

# Model Predictive Control for Temperature Regulation

The purpose of this lab is to reinforce the concepts taught in class about dynamic process models, estimation, and model predictive control. It involves real-time sensing and calculations on an Arduino device that is a portable pocket-sized lab experiment.

## Problem Statement

### Model Predictive Control with an Empirical Model

1. Perform a doublet test on the system, varying the control output in manual mode. See the exercise at <http://apmonitor.com/do/index.php/Main/DataSimulation> to download sample data.
  2. From the manual-mode test calculate linear, first-order constants ( $\tau_p$ ,  $K_p$ ) fitting the data to the equation  $\tau_p \frac{dx}{dt} = -x + K_p u$ .
  3. Develop a model predictive controller based on the linear, first-order model.
  4. Perform set point changes and introduce disturbances (blow on the thermistor) to test the performance of the controller. Comment on the performance of the controller using the empirically derived constants.
  5. Tune the controller to improve performance.
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### Model Predictive Control with a First-Principles / Hybrid Model

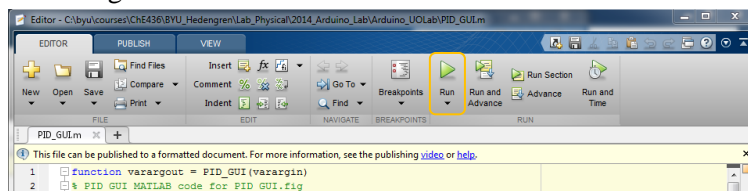
1. Derive the form of a first principles model for the relationship between input voltage and output temperature. There is no need to directly measure all parameters in the model; engineering judgement is sufficient.
2. Simulate the first principles model and compare the results to the data that were collected during the doublet test. Adjust the parameters in your model to align the model and measured values.
- 3a. For linear MPC: Linearize the adjusted first principles model and compare it to the empirical model. Comment on the similarities or differences between the two.
- 3b. For nonlinear MPC: Create a model predictive controller to adjust transistor voltage to regulate temperature.
4. Perform set point changes and introduce disturbances (blow on the thermistor) to test the performance of the controller. Comment on the performance of the controller using the empirically derived constants.
5. Tune the controller to improve performance.

## Setup for the Temperature Control Device

1. Plug in power supply to electrical outlet and USB connection to UO Lab computer
2. Download required files from course website and extract files from zipped archive
3. Open PID\_GUI.m from extracted folder (not zipped folder)

|                       |                   |             |       |
|-----------------------|-------------------|-------------|-------|
| apm                   | 5/8/2015 10:03 AM | File folder |       |
| ArduinoCode           | 5/8/2015 8:53 AM  | File folder |       |
| ArduinoDriver_Windows | 5/8/2015 8:53 AM  | File folder |       |
| Collected Data        | 5/8/2015 11:05 AM | File folder |       |
| Excel_FOPDT           | 5/8/2015 8:53 AM  | File folder |       |
| FirstPrinciples       | 5/8/2015 8:53 AM  | File folder |       |
| MatlabCode            | 5/8/2015 8:53 AM  | File folder |       |
| control.apm           | 5/8/2015 10:53 AM | APM File    | 1 KB  |
| control.csv           | 5/8/2015 10:22 AM | CSV File    | 1 KB  |
| mpc.m                 | 5/8/2015 11:03 AM | MATLAB Code | 2 KB  |
| mpc_init.m            | 5/8/2015 10:59 AM | MATLAB Code | 2 KB  |
| PID_GUI.m             | 5/8/2015 10:56 AM | MATLAB Code | 23 KB |

4. Click the green Run button

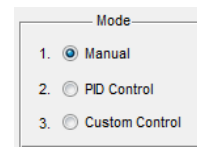


## Obtain a Dynamic Model from Step Test Data

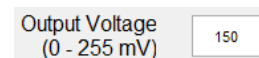
1. Click the green **Start** button.



2. Once the module has initialized, select **Manual** mode.



3. An input box will appear to allow changes to the input voltage.



4. Input manual values of output voltage to implement either a step, doublet, or PRBS input signal to the Arduino device. The *Enter* key is required to implement a change. The temperature may take a couple minutes to reach a new steady state value.

**Caution!**  
Unit is Hot!  
Do not Touch!

5. When the test is complete, select **Stop**.



6. Retrieve data from the Folder **Collected Data**. If multiple tests were performed, the data files are named according to the test time stamp.

## Determine a Linear, First-Order Model

Fit data to a FOPDT model using Excel, MATLAB, Python, or another analysis tool. There is an Excel template in the folder **Excel\_FOPDT**. Values from the generated data should be copied into the appropriate locations on the Excel worksheet as shown below.

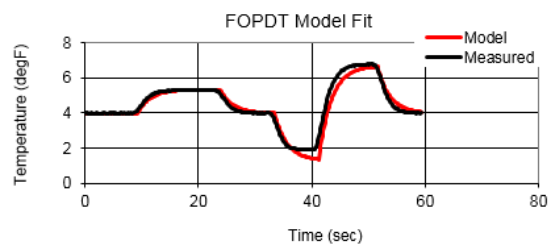
| Insert<br>time | These<br>Input | Values<br>Measured |
|----------------|----------------|--------------------|
| 0.01667        |                | 70                 |
| 0.16667        |                | 70                 |
| 0.31667        |                | 70                 |
| 0.46667        |                | 70                 |
| 0.61667        | 70             |                    |
| 0.76667        | 70             |                    |

While it is not necessary to modify the columns that calculate the model mismatch, it may be necessary to fill down the equations in these columns if the number of data points exceeds the template default.

| Don't change these columns, only copy down to match number of measurements |                |                    |                     |            |             |
|--|----------------|--------------------|---------------------|------------|-------------|
| Model  | Model<br>Slope | Model<br>Intercept | Model<br>with Delay | abs(error) | error^2     |
| 3.9885658  | 0              | 3.9885658          | 3.9885658           | 0          | 0           |
| 3.9885658  | 0              | 3.9885658          | 3.9885658           | 0.0127675  | 0.000163009 |
| 3.9885658  | 0              | 3.9885658          | 3.9885658           | 0.0100581  | 0.000101165 |
| 3.9885658  | 0              | 3.9885658          | 3.9885658           | 0.0023061  | 5.3181E-06  |

Values of  $K_p$  and  $\tau_p$  can be obtained by either manually changing the values in the parameters section or using Excel Solver to find the values that minimize either the Sum of Squared Errors or else the Sum of Absolute Errors.

| Model Parameters                   |             |
|------------------------------------|-------------|
| Kp (Gain)                          | 0.135127622 |
| tau (Time Constant)                | 2.229643365 |
| theta (Time Delay)                 | 2           |
| Doesn't seem to change with solver |             |
| Minimize Either of These           |             |
| Sum of Squared Errors              | 157.5540833 |
| Sum of Absolute Errors             | 154.7466485 |



## PID Controller Tuning

Once the linear, first order model ( $K_p$  and  $\tau_p$ ) is determined, use tuning correlations to select acceptable starting values for the PID controller ( $K_c$ ,  $\tau_I$ , and  $\tau_D$ ).

1. Click the green Start button.
2. Select Control mode.
3. Enter **Proportional (P)**, **Integral (I)**, and **Derivative (D)** terms for the PID controller.
  - a. Proportional (P) =  $K_c$

- b. Integral (I) =  $\frac{K_c}{\tau_I}$
  - c. Derivative (D) =  $K_c \tau_D$
- 4. Tune controller to achieve improved performance.
- 5. Select **Stop** when the test is complete.
- 6. Retrieve the saved data file from the **Collected Data** folder.

## MPC Tuning (Custom Control Option)

Once the linear, first order model ( $K_p$  and  $\tau_p$ ) is determined, implement the model in **control.apm** and tune the controller in **mpc\_init.m** to achieve acceptable performance.

- 1. Click the green Start button.
- 2. Select Control mode.
- 3. Enter controller setpoint.
- 4. Tune controller to achieve improved performance.
  - a. Common MV tuning parameters
    - i. COST = (+) minimize MV, (-) maximize MV
    - ii. DCOST = penalty for MV movement
    - iii. DMAX = maximum that MV can move each cycle
    - iv. FSTATUS = feedback status with 1=measured, 0=off
    - v. LOWER = lower MV bound
    - vi. MV\_TYPE = MV type with 0=zero-order hold, 1=linear interpolation
    - vii. STATUS = turn on (1) or off (0) MV
    - viii. UPPER = upper MV bound
  - b. Common CV tuning parameters
    - i. COST = (+) minimize MV, (-) maximize MV
    - ii. CV\_TYPE = CV type with 1= $\ell_1$ -norm, 2=squared error
    - iii. FSTATUS = feedback status with 1=measured, 0=off
    - iv. SP = setpoint with CV\_TYPE = 2
    - v. SPLO = lower setpoint with CV\_TYPE = 1
    - vi. SPHI = upper setpoint with CV\_TYPE = 1
    - vii. STATUS = turn on (1) or off (0) MV
    - viii. TR\_INIT = trajectory type, 0=dead-band, 1,2=trajectory
    - ix. TR\_OPEN = opening at initial point of traj compared to end
- 5. Select **Stop** when the test is complete.
- 6. Retrieve the saved data file from the **Collected Data** folder.