ELEC4410

Control System Design

Lecture 6: Affine Parameterisation - Open Loop Unstable Model. Anti-windup Schemes

School of Electrical Engineering and Computer Science The University of Newcastle



Affine Parameterisation - Open Loop Unstable Model.



- Affine Parameterisation Open Loop Unstable Model.
- Saturation and Slew Rate Limitations.



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- Saturation and Slew Rate Limitations.
- Anti-windup Schemes.



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Reference: Control System Design, Goodwin, Graebe & Salgado.



Recall for affine parameterisation for the stable case:

$$C(s) = \frac{Q(s)}{1 - Q(s)G_o(s)}$$





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- To ensure $T_o(s)$, $S_o(s)$, $S_{io}(s)$ and $S_{uo}(s)$ are stable, we still need Q(s) stable and proper. In addition, we add further interpolation constraints:



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• $(1 - Q(s)G_o(s))G_o(s)$ stable.

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and assume all the poles of $A_o(s)$ are unstable.

Next we choose

$$Q(s) = \frac{\tilde{P}(s)}{\tilde{E}(s)}$$

where $\tilde{E}(s)$ is stable. In particular the zeros of $\tilde{E}(s)$ lie in a desirable region of the complex plane.



As we need unstable poles of $G_o(s)$ to be zeros in Q(s), we can write

$$Q(s) = \frac{A_o(s)\bar{P}(s)}{\tilde{E}(s)} \quad ; \quad \tilde{P}(s) = A_o(s)\bar{P}(s)$$



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> Then Q(s) is stable and has zeros to cancel unstable poles of $G_o(s)$.



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- We require the unstable poles of $G_o(s)$ to be zeros in $S_o(s)$.

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$$= 1 - \frac{\bar{P}(s)B_{o}(s)}{\tilde{E}(s)}$$
$$= \frac{\tilde{E}(s) - \bar{P}(s)B_{o}(s)}{\tilde{E}(s)}$$



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$$= 1 - \frac{\bar{P}(s)B_{o}(s)}{\tilde{E}(s)}$$
$$= \frac{\tilde{E}(s) - \bar{P}(s)B_{o}(s)}{\tilde{E}(s)}$$

• We thus require $A_o(s)$ to be a factor of $\tilde{E}(s) - \bar{P}(s)B_o(s)$.



Which we can write as

$$\tilde{E}(s) - \bar{P}(s)B_o(s) = \bar{L}(s)A_o(s)$$

or

$$\bar{L}(s)A_o(s) + \bar{P}(s)B_o(s) = \tilde{E}(s)$$
(1)

This is a standard pole assignment problem and choosing a desired $\tilde{E}(s)$ and the orders of $\bar{L}(s)$, $\bar{P}(s)$ will result in a unique solution.



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This is a standard pole assignment problem and choosing a desired $\tilde{E}(s)$ and the orders of $\bar{L}(s)$, $\bar{P}(s)$ will result in a unique solution.

Note: We set $\tilde{E}(s) = E(s)F(s)$.



Now we have

$$Q(s) = \frac{\tilde{P}(s)}{\tilde{E}(s)}$$

then

$$C(s) = \frac{Q(s)}{1 - Q(s)G_o(s)}$$

=
$$\frac{\tilde{P}(s)A_o(s)}{\tilde{E}(s)A_o(s) - \tilde{P}(s)B_o(s)}$$

=
$$\frac{\bar{P}(s)}{\bar{L}(s)}$$
(3)



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Step 2 Given $A_o(s)$, $B_o(s) \& \tilde{E}(s)$ solve $\bar{L}(s)A_o(s) + \bar{P}(s)B_o(s) = \tilde{E}(s)$ for a unique $\bar{L}(s)$ and $\bar{P}(s)$ that we denote as L(s) and P(s).



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Step 3

$$C(s) = \frac{P(s)}{L(s)}$$



Given a solution to equation (1), standard results in algebra state that any other solution can be expressed as

$$\frac{\bar{L}(s)}{E(s)} = \frac{L(s)}{E(s)} - Q_u(s)\frac{B_o(s)}{E(s)}$$
(4)
$$\frac{\bar{P}(s)}{E(s)} = \frac{P(s)}{E(s)} + Q_u(s)\frac{A_o(s)}{E(s)}$$
(5)

where $Q_u(s)$ is any stable proper transfer function having no undesirable poles.



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$$\frac{\bar{P}(s)}{E(s)} = \frac{P(s)}{E(s)} + Q_u(s)\frac{A_o(s)}{E(s)}$$
(7)

where $Q_u(s)$ is any stable proper transfer function having no undesirable poles.

Substitute (4) and (5) into (3) and we get

$$C(s) = \frac{\frac{P(s)}{E(s)} + Q_u(s)\frac{A_o(s)}{E(s)}}{\frac{L(s)}{E(s)} - Q_u(s)\frac{B_o(s)}{E(s)}}$$



Now say $A_o(s)$ contains both desirable and undesirable poles,

$$A_o(s) = A_d(s)A_u(s)$$

then we can write $E(s) = A_d(s)E(\bar{s})$ giving the pole assignment problem of

$$A_d(s)A_u(s)\bar{L}(s) + B_o(s)\bar{P}(s) = A_d(s)\bar{E}(s)F(s)$$

clearly this requires $\overline{P}(s) = \widetilde{P}(s)A_d(s)$.



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Therefore we have cancellations hence,

$$A_u(s)\bar{L}(s) + B_o(s)\tilde{P}(s) = \bar{E}(s)F(s)$$
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$$A_u(s)\bar{L}(s) + B_o(s)\tilde{P}(s) = \bar{E}(s)F(s)$$
(10)

For the unstable open loop case where the plant possesses some desirable poles then the same method as stated above applies except the pole assignment problem becomes equation (8).



The nominal complementary sensitivity for the class of stabilising controllers is,

$$\begin{split} T_{o}(s) &= \frac{B_{o}(s)\bar{P}(s)}{A_{o}(s)\bar{L}(s) + B_{o}(s)\bar{P}(s)} \\ &= \frac{B_{o}(s)\left(\frac{P(s)}{E(s)} + Q_{u}(s)\frac{A_{o}(s)}{E(s)}\right)}{A_{o}(s)\left(\frac{L(s)}{E(s)} - Q_{u}(s)\frac{B_{o}(s)}{E(s)}\right) + B_{o}(s)\left(\frac{P(s)}{E(s)} + Q_{u}(s)\frac{A_{o}(s)}{E(s)}\right)}{\frac{B_{o}(s)P(s)}{E(s)} + \frac{Q_{u}(s)B_{o}(s)A_{o}(s)}{E(s)}}{\frac{A_{o}(s)L(s)}{E(s)} + \frac{B_{o}(s)P(s)}{E(s)}}{\frac{B_{o}(s)P(s) + Q_{u}(s)B_{o}(s)A_{o}(s)}{E(s)}} \\ &= \frac{B_{o}(s)P(s) + Q_{u}(s)B_{o}(s)A_{o}(s)}{E(s)} \\ &= \frac{B_{o}(s)P(s) + Q_{u}(s)B_{o}(s)A_{o}(s)}{E(s)F(s)} \end{split}$$


Affine Parameterisation - Open Loop Unstable Model

Given the controller parameterisation for unstable plants as

$$C(s) = \frac{\frac{P(s)}{E(s)} + Q_u(s)\frac{A_o(s)}{E(s)}}{\frac{L(s)}{E(s)} - Q_u(s)\frac{B_o(s)}{E(s)}}$$

and that u = Ce where e is the error signal we can write (note the argument s has been dropped)

$$u = Ce$$

$$= \left(\frac{\frac{P}{E} + Q_u \frac{A_o}{E}}{\frac{L}{E} - Q_u \frac{B_o}{E}}\right)e$$

$$= \frac{E}{L}\left[\frac{P}{E}e + Q_u \frac{A_o}{E}e + Q_u \frac{B_o}{E}u\right]$$



Affine Parameterisation - Open Loop Unstable Model

Affine Parameterisation: Unstable Open Loop Case





Affine Parameterisation - Open Loop Unstable Model

The parametrisation as discussed above leads to the following parameterised version of the nominal sensitivities:

$$S_{o}(s) = \frac{A_{o}(s)L(s)}{E(s)F(s)} - Q_{u}(s)\frac{B_{o}(s)A_{o}(s)}{E(s)F(s)}$$
$$T_{o}(s) = \frac{B_{o}(s)P(s)}{E(s)F(s)} + Q_{u}(s)\frac{B_{o}(s)A_{o}(s)}{E(s)F(s)}$$
$$S_{io}(s) = \frac{B_{o}(s)L(s)}{E(s)F(s)} - Q_{u}(s)\frac{(B_{o}(s))^{2}}{E(s)F(s)}$$
$$S_{uo}(s) = \frac{A_{o}(s)P(s)}{E(s)F(s)} + Q_{u}(s)\frac{(A_{o}(s))^{2}}{E(s)F(s)}$$



What is Saturation?



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- **Objective:** How do we deal with the nonlinear affects of Saturation and Slew Rate Limitation? {u(t) and $\dot{u}(t)$ are limited}

Control System with Saturation on the Plant Input





Saturation

$$u = sat{\hat{u}} = \begin{cases} u_{min} & \text{if } \hat{u} < u_{min}, \\ \hat{u} & \text{if } u_{min} \leq \hat{u} \leq u_{max}, \\ u_{max} & \text{if } \hat{u} > u_{max}. \end{cases}$$





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- Ignoring the presence of saturations can cause long undesirable transients in the closed loop.
- The transients are due to the controller states having 'wound up' to large values.
- In a PID controller, there is only one state that is subject to wind-up the integrator state!
- Therefore in a PID controller, an anti-windup scheme involves limiting the integrator state in the some way.



Anti-windup scheme for a PI controller



When the controller is in the linear region of the saturation

$$\frac{\hat{u}}{e} = K_{p}(1 + \frac{K_{i}}{s}).$$

Thus the state of the controller is updated only with the actual plant input \hat{u} .



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- The key properties of an anti-windup scheme are:
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 - The states of the controller should have a stable realisation when driven by the actual plant input.



The Problem of Windup in IMC

We consider here open loop stable plants.



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- We consider here open loop stable plants.
- The block diagram shows an IMC structure where saturation is included on the input to the plant. The initial states of the system are depicted as x(t₀) and x(t₀). We also assume G_o(s) = G(s).

Internal Model Control with Saturation on Plant Input





When \hat{u} is in the linear region of the saturation,

 $y = Q(s)G_o(s)r + (1 - Q(s)G_o(s))d_o + lC(x - \hat{x})$

where IC is a transfer function relating the initial conditions to the output.



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Once the initial transient has decayed we see that the output is what we would expect, i.e. only a function of the reference and disturbance.



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 - **But** $x(t_0) \neq \hat{x}(t_0)$ since we had $\hat{u} \neq u$
- Hence we will again see a transient in the output.
- This transient is due to the mismatched states in the plant and the controller, i.e. $x(t_0) \neq \hat{x}(t_0)$, and is called windup.



Anti-windup for IMC

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Now $x(t_0) = \hat{x}(t_0)$ during and after saturation.



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 - 2. *e* and \hat{u} are completely independent of the saturation.
- For further insight into this problem, we digress slightly, and examine a feedback realisation of Q(s).


Assume, for simplicity, that $G_o(s)$ is minimum phase. Then

 $G_o^i = G_o^{-1}$ and $Q(s) = F_q(s)G_o^i(s)$.



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Lets also assume here that $F_q(s)$ is chosen to make Q(s) bi-proper

$$Q(s) = \frac{n_n s^n + \dots + n_1 s + n_0}{m_n s^n + \dots + m_1 s + m_0}$$

where $n_n \neq 0$ and $m_n \neq 0$.



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$$q_{\infty} = \lim_{s \to \infty} Q(s) = \frac{n_n}{m_n} \neq 0.$$

We can then write

$$Q(s) = q_{\infty} + \bar{Q}(s)$$

where $\bar{Q}(s)$ is strictly proper.

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Now

 $\hat{u} = Q(s)e$ then $e = Q(s)^{-1}\hat{u}$ $= \left(q_{\infty}^{-1} + \bar{Q}^{-1}(s)\right)\hat{u}$ $\therefore \hat{u} = \frac{1}{q_{\infty}^{-1}}\left(e - \bar{Q}^{-1}(s)\hat{u}\right)$



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Then,

$$\bar{Q}^{-1}(s) = Q^{-1}(s) - q_{\infty}^{-1}$$

then $\hat{u} = \frac{1}{q_{\infty}^{-1}} \left(e - \left(Q^{-1}(s) - q_{\infty}^{-1} \right) \hat{u} \right)$

which is a feedback representation of Q(s).



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Feedback Representation of *Q*(*s*)



Check:

$$\hat{u} = q_{\infty} \left(e - \left(Q^{-1}(s) - q_{\infty}^{-1} \right) \hat{u} \right)$$
$$\frac{\hat{u}}{e} = \frac{q_{\infty}}{1 + q_{\infty} Q^{-1}(s) - 1}$$
$$= \frac{1}{Q^{-1}(s)}$$
$$= Q(s)$$



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Now how can we use this to improve the anti-windup scheme for IMC.

Anti-windup scheme utilising the feedback representation of Q(s)





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Anti-windup scheme utilising the feedback representation of Q(s)



In this scheme the controller state is updated based on the controller action which is effectively applied to the linear plant.



Now how can we use this to improve the anti-windup scheme for IMC.

Anti-windup scheme utilising the feedback representation of Q(s)



- In this scheme the controller state is updated based on the controller action which is effectively applied to the linear plant.
- NOTE: For this to be feasible, we require Q(s) to be stable and bi-proper.



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- The scheme can be generalised (including unstable plants).
- Let C(s) be a bi-proper controller, then

$$C(s) = C_{\infty} + \bar{C}(s)$$

Anti-windup scheme for bi-proper C(s)





Slew Rate Limitation

Actuators may also be slew rate limited,

$$\dot{u}(t) = sat\{\dot{\hat{u}}(t)\} = \begin{cases} \sigma_{min} & \text{if } \dot{\hat{u}}(t) < \sigma_{min}, \\ \dot{\hat{u}}(t) & \text{if } \sigma_{min} \leq \dot{\hat{u}}(t) \leq \sigma_{max}, \\ \sigma_{max} & \text{if } \dot{\hat{u}}(t) > \sigma_{max}. \end{cases}$$



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We can approximate the derivative using Euler's Method

$$\dot{u}(t) \approx \frac{u(t) - u(t - \Delta)}{\Delta}$$



This leads to a Slew Rate Limiter (SRL)

Model of Slew Rate Limiter





The previous idea for the anti-windup scheme can be applied to windup due to slew rate limitations.

Anti-windup Scheme for Slew Rate Limitation





Saturation and Slew Rate Limitation

Saturation can be added to the slew rate limiter model.

Saturation and Slew Rate Limiter Model





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Saturation and Slew Rate Limiter Model



This model, which includes both saturation and slew rate limitation, can be incorporated in the same anti windup scheme as shown in the previous examples.

