\_response}(\text{clsys}, \text{tau}, \text{y\_ref}, [0.4, 0, 0.0, 0])
    # Calcaluate the input used to generate the control response
    u\_sfb = kf * y\_ref - K @ x[0:2]
    u\_ofb = kf * y\_ref - K @ x[2:4]
    # Configure matplotlib plots to be a bit bigger and optimize layout
    plt.figure(figsize=[9, 4.5])
    # Plot the actual and estimated states
    ax = plt.subplot(1, 2, 1)
    plt.plot(t, x[0])
    plt.plot(t, x[2])
    plt.plot(t, y_ref, 'k-.')
    ax.set(xlim=[0, 30])
    plt.legend(['state feedback', 'output feedback', 'reference'])
    plt.xlabel('Normalized time $v_0 t / b$')
```

```
plt.ylabel('Lateral position $y/b$')
ax = plt.subplot(2, 2, 2)
plt.plot(t, x[1])
plt.plot(t, x[3])
plt.plot(t, theta_ref, 'k-.')
ax.set(xlim=[0, 15])
plt.ylabel('Vehicle angle $\\theta$')
ax = plt.subplot(2, 2, 4)
plt.plot(t, u_sfb[0])
plt.plot(t, u_ofb[0])
plt.plot(t, delta_curvy, 'k-.')
ax.set(xlim=[0, 15])
plt.xlabel('Normalized time $v_0 t / b$')
plt.ylabel('Steering angle $\\delta$')
plt.tight_layout()
K = [0.49]
               0.744811
kf = [[0.49]]
                                     state feedback
     0.4
                                                            0.1
                                     output feedback
                                                       Vehicle angle \theta
                                     reference
     0.3
                                                            0.0
      0.2
 ateral position y/b
                                                           -0.1
     0.1
                                                               0
                                                                     2
                                                                           4
                                                                                      8
                                                                                            10
                                                                                                 12
                                                                                                       14
     0.0
                                                       Steering angle \delta
                                                            0.1
    -0.1
                                                            0.0
    -0.2
                                                           -0.1
    -0.3
                5
                       10
                              15
                                     20
                                            25
                                                                     2
                                                                                      8
                                                                                            10
                                                                                                 12
                                                                                                       14
                                                   30
                                                               0
                                                                                 6
                      Normalized time v_0t/b
                                                                            Normalized time v_0t/b
```

#### **Trajectory Generation (Example 8.8)**

To illustrate how we can use a two degree-of-freedom design to improve the performance of the system, consider the problem of steering a car to change lanes on a road. We use the non-normalized form of the dynamics, which were derived in Example 3.11.

KJA comment, 1 Jul 2019: 1. I think the reference trajectory is too much curved in the end compare with Example 3.11

In summary I think it is OK to change the reference trajectories but we should make sure that the curvature is less than  $\rho = 600m$  not to have too high acceleratarion.

RMM response, 16 Jul 2019: \* Not sure if the comment about the trajectory being too curved is referring to this example. The steering angles (and hence radius of curvature/acceleration) are quite low. ??

KJA response, 20 Jul 2019: You are right the curvature is not too small. We could add the sentence "The small deviations can be eliminated by adding feedback."

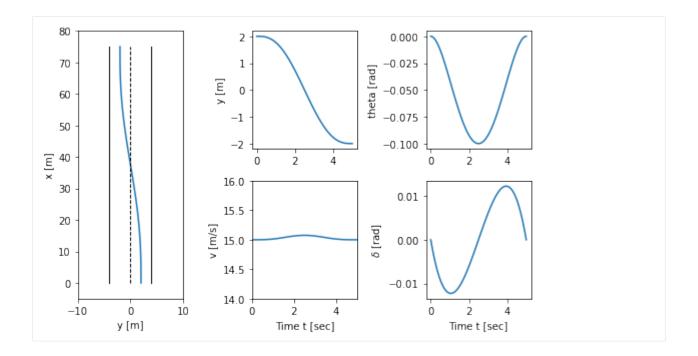
RMM response, 23 Jul 2019: I think the small deviation you are referring to is in the velocity trace. This occurs because I gave a fixed endpoint in time and so the velocity had to be adjusted to hit that exact point at that time. This doesn't show up in the book, so it won't be a problem ( $\implies$  no additional explanation required).

```
[9]: import control.flatsys as fs
    # Function to take states, inputs and return the flat flag
    def vehicle_flat_forward(x, u, params={}):
        # Get the parameter values
        b = params.get('wheelbase', 3.)
        # Create a list of arrays to store the flat output and its derivatives
        zflag = [np.zeros(3), np.zeros(3)]
        # Flat output is the x, y position of the rear wheels
        zflag[0][0] = x[0]
        zflag[1][0] = x[1]
        # First derivatives of the flat output
        zflag[0][1] = u[0] * np.cos(x[2]) # dx/dt
        zflag[1][1] = u[0] * np.sin(x[2]) # dy/dt
        # First derivative of the angle
        thdot = (u[0]/b) * np.tan(u[1])
        # Second derivatives of the flat output (setting vdot = 0)
        zflag[0][2] = -u[0] * thdot * np.sin(x[2])
        zflag[1][2] = u[0] * thdot * np.cos(x[2])
        return zflag
    # Function to take the flat flag and return states, inputs
    def vehicle_flat_reverse(zflag, params={}):
        # Get the parameter values
        b = params.get('wheelbase', 3.)
        # Create a vector to store the state and inputs
        x = np.zeros(3)
        u = np.zeros(2)
        # Given the flat variables, solve for the state
        x[0] = zflag[0][0] # x position
        x[1] = zflag[1][0] # y position
        x[2] = np.arctan2(zflag[1][1], zflag[0][1]) # tan(theta) = ydot/xdot
        # And next solve for the inputs
        u[0] = zflag[0][1] * np.cos(x[2]) + zflag[1][1] * np.sin(x[2])
        thdot_v = zflag[1][2] * np.cos(x[2]) - zflag[0][2] * np.sin(x[2])
        u[1] = np.arctan2(thdot_v, u[0]**2 / b)
        return x, u
    vehicle_flat = fs.FlatSystem(vehicle_flat_forward, vehicle_flat_reverse, inputs=2,__
     →states=3)
```

To find a trajectory from an initial state  $x_0$  to a final state  $x_f$  in time  $T_f$  we solve a point-to-point trajectory generation

problem. We also set the initial and final inputs, which sets the vehicle velocity v and steering wheel angle  $\delta$  at the endpoints.

```
[10]: # Define the endpoints of the trajectory
     x0 = [0., 2., 0.]; u0 = [15, 0.]
     xf = [75, -2., 0.]; uf = [15, 0.]
     Tf = xf[0] / uf[0]
     # Define a set of basis functions to use for the trajectories
     poly = fs.PolyFamily(6)
     # Find a trajectory between the initial condition and the final condition
     traj = fs.point_to_point(vehicle_flat, x0, u0, xf, uf, Tf, basis=poly)
     # Create the trajectory
     t = np.linspace(0, Tf, 100)
     x, u = traj.eval(t)
     # Configure matplotlib plots to be a bit bigger and optimize layout
     plt.figure(figsize=[9, 4.5])
     # Plot the trajectory in xy coordinate
     plt.subplot(1, 4, 2)
     plt.plot(x[1], x[0])
     plt.xlabel('y [m]')
     plt.ylabel('x [m]')
     # Add lane lines and scale the axis
     plt.plot([-4, -4], [0, x[0, -1]], 'k-', linewidth=1)
     plt.plot([0, 0], [0, x[0, -1]], 'k--', linewidth=1)
     plt.plot([4, 4], [0, x[0, -1]], 'k-', linewidth=1)
     plt.axis([-10, 10, -5, x[0, -1] + 5])
     # Time traces of the state and input
     plt.subplot(2, 4, 3)
     plt.plot(t, x[1])
     plt.ylabel('y [m]')
     plt.subplot(2, 4, 4)
     plt.plot(t, x[2])
     plt.ylabel('theta [rad]')
     plt.subplot(2, 4, 7)
     plt.plot(t, u[0])
     plt.xlabel('Time t [sec]')
     plt.ylabel('v [m/s]')
     plt.axis([0, Tf, u0[0] - 1, uf[0] +1])
     plt.subplot(2, 4, 8)
     plt.plot(t, u[1]);
     plt.xlabel('Time t [sec]')
     plt.ylabel('$\delta$ [rad]')
     plt.tight_layout()
```



## Vehicle transfer functions for forward and reverse driving (Example 10.11)

The vehicle steering model has different properties depending on whether we are driving forward or in reverse. The figures below show step responses from steering angle to lateral translation for a the linearized model when driving forward (dashed) and reverse (solid). In this simulation we have added an extra pole with the time constant T=0.1 to approximately account for the dynamics in the steering system.

With rear-wheel steering the center of mass first moves in the wrong direction and the overall response with rear-wheel steering is significantly delayed compared with that for front-wheel steering. (b) Frequency response for driving forward (dashed) and reverse (solid). Notice that the gain curves are identical, but the phase curve for driving in reverse has non-minimum phase.

RMM note, 27 Jun 2019: \* I cannot recreate the figures in Example 10.11. Since we are looking at the lateral *velocity*, there is a differentiator in the output and this takes the step function and creates an offset at t=0 (intead of a smooth curve). \* The transfer functions are also different, and I don't quite understand why. Need to spend a bit more time on this one.

KJA comment, 1 Jul 2019: The reason why you cannot recreate figures i Example 10.11 is because the caption in figure is wrong, sorry my fault, the y-axis should be lateral position not lateral velocity. The approximate expression for the transfer functions

$$G_{y\delta} = \frac{av_0s + v_0^2}{bs} = \frac{1.5s + 1}{3s^2} = \frac{0.5s + 0.33}{s}$$

are quite close to the values that you get numerically

In this case I think it is useful to have v=1 m/s because we do not drive to fast backwards.

RMM response, 17 Jul 2019 \* Updated figures below use the same parameters as the running example (the current text uses different parameters) \* Following the material in the text, a pole is added at s = -1 to approximate the dynamics of the steering system. This is not strictly needed, so we could decide to take it out (and update the text)

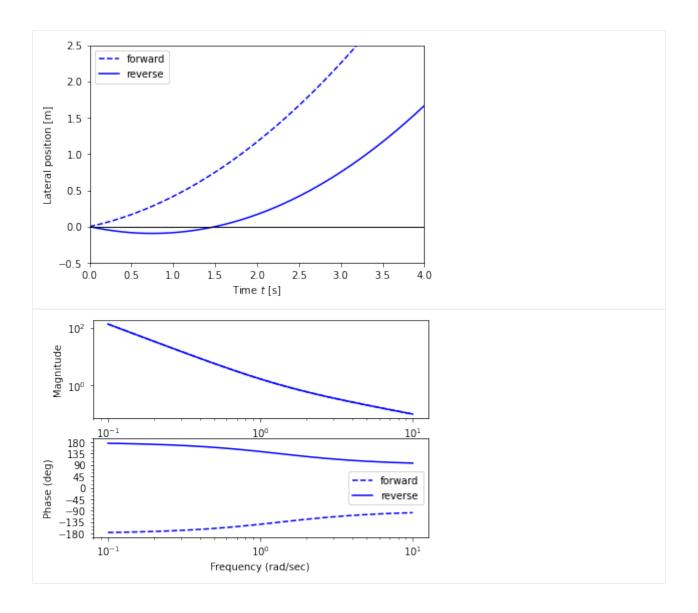
KJA comment, 20 Jul 2019: I have been oscillating a bit about this example. Of course it does not make sense to drive in reverse in 30 m/s but it seems a bit silly to change parameters just in this case (if we do we have to motivate it). On the other hand what we are doing is essentially based on transfer functions and a RHP zero. My current view which

has changed a few times is to keep the standard parameters. In any case we should eliminate the extra time constant. A small detail, I could not see the time response in the file you sent, do not resend it!, I will look at the final version.

RMM comment, 23 Jul 2019: I think it is OK to have the speed be different and just talk about this in the text. I have removed the extra time constant in the current version.

```
[11]: # Magnitude of the steering input (half maximum)
     Msteer = vehicle_params['maxsteer'] / 2
     # Create a linearized model of the system going forward at 2 m/s
     forward_lateral = ct.linearize(lateral, [0, 0], [0], params={'velocity': 2})
     forward_tf = ct.ss2tf(forward_lateral)[0, 0]
     print("Forward TF = ", forward_tf)
     # Create a linearized model of the system going in reverise at 1 m/s
     reverse_lateral = ct.linearize(lateral, [0, 0], [0], params={'velocity': -2})
     reverse_tf = ct.ss2tf(reverse_lateral)[0, 0]
     print("Reverse TF = ", reverse_tf)
     Forward TF =
              s + 1.333
     s^2 + 7.828e - 16 s - 1.848e - 16
     Reverse TF =
             -s + 1.333
     ______
     s^2 - 7.828e - 16 s - 1.848e - 16
```

```
[12]: # Configure matplotlib plots to be a bit bigger and optimize layout
     plt.figure()
     # Forward motion
     t, y = ct.step_response(forward_tf * Msteer, np.linspace(0, 4, 500))
     plt.plot(t, y, 'b--')
     # Reverse motion
     t, y = ct.step_response(reverse_tf * Msteer, np.linspace(0, 4, 500))
     plt.plot(t, y, 'b-')
     # Add labels and reference lines
     plt.axis([0, 4, -0.5, 2.5])
     plt.legend(['forward', 'reverse'], loc='upper left')
     plt.xlabel('Time $t$ [s]')
     plt.ylabel('Lateral position [m]')
     plt.plot([0, 4], [0, 0], 'k-', linewidth=1)
     # Plot the Bode plots
     plt.figure()
     plt.subplot(1, 2, 2)
     ct.bode_plot(forward_tf[0, 0], np.logspace(-1, 1, 100), color='b', linestyle='--')
     ct.bode_plot(reverse_tf[0, 0], np.logspace(-1, 1, 100), color='b', linestyle='-')
     plt.legend(('forward', 'reverse'));
```



#### Feedforward Compensation (Example 12.6)

For a lane transfer system we would like to have a nice response without overshoot, and we therefore consider the use of feedforward compensation to provide a reference trajectory for the closed loop system. We choose the desired response as  $F_{\rm m}(s)=a^22/(s+a)^2$ , where the response speed or aggressiveness of the steering is governed by the parameter a.

RMM note, 27 Jun 2019: \* a was used in the original description of the dynamics as the reference offset. Perhaps choose a different symbol here? \* In current version of Ch 12, the y axis is labeled in absolute units, but it should actually be in normalized units, I think. \* The steering angle input for this example is quite high. Compare to Example 8.8, above. Also, we should probably make the size of the "lane change" from this example match whatever we use in Example 8.8

KJA comments, 1 Jul 2019: Chosen parameters look good to me

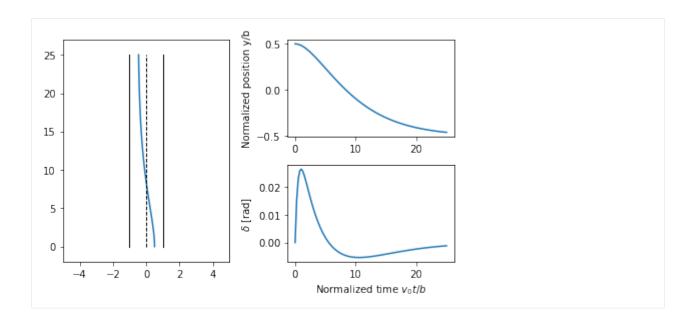
RMM response, 17 Jul 2019 \* I changed the time constant for the feedforward model to give something that is more reasonable in terms of turning angle at the speed of  $v_0 = 30$  m/s. Note that this takes about 30 body lengths to change lanes (= 9 seconds at 105 kph). \* The time to change lanes is about 2X what it is using the differentially flat trajectory

above. This is mainly because the feedback controller applies a large pulse at the beginning of the trajectory (based on the input error), whereas the differentially flat trajectory spreads the turn over a longer interval. Since are living the steering angle, we have to limit the size of the pulse => slow down the time constant for the reference model.

KJA response, 20 Jul 2019: I think the time for lane change is too long, which may depend on the small steering angles used. The largest steering angle is about 0.03 rad, but we have admitted larger values in previous examples. I suggest that we change the design so that the largest sterring angel is closer to 0.05, see the remark from Bjorn O a lane change could take about 5 s at 30m/s.

RMM response, 23 Jul 2019: I reset the time constant to 0.2, which gives something closer to what we had for trajectory generation. It is still slower, but this is to be expected since it is a linear controller. We now finish the trajectory in 20 body lengths, which is about 6 seconds.

```
[13]: # Define the desired response of the system
     a = 0.2
     P = ct.ss2tf(lateral normalized)
     Fm = ct.TransferFunction([a**2], [1, 2*a, a**2])
     Fr = Fm / P
      # Compute the step response of the feedforward components
     t, y_ffwd = ct.step_response(Fm, np.linspace(0, 25, 100))
     t, delta_ffwd = ct.step_response(Fr, np.linspace(0, 25, 100))
      # Scale and shift to correspond to lane change (-2 to +2)
     y_ffwd = 0.5 - 1 * y_ffwd
     delta_ffwd *= 1
      # Overhead view
     plt.subplot(1, 2, 1)
     plt.plot(y_ffwd, t)
     plt.plot(-1*np.ones(t.shape), t, 'k-', linewidth=1)
     plt.plot(0*np.ones(t.shape), t, 'k--', linewidth=1)
     plt.plot(1*np.ones(t.shape), t, 'k-', linewidth=1)
     plt.axis([-5, 5, -2, 27])
      # Plot the response
     plt.subplot(2, 2, 2)
     plt.plot(t, y_ffwd)
      # plt.axis([0, 10, -5, 5])
     plt.ylabel('Normalized position y/b')
     plt.subplot(2, 2, 4)
     plt.plot(t, delta_ffwd)
      # plt.axis([0, 10, -1, 1])
     plt.ylabel('$\\delta$ [rad]')
     plt.xlabel('Normalized time $v_0 t / b$');
     plt.tight_layout()
```



#### **Fundamental Limits (Example 14.13)**

Consider a controller based on state feedback combined with an observer where we want a faster closed loop system and choose  $\omega_c = 10$ ,  $\zeta_c = 0.707$ ,  $\omega_o = 20$ , and  $\zeta_o = 0.707$ .

KJA comment, 20 Jul 2019: This is a really troublesome case. If we keep it as a vehicle steering problem we must have an order of magnitude lower valuer for  $\omega_c$  and  $\omega_o$  and then the zero will not be slow. My recommendation is to keep it as a general system with the transfer function.  $P(s) = (s+1)/s^2$ . The text then has to be reworded.

RMM response, 23 Jul 2019: I think the way we have it is OK. Our current value for the controller and observer is  $\omega_c = 0.7$  and  $\omega_o = 1$ . Here we way we want something faster and so we got to  $\omega_c = 7$  (10X) and  $\omega_o = 10$  (10X).

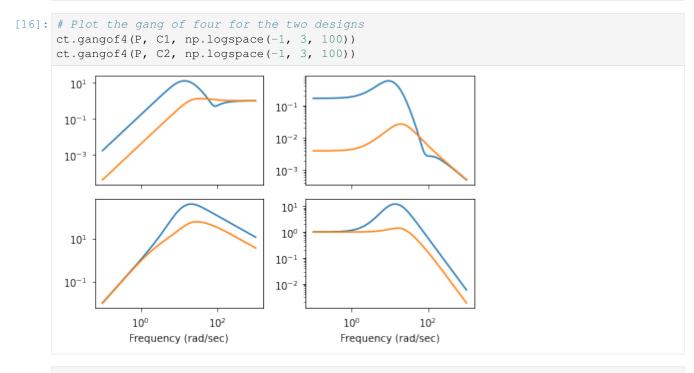
```
[14]: # Compute the feedback gain using eigenvalue placement
     wc = 10
     zc = 0.707
     eigs = np.roots([1, 2*zc*wc, wc**2])
     K = ct.place(A, B, eigs)
     kr = np.real(1/clsys.evalfr(0))
     print("K = ", np.squeeze(K))
     # Compute the estimator gain using eigenvalue placement
     wo = 20
     zo = 0.707
     eigs = np.roots([1, 2*zo*wo, wo**2])
     L = np.transpose(
         ct.place(np.transpose(A), np.transpose(C), eigs))
     print("L = ", np.squeeze(L))
     # Construct an output-based controller for the system
     C1 = ct.ss2tf(ct.StateSpace(A - B@K - L@C, L, K, 0))
     print("C(s) = ", C1)
     # Compute the loop transfer function and plot Nyquist, Bode
     L1 = P * C1
     plt.figure(); ct.nyquist_plot(L1, np.logspace(0.5, 3, 500))
     plt.figure(); ct.bode_plot(L1, np.logspace(-1, 3, 500));
```

```
K = [100. -35.86]
L = [28.28400.]
C(s) =
-1.152e+04 s + 4e+04
s^2 + 42.42 s + 6658
      1.5
      1.0
 Imaginary axis
      0.5
      0.0
     -0.5
    -1.0
    -1.5
                            -1.0
                                            -0.5
                                                             0.0
                                                                            0.5
           -1.5
                                            Real axis
       10^{3}
 Magnitude
10<sup>-1</sup>
                               10°
                                                                 10<sup>2</sup>
                                                10<sup>1</sup>
                                                                                  10^{3}
     -180
Phase (deg)
     -225
    -270
    -315
     -360
             10^{-1}
                               10°
                                                10<sup>1</sup>
                                                                                  10^{3}
                                                                 10^{2}
                                       Frequency (rad/sec)
```

```
[15]: # Modified control law
wc = 10
zc = 2.6
eigs = np.roots([1, 2*zc*wc, wc**2])
K = ct.place(A, B, eigs)
kr = np.real(1/clsys.evalfr(0))
print("K = ", np.squeeze(K))

# Construct an output-based controller for the system
C2 = ct.ss2tf(ct.StateSpace(A - B@K - L@C, L, K, 0))
print("C(s) = ", C2)
```

```
K = [100. 2.]
C(s) =
    3628 s + 4e+04
-----s^2 + 80.28 s + 156.6
```



## 8.2.3 Vertical takeoff and landing aircraft

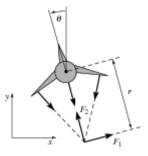
This notebook demonstrates the use of the python-control package for analysis and design of a controller for a vectored thrust aircraft model that is used as a running example through the text *Feedback Systems* by Astrom and Murray. This example makes use of MATLAB compatible commands.

Additional information on this system is available at

 $http://www.cds.caltech.edu/\sim murray/wiki/index.php/Python-control/Example:\_Vertical\_takeoff\_and\_landing\_aircraft$ 

### **System Description**

This example uses a simplified model for a (planar) vertical takeoff and landing aircraft (PVTOL), as shown below:



$$m\ddot{x}=F_1\cos heta-F_2\sin heta-c\dot{x}, \ m\ddot{y}=F_1\sin heta+F_2\cos heta-mg-c\dot{y}, \ J\ddot{ heta}=rF_1.$$

The position and orientation of the center of mass of the aircraft is denoted by  $(x, y, \theta)$ , m is the mass of the vehicle, J the moment of inertia, g the gravitational constant and c the damping coefficient. The forces generated by the main downward thruster and the maneuvering thrusters are modeled as a pair of forces  $F_1$  and  $F_2$  acting at a distance r below the aircraft (determined by the geometry of the thrusters).

Letting  $z=(x,y,\theta,\dot{x},\dot{y},\dot{\theta})$ , the equations can be written in state space form as:

$$\frac{dz}{dt} = \begin{bmatrix} z_4 \\ z_5 \\ z_6 \\ -\frac{c}{m} z_4 \\ -g - \frac{c}{m} z_5 \\ 0 \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \\ \frac{1}{m} \cos \theta F_1 + \frac{1}{m} \sin \theta F_2 \\ \frac{1}{m} \sin \theta F_1 + \frac{1}{m} \cos \theta F_2 \end{bmatrix}$$

#### LQR state feedback controller

This section demonstrates the design of an LQR state feedback controller for the vectored thrust aircraft example. This example is pulled from Chapter 6 (Linear Systems, Example 6.4) and Chapter 7 (State Feedback, Example 7.9) of Astrom and Murray. The python code listed here are contained the the file pytol-lqr.py.

To execute this example, we first import the libraries for SciPy, MATLAB plotting and the python-control package:

```
[1]: from numpy import * # Grab all of the NumPy functions
from matplotlib.pyplot import * # Grab MATLAB plotting functions
from control.matlab import * # MATLAB-like functions
%matplotlib inline
```

The parameters for the system are given by

```
[2]: m = 4 # mass of aircraft

J = 0.0475 # inertia around pitch axis

r = 0.25 # distance to center of force

g = 9.8 # gravitational constant

c = 0.05 # damping factor (estimated)
```

Choosing equilibrium inputs to be  $u_e = (0, mg)$ , the dynamics of the system  $\frac{dz}{dt}$ , and their linearization A about

equilibrium point  $z_e = (0, 0, 0, 0, 0, 0)$  are given by

$$\frac{dz}{dt} = \begin{bmatrix} z_4 \\ z_5 \\ z_6 \\ -g\sin z_3 - \frac{c}{m}z_4 \\ g(\cos z_3 - 1) - \frac{c}{m}z_5 \end{bmatrix} \qquad A = \begin{bmatrix} 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \\ 0 & 0 - g & -c/m & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & -c/m & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

```
[3]: # State space dynamics xe = [0, 0, 0, 0, 0] # equilibrium point of interest ue = [0, m*g] # (note these are lists, not matrices)
```

```
[4]: # Dynamics matrix (use matrix type so that * works for multiplication)
     # Note that we write A and B here in full generality in case we want
     # to test different xe and ue.
    A = matrix(
       [[ 0,
                 0, 0, 1, 0, 0],
               0, 0, 0, 1, 0],
0, 0, 0, 0, 1],
        [ 0,
         [ 0,
         [ 0, 0, (-ue[0]*sin(xe[2]) - ue[1]*cos(xe[2]))/m, -c/m, 0, 0], [ 0, 0, (ue[0]*cos(xe[2]) - ue[1]*sin(xe[2]))/m, 0, -c/m, 0],
         [0, 0, 0, 0, 0, 0]
     # Input matrix
    B = matrix(
        [[0, 0], [0, 0], [0, 0],
        [\cos(xe[2])/m, -\sin(xe[2])/m],
         [\sin(xe[2])/m, \cos(xe[2])/m],
         [r/J, 0]])
     # Output matrix
    C = matrix([[1, 0, 0, 0, 0, 0], [0, 1, 0, 0, 0]])
    D = matrix([[0, 0], [0, 0]])
```

To compute a linear quadratic regulator for the system, we write the cost function as

$$J = \int_0^\infty (\xi^T Q_{\xi} \xi + v^T Q_v v) dt,$$

where  $\xi = z - z_e$  and  $v = u - u_e$  represent the local coordinates around the desired equilibrium point  $(z_e, u_e)$ . We begin with diagonal matrices for the state and input costs:

```
[5]: Qx1 = diag([1, 1, 1, 1, 1, 1])
Qu1a = diag([1, 1])
(K, X, E) = lqr(A, B, Qx1, Qu1a); K1a = matrix(K)
```

This gives a control law of the form  $v = -K\xi$ , which can then be used to derive the control law in terms of the original variables:

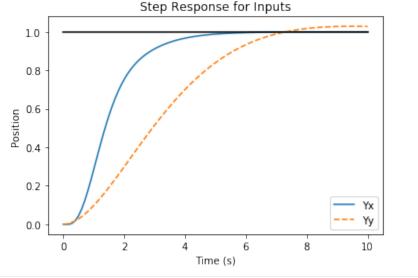
$$u = v + u_e = -K(z - z_d) + u_d.$$
  
where : math : ' $u_e = (0, mq)$ 'and : math : ' $z_d = (x_d, y_d, 0, 0, 0, 0)$ '

The way we setup the dynamics above, A is already hardcoding  $u_d$ , so we don't need to include it as an external input. So we just need to cascade the  $-K(z-z_d)$  controller with the PVTOL aircraft's dynamics to control it. For didactic purposes, we will cheat in two small ways:

- First, we will only interface our controller with the linearized dynamics. Using the nonlinear dynamics would require the NonlinearIOSystem functionalities, which we leave to another notebook to introduce.
- 2. Second, as written, our controller requires full state feedback (K multiplies full state vectors z), which we do not have access to because our system, as written above, only returns x and y (because of C matrix). Hence, we would need a state observer, such as a Kalman Filter, to track the state variables. Instead, we assume that we have access to the full state.

The following code implements the closed loop system:





The plot above shows the x and y positions of the aircraft when it is commanded to move 1 m in each direction. The following shows the x motion for control weights  $\rho = 1, 10^2, 10^4$ . A higher weight of the input term in the cost function causes a more sluggish response. It is created using the code:

```
[8]: # Look at different input weightings
Qu1a = diag([1, 1])
K1a, X, E = lqr(A, B, Qx1, Qu1a)
(continues on next page)
```

(continued from previous page)

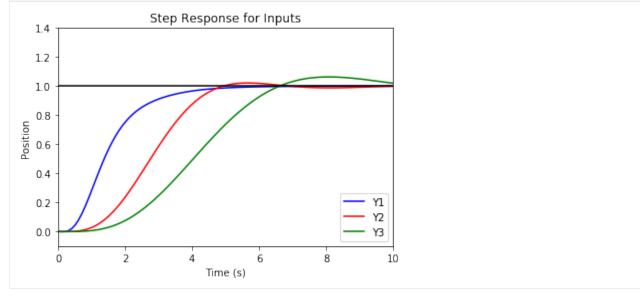
```
Hlax = H = ss(A-B*Kla,B*Kla*Xd,C,D)

Qulb = (40**2)*diag([1, 1])
Klb, X, E = lqr(A, B, Qx1, Qulb)
Hlbx = H = ss(A-B*Klb,B*Klb*Xd,C,D)

Qulc = (200**2)*diag([1, 1])
Klc, X, E = lqr(A, B, Qx1, Qulc)
Hlcx = ss(A-B*Klc,B*Klc*Xd,C,D)

[Y1, T1] = step(Hlax, T=linspace(0,10,100), input=0,output=0)
[Y2, T2] = step(Hlbx, T=linspace(0,10,100), input=0,output=0)
[Y3, T3] = step(Hlcx, T=linspace(0,10,100), input=0,output=0)
```

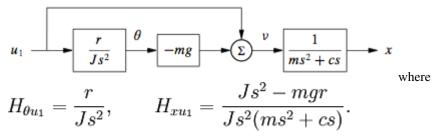
```
[9]: plot(T1, Y1.T, 'b-', T2, Y2.T, 'r-', T3, Y3.T, 'g-')
plot([0,10], [1, 1], 'k-')
title('Step Response for Inputs')
ylabel('Position')
xlabel('Time (s)')
legend(('Y1','Y2','Y3'),loc='lower right')
axis([0, 10, -0.1, 1.4])
show()
```



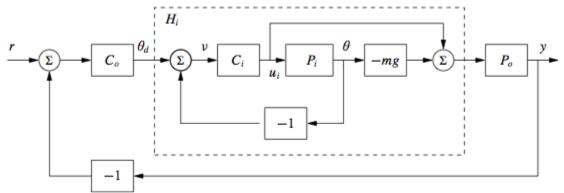
### Lateral control using inner/outer loop design

This section demonstrates the design of loop shaping controller for the vectored thrust aircraft example. This example is pulled from Chapter 11 (Frequency Domain Design) of Astrom and Murray.

To design a controller for the lateral dynamics of the vectored thrust aircraft, we make use of a "inner/outer" loop design methodology. We begin by representing the dynamics using the block diagram



The controller is constructed by splitting the process dynamics and controller into two components: an inner loop consisting of the roll dynamics  $P_i$  and control  $C_i$  and an outer loop consisting of the lateral position dynamics  $P_o$  and controller  $C_o$ .



The closed

inner loop dynamics  $H_i$  control the roll angle of the aircraft using the vectored thrust while the outer loop controller  $C_o$  commands the roll angle to regulate the lateral position.

The following code imports the libraries that are required and defines the dynamics:

```
[10]: from matplotlib.pyplot import * # Grab MATLAB plotting functions
     from control.matlab import * # MATLAB-like functions
[12]: # System parameters
     m = 4
                                   # mass of aircraft
     J = 0.0475
                                   # inertia around pitch axis
     r = 0.25
                                   # distance to center of force
     g = 9.8
                                   # gravitational constant
     c = 0.05
                                   # damping factor (estimated)
[13]: # Transfer functions for dynamics
     Pi = tf([r], [J, 0, 0])
                              # inner loop (roll)
```

For the inner loop, use a lead compensator

Po = tf([1], [m, c, 0])

```
[14]: k = 200
a = 2
b = 50
Ci = k*tf([1, a], [1, b]) # lead compensator
Li = Pi*Ci
```

# outer loop (position)

The closed loop dynamics of the inner loop,  $H_i$ , are given by

```
[15]: Hi = parallel(feedback(Ci, Pi), -m*g*feedback(Ci*Pi, 1))
```

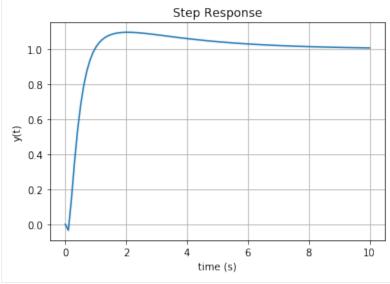
Finally, we design the lateral compensator using another lead compenstor

```
[16]: # Now design the lateral control system
a = 0.02
b = 5
K = 2
Co = -K*tf([1, 0.3], [1, 10]) # another lead compensator
Lo = -m*g*Po*Co
```

The performance of the system can be characterized using the sensitivity function and the complementary sensitivity function:

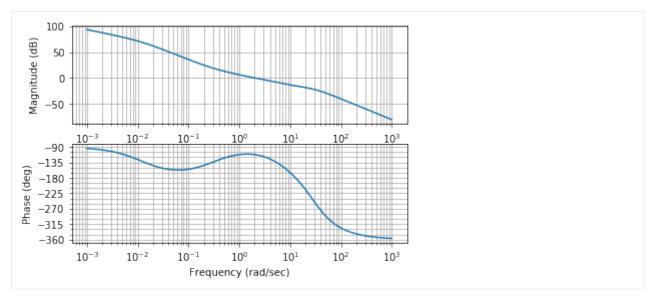
```
[17]: L = Co*Hi*Po
S = feedback(1, L)
T = feedback(L, 1)
```

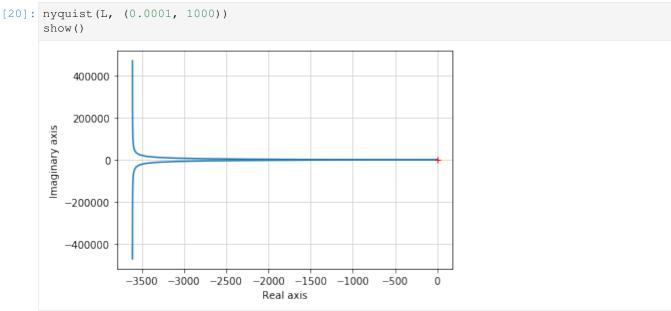




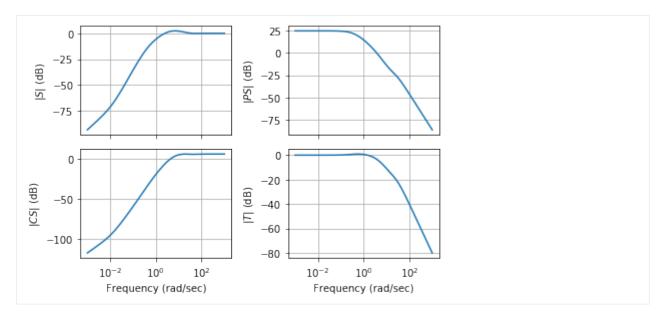
The frequency response and Nyquist plot for the loop transfer function are computed using the commands

```
[19]: bode(L) show()
```





[21]: gangof4(Hi\*Po, Co)



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### **Development**

You can check out the latest version of the source code with the command:

```
git clone https://github.com/python-control/python-control.git
```

You can run the unit tests with pytest to make sure that everything is working correctly. Inside the source directory, run:

```
pytest -v
```

or to test the installed package:

```
pytest --pyargs control -v
```

Your contributions are welcome! Simply fork the GitHub repository and send a pull request.

#### Links

- Issue tracker: https://github.com/python-control/jython-control/issues
- Mailing list: http://sourceforge.net/p/python-control/mailman/

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