

Motor Control with Arduino

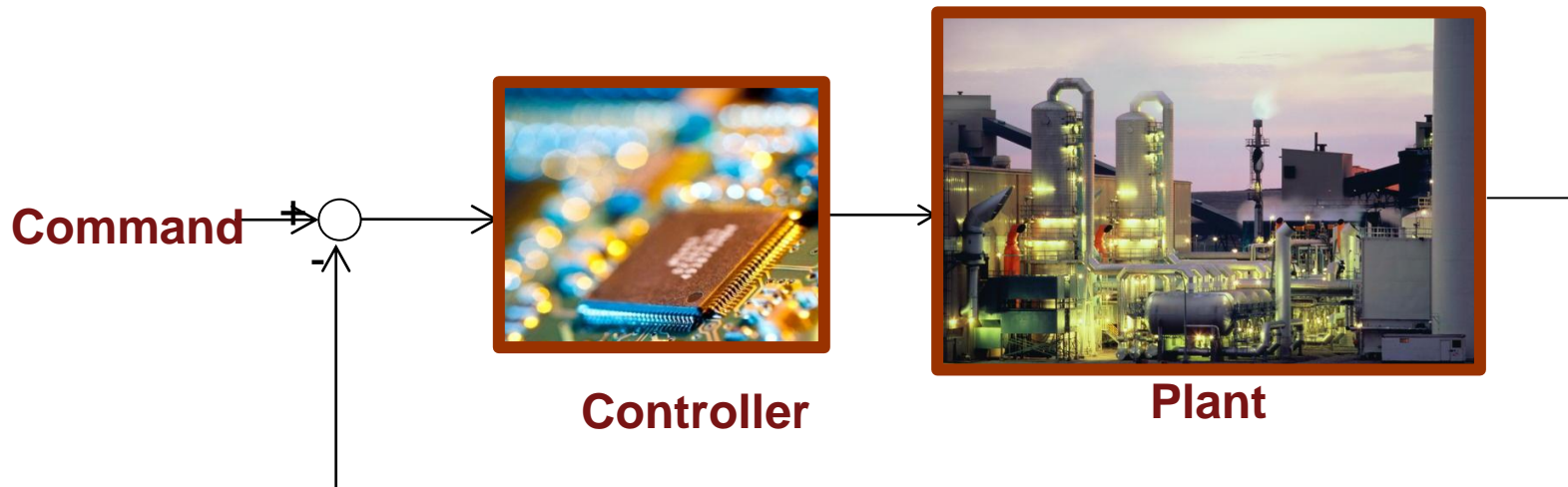
A Case Study in Data-Driven Modeling and Control Design

Toledo, 22-II-2013

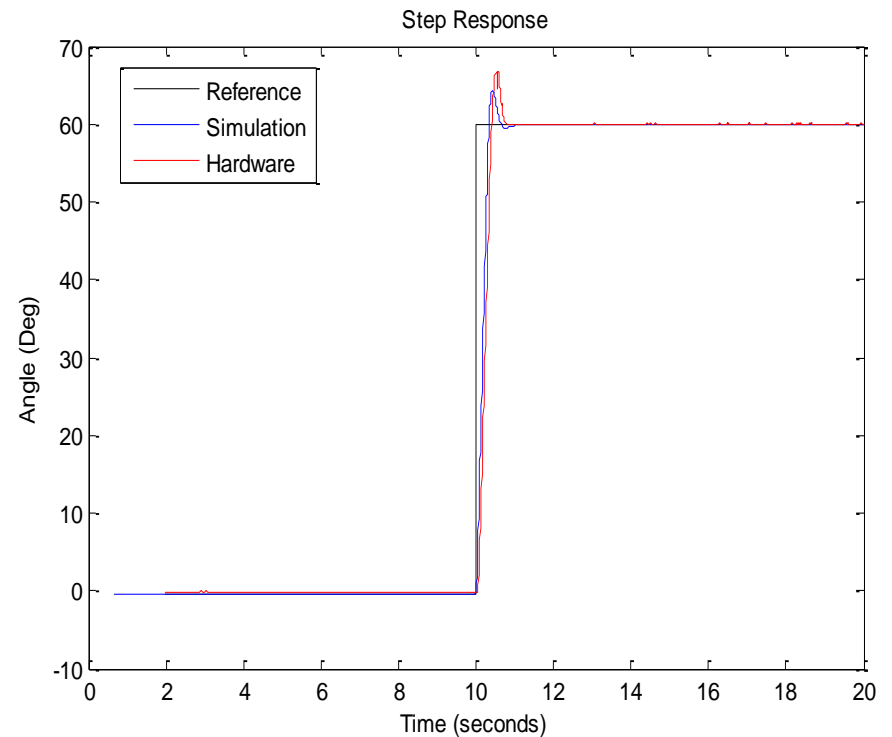
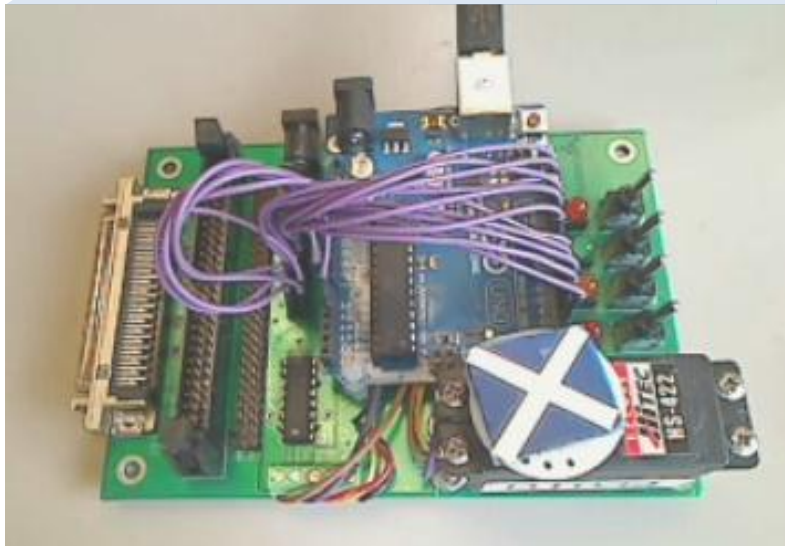
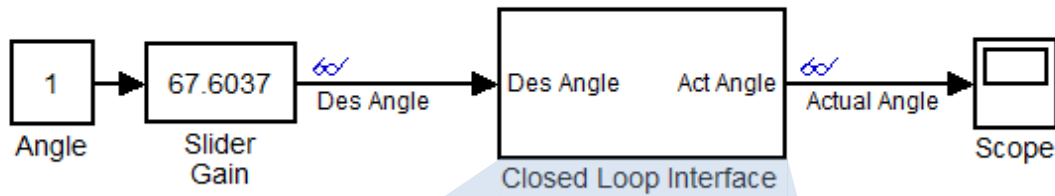
Why Data-Driven Control?

- Two approaches to modeling
 - **First-principles:** requires knowledge of math / physics of the system
 - **Data-driven:** requires measured input-output data

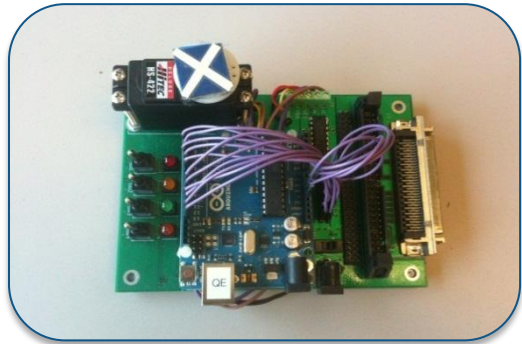
- Modeling from first-principles can get challenging



Demo: DC Motor Controller using Arduino Uno



Workflow

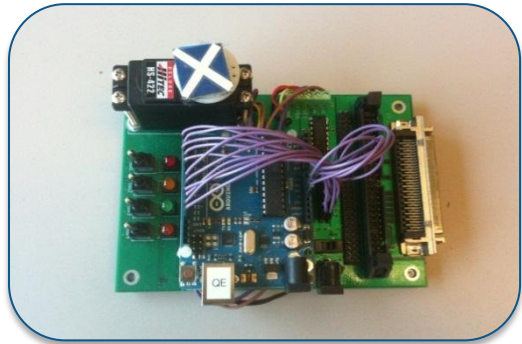


Hardware

1

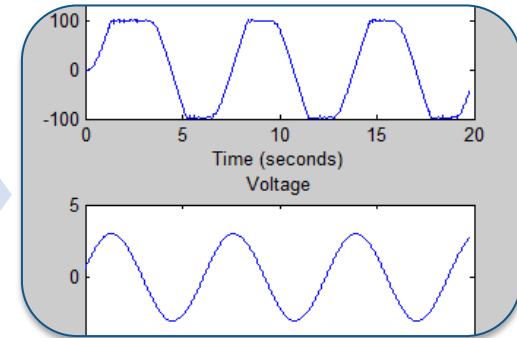
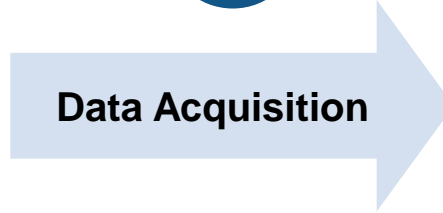
Data Acquisition

Workflow



Hardware

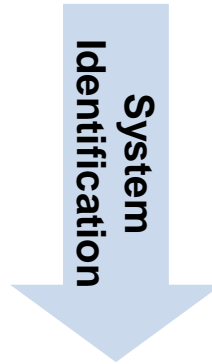
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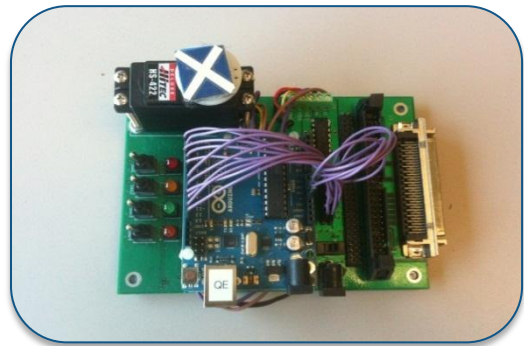
Datasets

System
Identification

2

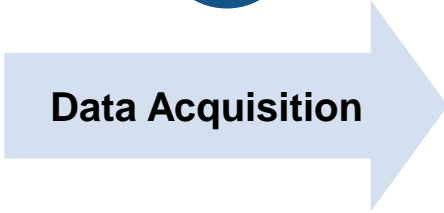


Workflow

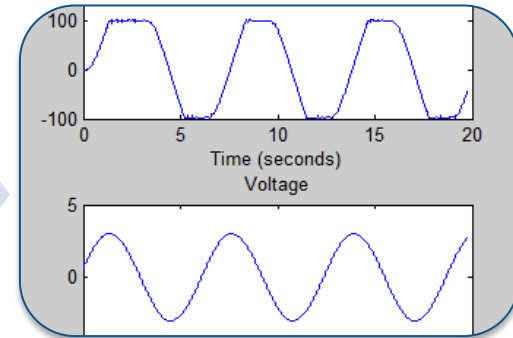


Hardware

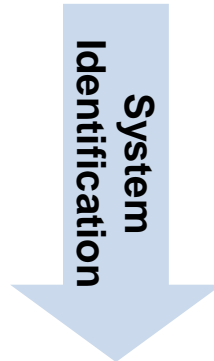
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Data Acquisition

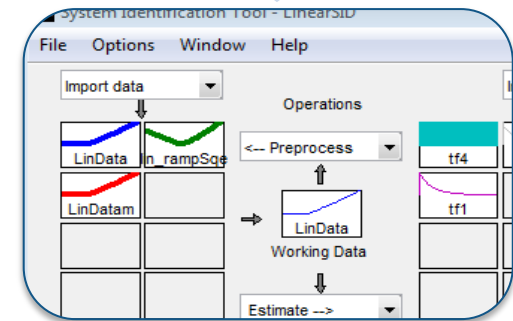


Datasets

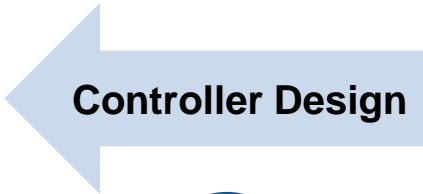


System Identification

2



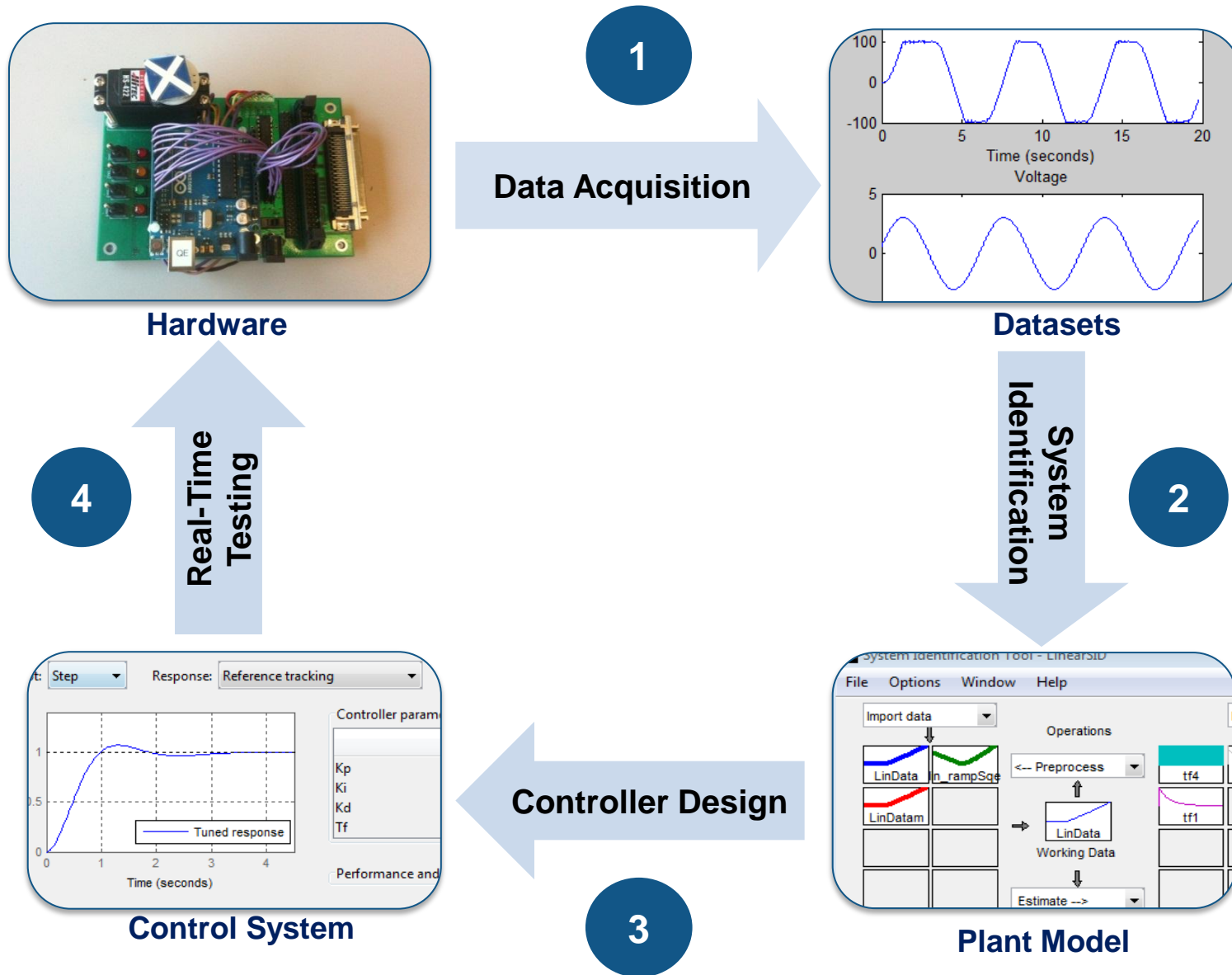
Plant Model



Controller Design

3

Workflow



Agenda

- Data Acquisition
 - Hardware setup
 - Run on Target Hardware

- System Identification
 - Linear model estimation
 - Nonlinear model estimation

- Controller Design
 - PID controller tuning
 - Desktop simulation with a nonlinear model

- Real-Time Testing and Controller Implementation
 - Deployment to Arduino Uno
 - Real-time controller evaluation

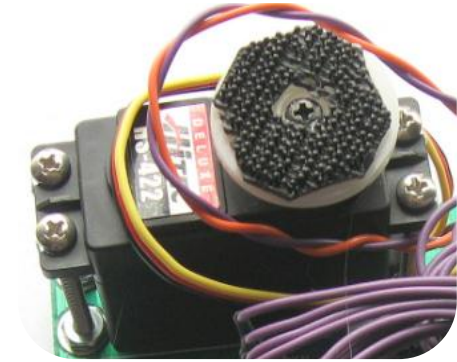
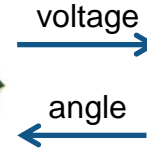
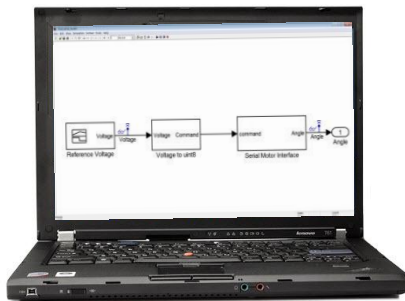
DATA ACQUISITION

Hardware Setup

Host Computer

Arduino Uno

DC Motor



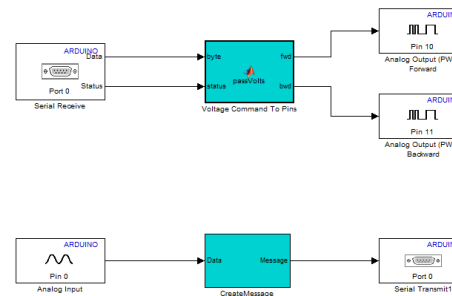
Serial communication

Motor driver

Run on Target Hardware

A Simulink feature in R2012a that:

- Creates an executable file from a model, and runs it on target hardware
- Is available from the model's Tools menu
 - Tools > Run on Target Hardware**
- Uses a *Target Installer* to install support packages for specific target hardware



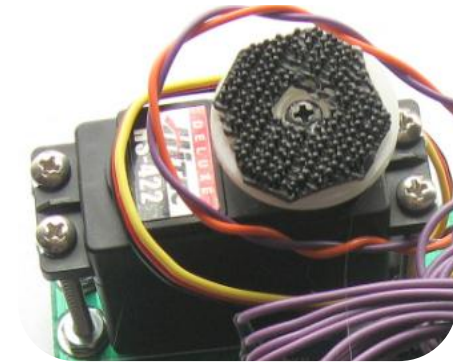
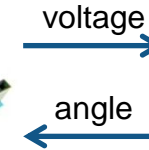
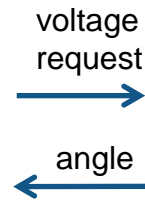
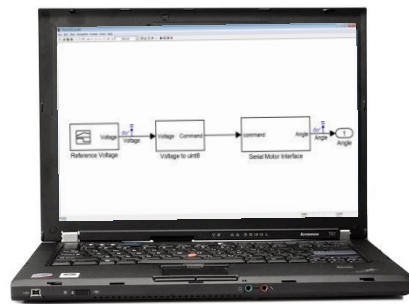
Requires Simulink®

Hardware Setup

Host Computer

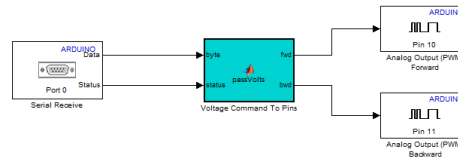
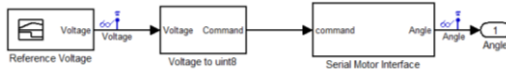
Arduino Uno

DC Motor



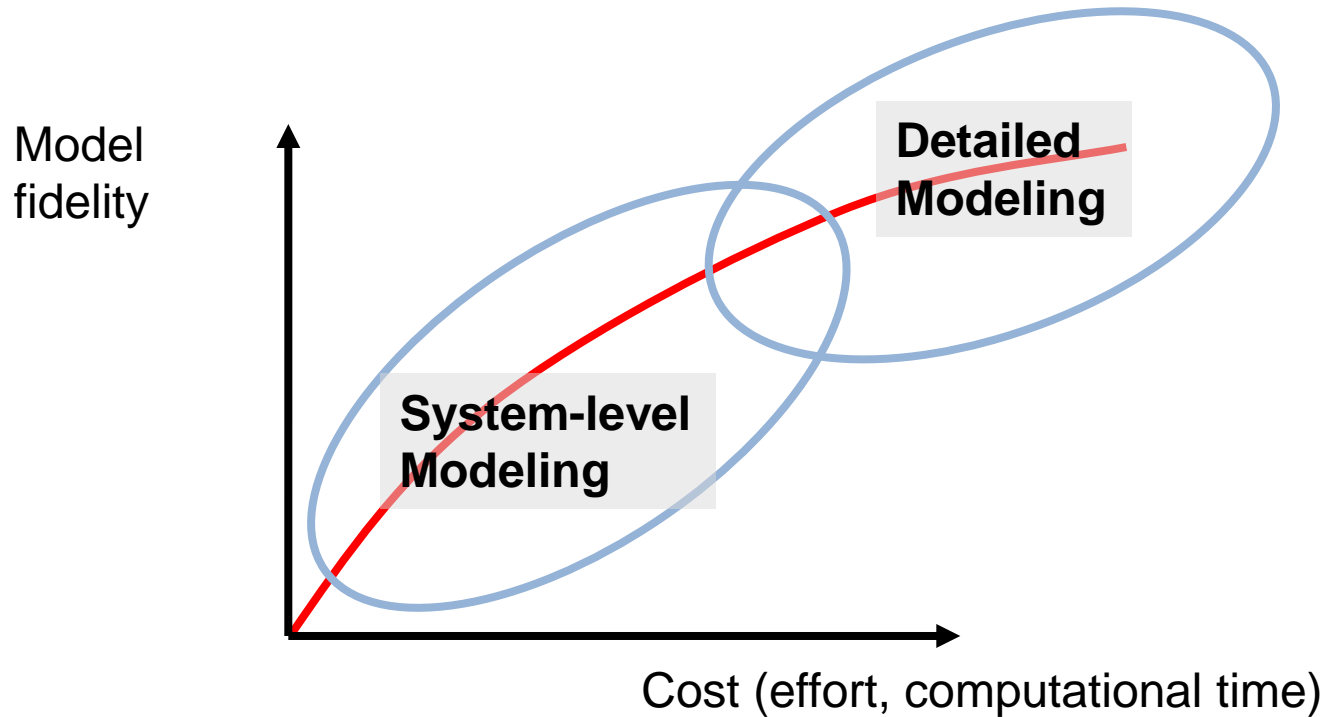
Serial communication
(using Run On
Target Hardware)

Motor driver



SYSTEM IDENTIFICATION

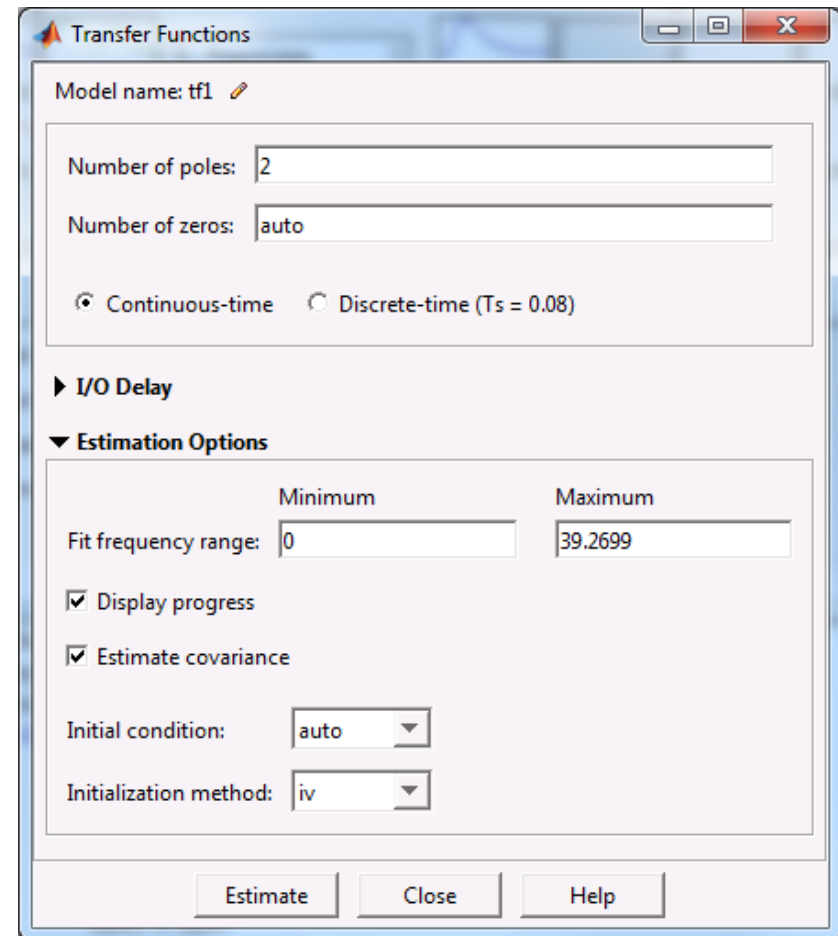
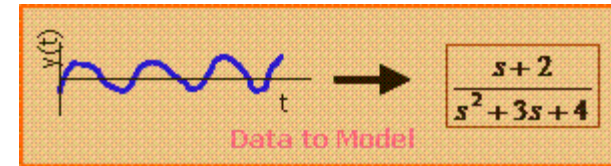
Model Fidelity vs. Cost



- Model the dynamics that matter for your analysis
- Balance cost and model fidelity

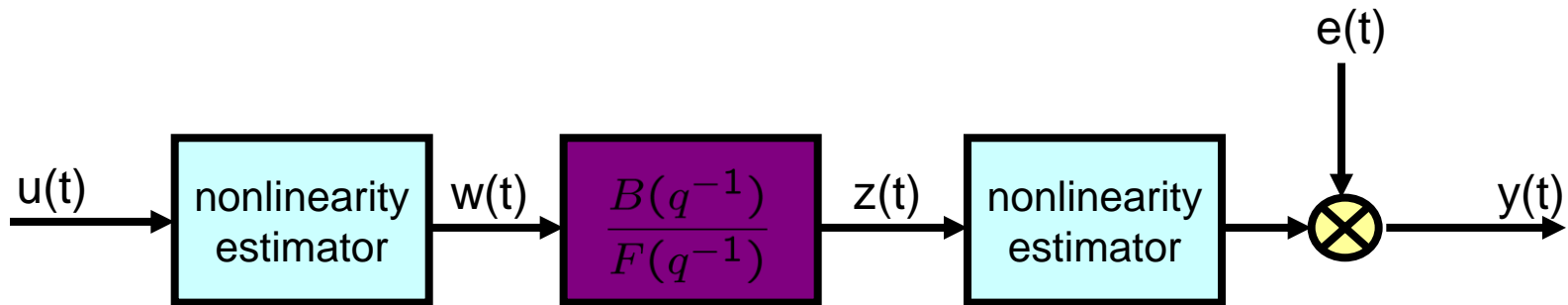
Linear System Identification

- Create continuous-time transfer functions from data
- Use either time or frequency domain data containing an arbitrary number of inputs and outputs
- Estimate other linear models: state-space, process models, and parametric models

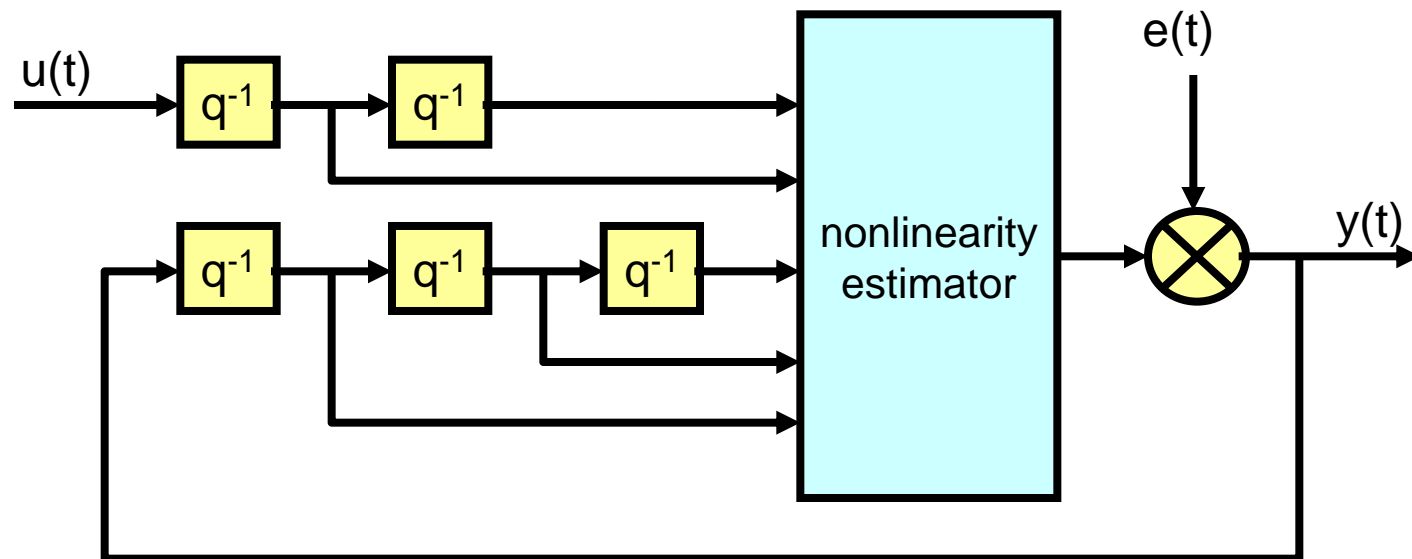


Requires System Identification Toolbox™

Hammerstein-Wiener structure



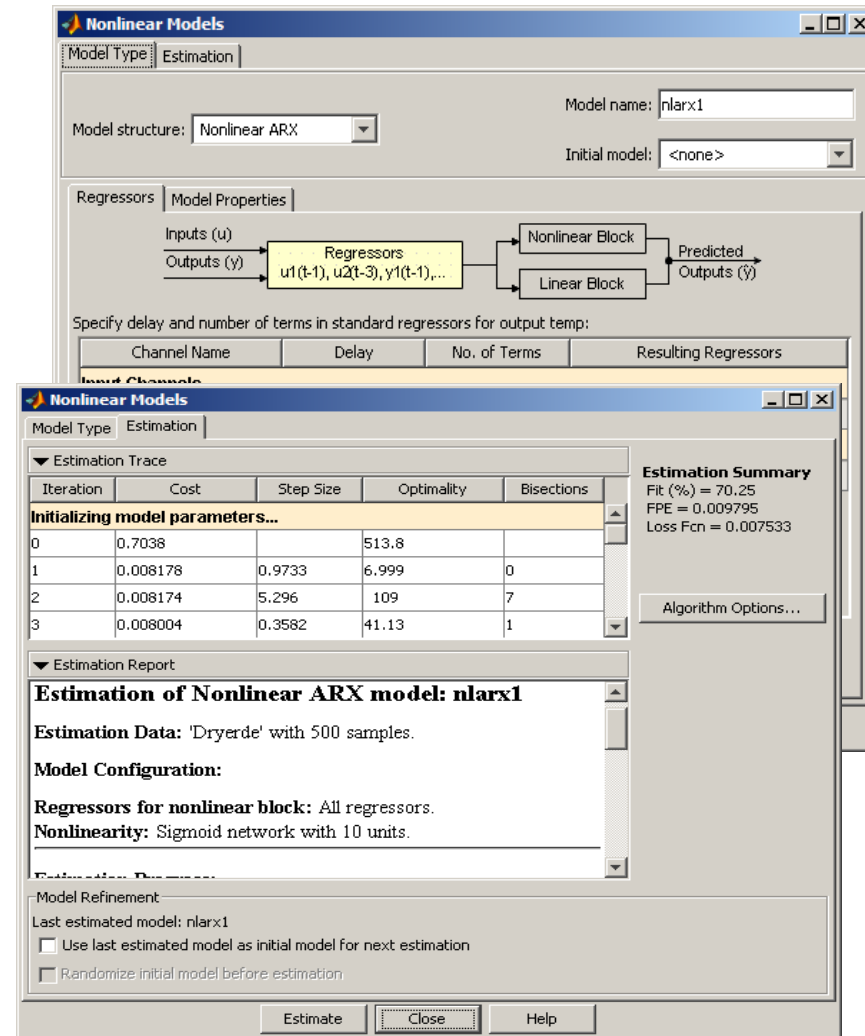
Example of NLARX model



$$y(t) = f[y(t-1), y(t-2), y(t-3), u(t-1), u(t-2)] + e(t)$$

Nonlinear System Identification

- Estimate Hammerstein-Wiener models and nonlinear ARX models
- Use a variety of dynamic nonlinearities such as wavelets, neural networks, piecewise linear, etc.
- Estimate signal saturation and dead-zone behaviors affecting linear systems
- Use custom regressors in nonlinear ARX models for greater flexibility in modeling



Nonlinear Models - Estimation

Model name: nlarx1
 Model structure: Nonlinear ARX
 Initial model: <none>

Regressors | Model Properties

Inputs (u) → Regressors (u1(t-1), u2(t-3), y1(t-1), ...) → Nonlinear Block / Linear Block → Predicted Outputs (ŷ)

Specify delay and number of terms in standard regressors for output temp:

Channel Name	Delay	No. of Terms	Resulting Regressors

Nonlinear Models - Estimation

Estimation Trace

Iteration	Cost	Step Size	Optimality	Bisections
Initializing model parameters...				
0	0.7038		513.8	
1	0.008178	0.9733	6.999	0
2	0.008174	5.296	109	7
3	0.008004	0.3582	41.13	1

Estimation Summary

Fit (%) = 70.25
 FPE = 0.009795
 Loss Fcn = 0.007533

Algorithm Options...

Estimation Report

Estimation of Nonlinear ARX model: nlarx1

Estimation Data: 'Dryerde' with 500 samples.

Model Configuration:

Regressors for nonlinear block: All regressors.
 Nonlinearity: Sigmoid network with 10 units.

Model Refinement

Last estimated model: nlarx1

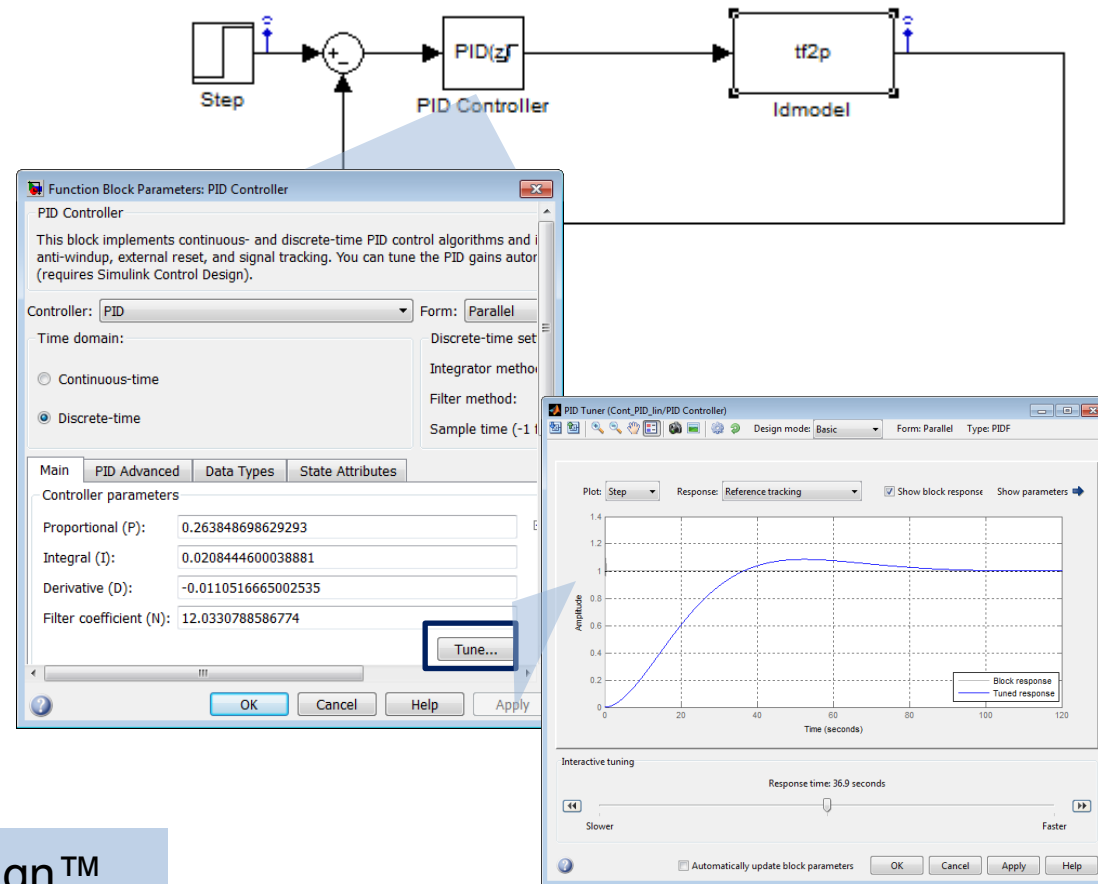
Use last estimated model as initial model for next estimation
 Randomize initial model before estimation

Estimate Close Help

CONTROLLER DESIGN

Automatic PID Tuning

- Automatically linearizes Simulink models and finds gain values to meet specifications
- Provides additional fine-tuning capability with simple sliders



Requires Simulink Control Design™

For more info, watch our webinar “*PID Control Made Easy*”.

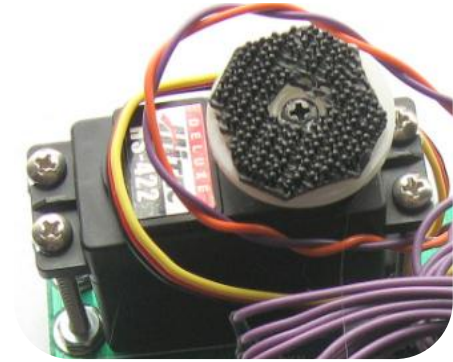
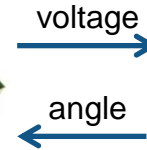
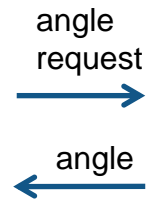
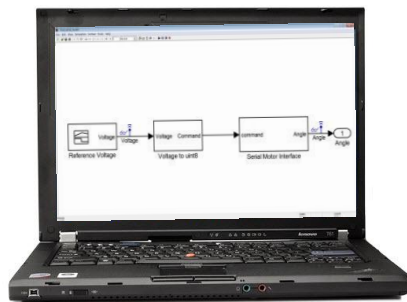
REAL-TIME TESTING AND CONTROLLER IMPLEMENTATION

Real-Time Testing

Host Computer

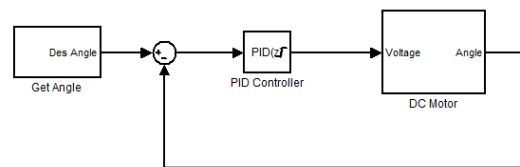
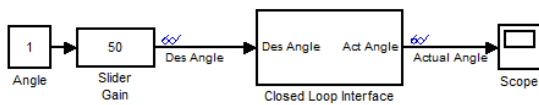
Arduino Uno

DC Motor

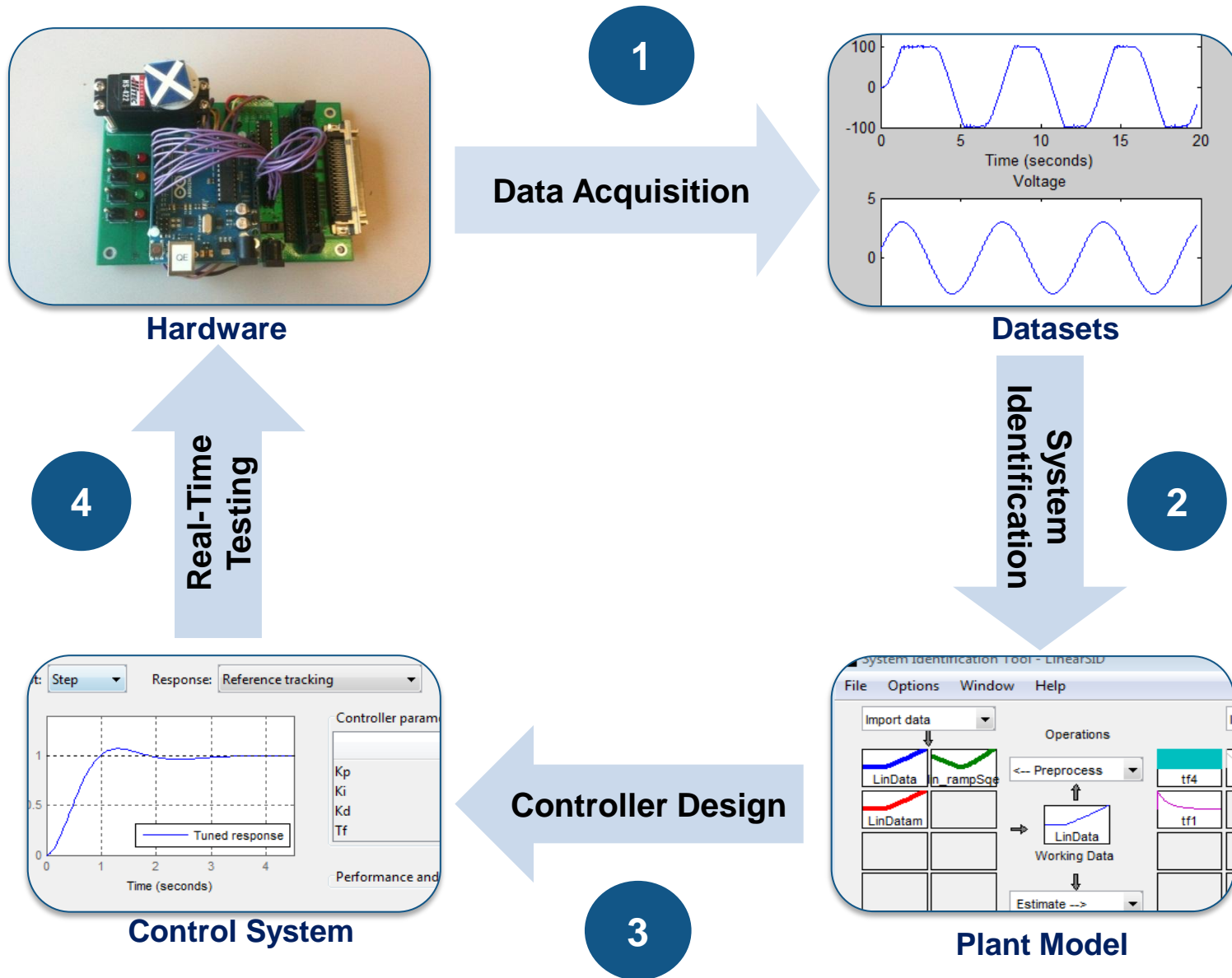


Serial communication
(Using Run On
Target Hardware)

Motor driver



Workflow



Summary

- MATLAB and Simulink support data-driven control design

- MATLAB and Simulink provide an environment for
 - Data acquisition
 - System identification
 - Control design
 - Real-Time testing