



$$y(t) = \begin{bmatrix} c_1 & c_2 & c_3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} + Du_i \quad (3)$$

The variables  $x_1$ ,  $x_2$  and  $x_3$  are time varying functions called internal states of the system. In our cavity tuning problem they can be estimated. A brief discussion of this is given later. These estimated states are used in the controller to obtain the control signal. The control signal  $u_i$  can be assumed to have two inputs,  $u$  and  $\Delta u_d$ , where  $u$  is the signal generated by the controller and  $\Delta u_d$  the input disturbance:

$$u_i(t) = u(t) + \Delta u_d(t) \quad (4)$$

Using the estimated states a time dependent sliding variable  $S$  is defined as follows:

$$S = \begin{bmatrix} g_1 & g_2 & g_3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = g^T x \quad (5)$$

The components  $g_1$ ,  $g_2$  and  $g_3$  of the matrix  $g^T$  are assumed to be known at this stage. However, later in this paper we discuss briefly a method to calculate them. To design a stable feedback loop we need to choose a suitable, positive definite Lyapunov function. In this particular case we can use the function as

$$V(t) = \frac{1}{2} S^2 \quad (6)$$

For global stability the Lyapunov function,  $V$ , must be positive definite, and its first derivative,  $\dot{V}$ , must be less than zero. In other words,

$$S\dot{S} < 0 \quad (7)$$

where  $\dot{S}$  is the time-derivative of Equation 5 and is given by

$$\begin{aligned} \dot{S} &= g^T \{ A x + b(u + \Delta u_d) \} \\ &= g^T \begin{bmatrix} a_1 & a_2 & a_3 \end{bmatrix} x + g^T b(u + \Delta u_d) \\ &= \begin{bmatrix} \alpha_1 & \alpha_2 & \alpha_3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} + \beta(u + \Delta u_d) \end{aligned} \quad (8)$$

with

$$a_1 = \begin{bmatrix} a_{11} \\ a_{21} \\ a_{31} \end{bmatrix}, \quad a_2 = \begin{bmatrix} a_{12} \\ a_{22} \\ a_{32} \end{bmatrix}, \quad a_3 = \begin{bmatrix} a_{13} \\ a_{23} \\ a_{33} \end{bmatrix}, \quad \alpha_i = g^T a_i, \quad i = 1, 2, 3 \\ \beta = g^T b \quad (9)$$

We can split the parameters,  $\alpha_1$ ,  $\alpha_2$ ,  $\alpha_3$ , and  $\beta$  into the nominal parameters,  $\alpha_1^\circ$ ,  $\alpha_2^\circ$ ,  $\alpha_3^\circ$ , and  $\beta^\circ$  and unknown parameters,  $\Delta\alpha_1$ ,  $\Delta\alpha_2$ ,  $\Delta\alpha_3$ , and  $\Delta\beta$  as follows:

$$\alpha_i = \alpha_i^\circ + \Delta\alpha_i \quad \left| \quad i = 1, 2, 3 \right. \\ \beta = \beta^\circ + \Delta\beta \quad (10)$$

The nominal parameters were calculated using the measured system matrices  $\{A, b\}$  and the matrix  $g^T$  of the controller. The unknown parameters are associated with the amount of system uncertainties excluding the disturbance signal. Also, let the control law,  $u$ , calculated by the controller, be divided into two parts: the continuous part,  $u_c$ , and the switching part,  $u_s$ . The

continuous part will hold the tuning phase error zero under ideal plant conditions; at the same time the switching part will drive the phase error zero whenever there is uncertainty. Thus

$$u = u_c + u_s \quad (11)$$

The control signals  $u_c$  and  $u_s$  are designed such that the Lyapunov stability condition dictated by Equation 7 is satisfied under the normal operating conditions. Also the control signals must not exceed the upper limits set by the bias regulator. Since  $u_c$  is used as the control function for the continuous part, we can group all the nominal parameters as follows:

$$u_c = -\frac{1}{\beta^\circ} \sum_{i=1}^3 \alpha_i^\circ x_i \quad (12)$$

Substituting Equations 10, 11 and 12 into Equation 8 and rearranging, we obtain

$$\dot{S} = \beta u_s + \beta \Delta u_d + \sum_{i=1}^3 (\Delta\alpha_i - \frac{\Delta\beta}{\beta^\circ} \alpha_i^\circ) x_i \quad (13)$$

The switching part of the control signal,  $u_s$ , is arranged with gains to overcome the uncertainties as follows:

$$u_s = -[k_1|x_1| + k_2|x_2| + k_3|x_3| + k_0] \text{sgn}S \quad (14)$$

The function  $\text{sgn}S$  in Equation 14 is the *signum* function which has a value either +1 or -1 when  $S \geq 0$  and  $S < 0$ , respectively. The constants,  $k_0$ ,  $k_1$ ,  $k_2$ ,  $k_3$  are selected such that Equation 7 is always satisfied. Clearly, with the following conditions on the gains, we can keep the loop stable if

$$k_i > \text{sup} \left| \frac{1}{\beta} (\Delta\alpha_i - \frac{\Delta\beta}{\beta^\circ} \alpha_i^\circ) \right| \quad i = 1, 2, 3 \\ k_0 > |\Delta u_d| \quad (15)$$

The abbreviation "sup" used in Equation 15 is pronounced as "supremum" to represent the maximum value of the function. If the system parameters  $\{A, b\}$  were accurately measured and if the variation due to temperature or other unknown effects is ignored, then the gains  $k_1$ ,  $k_2$  and  $k_3$  can be set to zero. Whereas the gain  $k_0$  is still required to handle the input disturbance,  $\Delta u_d$ , when the beam is turned on. The choice of these gains gives different weightings to the cost of control. Precise values can be set by actually working on the system. Also, when the feedback gains,  $k_0 \rightarrow k_3$ , are zero in Equation 14, then  $u_s$  is zero. Under this condition the control signal is  $u = u_c$ , obtained by solving Equation 12, which appears like a linear state feedback controller. Since this type of controller may give oscillatory control signal, a saturation function could be defined in place of  $\text{sgn}S$ . It is defined with a constant  $\delta$  such that  $\text{sgn}S = 1$  for  $S > \delta$ ,  $\text{sgn}S = -1$  for  $S < -\delta$ , and  $\text{sgn}S = S/\delta$  for  $\delta \geq S \geq -\delta$ .

#### IV. ESTIMATION OF THE STATES

From the previous section we noted that the required control signal,  $u$ , can be generated by solving Equations 5, 11, 12, and 14. We can do this provided the internal states,  $x_1$ ,  $x_2$  and  $x_3$  are known. In our problem they must be estimated. The state estimator is known as the "observer". We use the output signal,  $y(t)$ , and the input signal,  $u_i(t)$ , and obtain a standard Luenberger observer. A simple design technique is discussed by

Kailath<sup>5</sup>. Hence, we simply quote the equation below:

$$\dot{\hat{x}} = A\hat{x} + bu_i + m[y - C\hat{x}] - mDu_i \quad (16)$$

Where,  $\hat{x}$  is the estimated state vector used to calculate the sliding variable,  $S$ , and  $m$  is the feedback gain vector. This gain vector is obtained from the system parameters,  $\{A, C\}$ , and an arbitrary set of eigenvalues<sup>5</sup>. As a rule of thumb, the eigenvalues of the observer are chosen such that the observer states converge to actual values almost 10 times faster than the controller eigenvalues corresponding to  $g_1, g_2$ , and  $g_3$ . For designing the observer we have assumed that the input signal,  $u_i(t)$ , is measurable, meaning the disturbance signal,  $\Delta u_d$ , is accessible. In other words, Equation 16 will not estimate the states accurately when the beam comes on and hence may give problems, especially when the eigenvalues are chosen close to the controller. Further work is underway to overcome the observer defects.

For overall stability the eigenvalues of the observer and controller must be negative. The  $g$  matrix for the controller is selected by trial and error method or by using eigenvalue assignment technique shown in Reference 6. In both cases the equivalent closed loop system, described by

$$\dot{x} = [A - b(g^T b)^{-1} g^T A] x, \quad (17)$$

must have negative eigenvalues for stability. Equation 17 is obtained by substituting the condition<sup>7</sup>  $\dot{S} \equiv 0$  in Equation 8 and using the resulting expression for the equivalent control signal,  $u_i$ , in Equation 1. When  $\dot{S} \equiv 0$  one of the eigenvalues of Equation 17 is zero<sup>6</sup>. Hence, for our system we specify only two eigenvalues,  $\lambda_1$  and  $\lambda_2$ , and ignore the third. The  $g$  matrix is then obtained from the following equation:

$$g^T = q^T \alpha(A), \quad (18)$$

where the function  $\alpha(A) = (A - \lambda_1)(A - \lambda_2)$ , and the matrix,  $q^T$  is equal to the last row of the inverse of the controllability matrix of the system (Equation 1), and the symbol  $T$  is used to signify the transpose of the matrix.

## V. IMPLEMENTATION AND SIMULATION

The feedback loop can be implemented, as always, in two ways, using analogue or digital circuits. A schematic layout for analogue implementation is shown in Figure 2. The controller implementation would require a multiplexer to determine the sign change in the sliding variable. For digital implementation, a DSP chip, TMS320C30, from Texas Instruments with a 32-bit floating point multiplication and accumulation time of 60ns can compute the control signal in under  $5\mu s$ , in real time.

We have simulated the loop performance with the controller at  $5\mu s$  sampling rate in Figure 3, with a step disturbance signal of  $\Delta u_d = +0.1V$  between 5 ms and 10 ms. Various parameters are shown in Figure 3. A saturation function with  $\delta = 1 \times 10^{-6}$  is used in place of  $\text{sgn}S$ . Clearly the output transients are controlled under less than  $0.4^\circ$ . At this stage it is recalled that the switching part of the control signal must not be made zero; otherwise the output of the system will become unbounded. This is because one of the eigenvalues of Equation 17 is close to zero. Also, the controllability matrix of the system is observed to be very close to singularity. Hence all the feedback parameters must be carefully chosen.

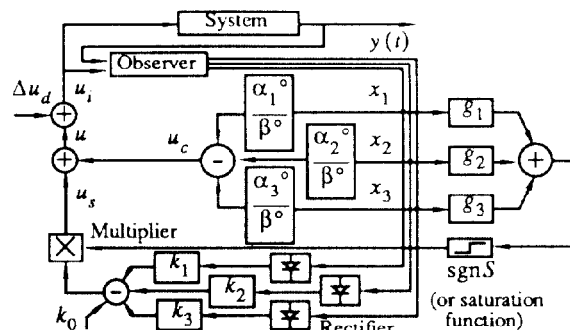
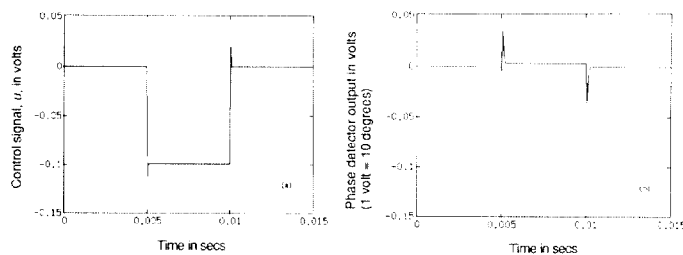


Figure 2. Analogue implementation of the sliding mode controller



$$\lambda_{Observer} = [-25 \ -50 \ -75] \times 10^4$$

$$\lambda_{Controller} = [-25 \ -250 \ 0] \times 10^3$$

$$k_0 = 0.11, k_1 = 200, k_2 = 100, k_3 = 200$$

$$g = \begin{bmatrix} 5.61223 \\ 19.90693 \\ -1.02001 \end{bmatrix} \times 10^{-3}$$

Figure 3. (a) Control signal,  $u$ , and (b) the phase error signal,  $y$ , under a step disturbance,  $\Delta u_d = 0.1V$ .

## VI. CONCLUSIONS

We have shown a modern control technique to design a robust feedback controller such as the "sliding-mode" starting from an experimental "Bode diagram" of the system. We retain all the simplicity of the state feedback controller and add robustness to handle variation in tuning errors due to beam loading or other uncertainties on the system. Although the controller is robust, a non-robust state estimator may give problems unless the eigenvalues are carefully selected.

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