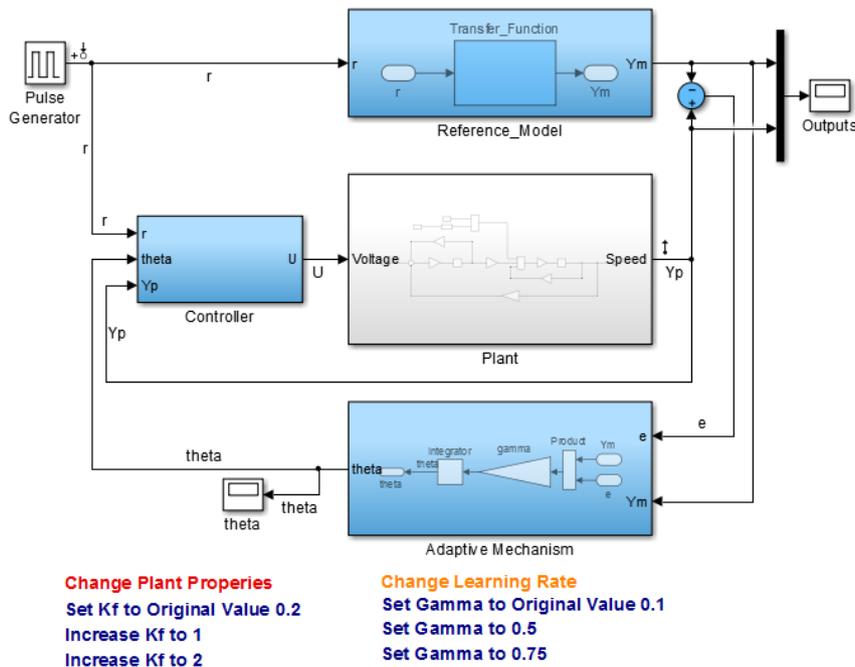


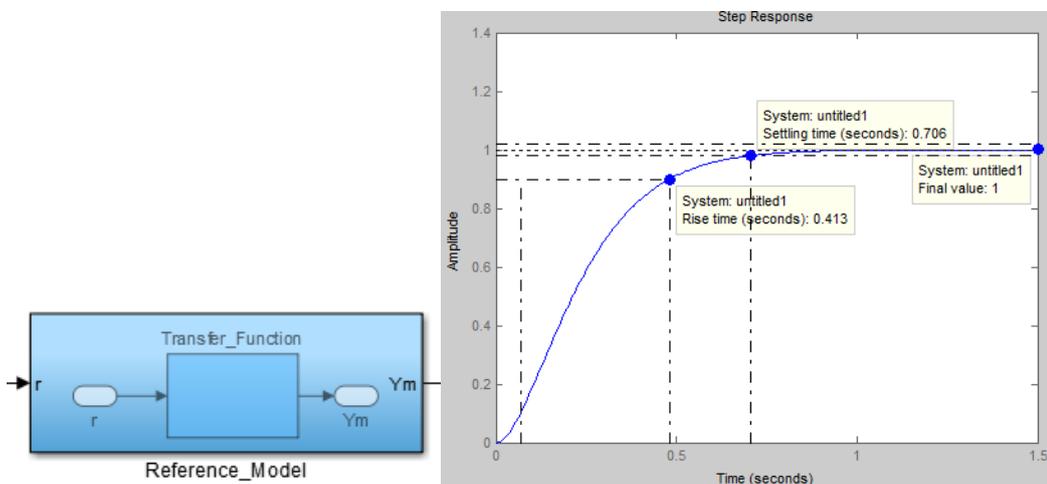
## Adaptive Controller Example

### Introduction:

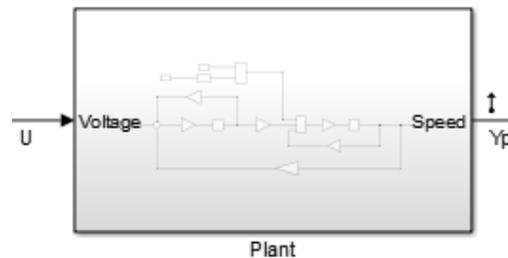
Objective of this example is to demonstrate how to design and model adaptive controller, tune and analyse its performance using Simulink®. For this example we have used direct adaptive method called Model Reference Adaptive Controller (MRAC). There are three main elements of this model: Reference Model, Plant Model and Adaptive Controller. Over all model looks like this:



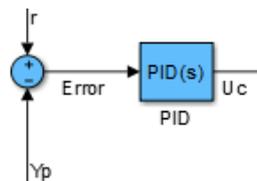
- 1.) Reference Model: This part of the controller captures\models the desired behaviour of closed-loop system. In other words, how you like your overall system to behave for a given input is modelled in this subsystem. In this example, reference behaviour is modelled as a transfer function. This can also come from closed-loop system specifications described in below figure as desired Rise Time (0.413sec), Settling Time (0.706sec) and Steady State Error (0). Reference Model output,  $Y_m$  is desired reference trajectory which Plant output ( $Y_p$ ) has to follow.



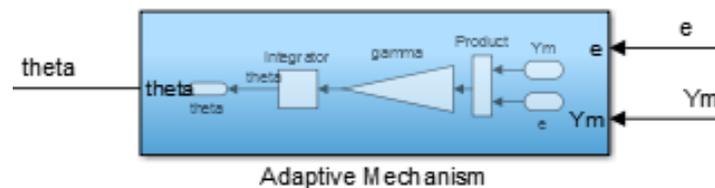
- 2.) Plant Model: In this example, plant is a DC Motor. One of the many motor parameters,  $K_f$  – Mechanical Damping, is considered to be varying. Initial value is assumed to be 0.2. And PID controller is tuned to achieve desired response with this initial value of  $K_f$ . Now as motor goes through ageing and impact of other environmental conditions,  $K_f$  changes, this will change motor behaviour. Plant output is  $Y_p$ . Hence controller has to adapt\change its parameter values to achieve desired response ( $Y_p - Y_m = \text{error } (e) = 0$ ).



- 3.) Adaptive Controller: There are two sub components of this controller.
- a. PID Controller: This part of controller is fixed and gains have been tuned for keeping initial plant condition in mind and to achieve overall stability. Output of PID controller is  $U_c$ .



- b. Adaptive Mechanism: This goal of this part of controller is to change its output ( $\theta$ ) based on error ( $e$ ) between plant output ( $Y_p$ ) and reference model output ( $Y_m$ ). How fast it can adapt (or change its output) depends on parameter called learning rate,  $\gamma$ . Higher the value of  $\gamma$ , faster it can adapt to any changes in plant. But there are some side effects also. Controller output ( $U$ ) is calculated by:  $U = U_c * \theta$ .

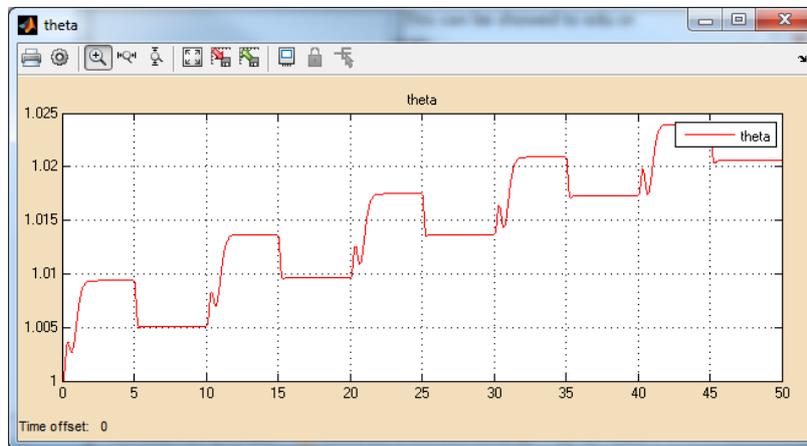


### Using This Example:

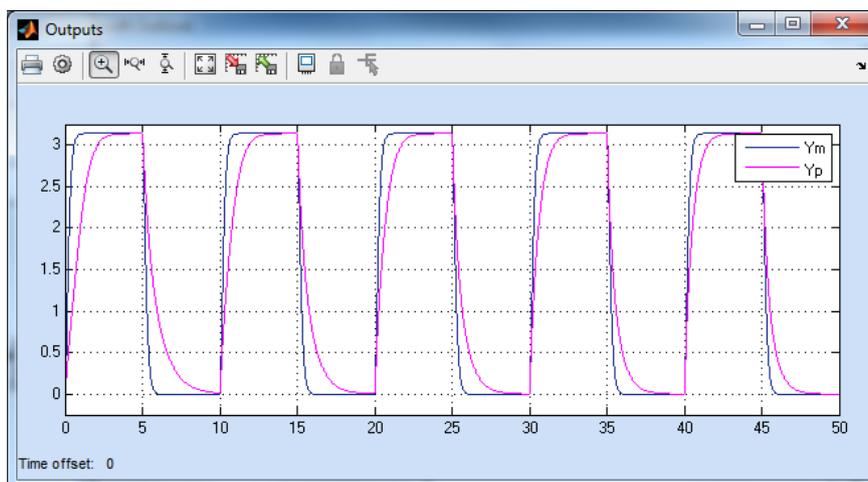
**Step 1:** Run this model with default values of plant, PID controller and learning rate. You can observe that overall all closed-loop system is behaving as per reference model. Look at the small difference between  $Y_m$  and  $Y_p$ . This small error is due to well-tuned PID controller for known plant values ( $K_f$  and others).



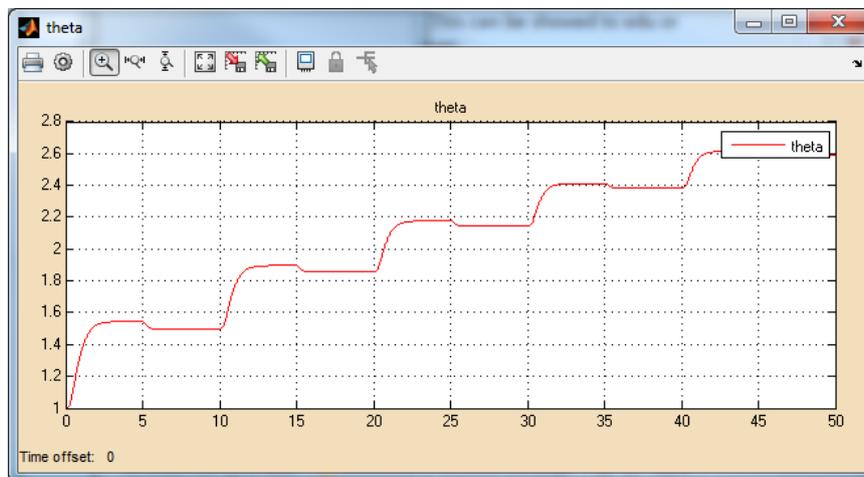
Also observe how theta, output of adaptive mechanism, changes as simulation progresses. It doesn't change much due to small error ( $e = Y_p - Y_m$ ).



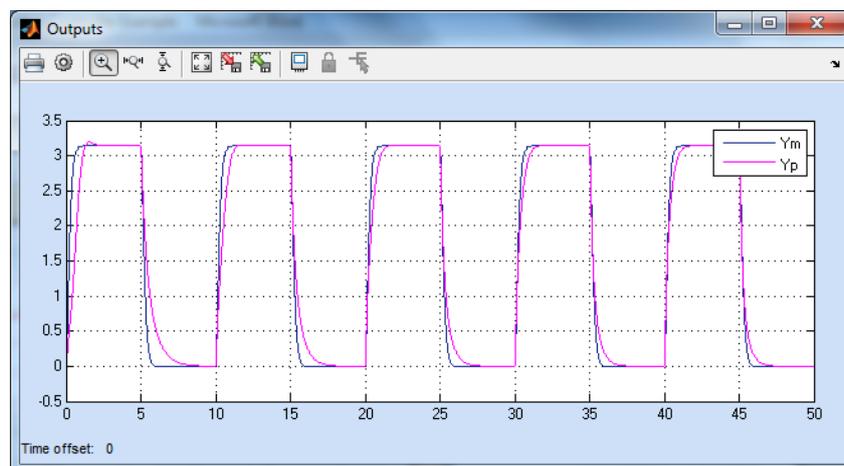
**Step 2:** Now change the plant behavior by increasing value of  $K_f$  from 0.2 to 1. This will result into different plant behavior initially and large error ( $e$ ). But if our adaptive mechanism is working fast, it should reduce error as simulation progresses. We can see from plot below,  $Y_p$  is approaching  $Y_m$  slowly. You can run simulation longer to see effect properly.

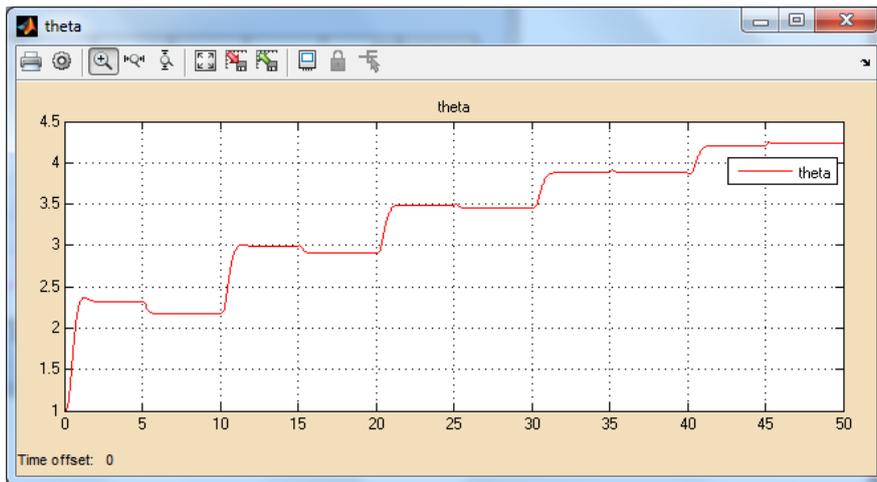


To understand how controller is adapting, look at *theta* plot below. As you can see, *theta* is going from its initial value 1 to 2.6 in 50 second and is not yet settled. If we run simulation long enough, *theta* will settle down at some higher values.



**Step 3:** Since results from Step 2 are not satisfactory, in terms of slow adoption rate, we can increase value of *gamma*, learning rate, to see if we can reduce the error ( $Y_p - Y_m$ ) to acceptable limit in 50 second. Go ahead and increase value of *gamma* from 0.1 to 0.5 and run model. You can see simulation results below, where plant response is improving during simulation due to fast changing *theta*. This is a result of faster learning rate, *gamma*.





**Step 4:**

To understand the impact of  $\gamma$ , plant parameters ( $K_f$  and others) and initial PID values, keep changing their values and understand how overall system behaves.