#### **ELEC4410**

#### Control System Design

Lecture 4: Affine Parameterisation. PID Revisited, Time Delays and Undesirable Closed Loop Poles

School of Electrical Engineering and Computer Science
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▶ Revisit PID design using the affine parametrisation.

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- ▶ Control of time delayed plants using the affine parametrisation. (Also showing the connections with the Smith controller.)

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- ▶ Control of time delayed plants using the affine parametrisation. (Also showing the connections with the Smith controller.)
- Use of interpolation constraints to remove undesirable open loop poles.

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

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- We employ the affine synthesis methodology. Since there are no unstable zeros, the model is exactly invertible.
- We then choose

$$G_o^i(s) = (G_o(s))^{-1} = \frac{V_o s + 1}{K_o}$$

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▶ This implies that our final choice for Q(s) is of the form:

$$Q(s) = F_Q(s)G_o^i(s) = \frac{v_o s + 1}{K_o(\alpha s + 1)}$$

and the controller becomes

$$C(s) = \frac{Q(s)}{1 - Q(s)G_o(s)} = \frac{v_o s + 1}{K_o \alpha s} = \frac{v_o}{K_o \alpha} + \frac{1}{K_o \alpha s}$$

which is a PI controller.

With the PI controller parameters found above the nominal complementary sensitivity becomes

$$T_o(s) = Q(s)G_o(s) = F_Q(s) = \frac{1}{\alpha s + 1}$$

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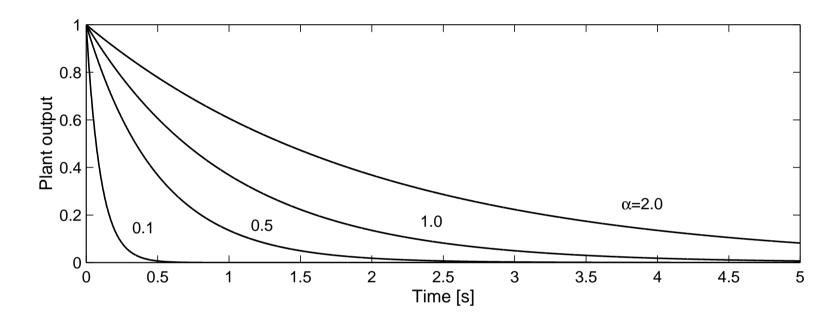
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- Choosing  $\alpha$  smaller makes the loop faster, whereas a larger value for  $\alpha$  slows the loop down.
- We thus see a direct connection between the design variable  $\alpha$  and the final closed loop performance.
- ▶ This is one of the principal advantages of the affine parametrisation methodology.

#### Effect of $\alpha$ on output disturbance rejection



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- Inserting a first order model into the affine structure automatically generates a PI controller.
- Inserting a second order model into the Q-structure automatically generates a PID controller.

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- Using this method, the control engineer works directly in terms of observable process properties (rise time, gain, etc.) and closed loop parameters providing an insightful basis for making trade-off decisions. The PI(D) parameters follow automatically.
- ▶ The approach does not preempt the design choice of cancelling or shifting the open-loop poles - both are possible and associated with different trade-offs.

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$$e^{-s\tau_o} \approx \frac{2 - s\tau_o}{2 + s\tau_o}$$

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- The Pade approximation usually works well if the time delay is smaller than the dominant time constant of the plant.
- ▶ For the case where the time delay is comparable or larger than the dominant time constant the Smith controller should be considered.

We consider here a special class of linear systems, namely those that be written as

$$G_o(s) = e^{-s\tau_o} \overline{G_o}(s)$$

where  $G_o(s)$  is a stable rational transfer function.

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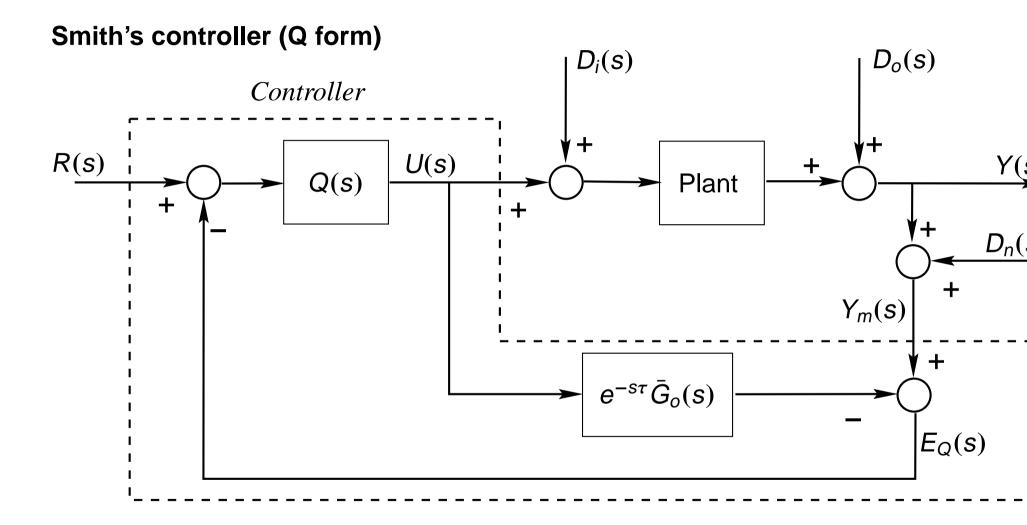
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- ▶ This idea was introduced by Otto Smith in the 1950's.

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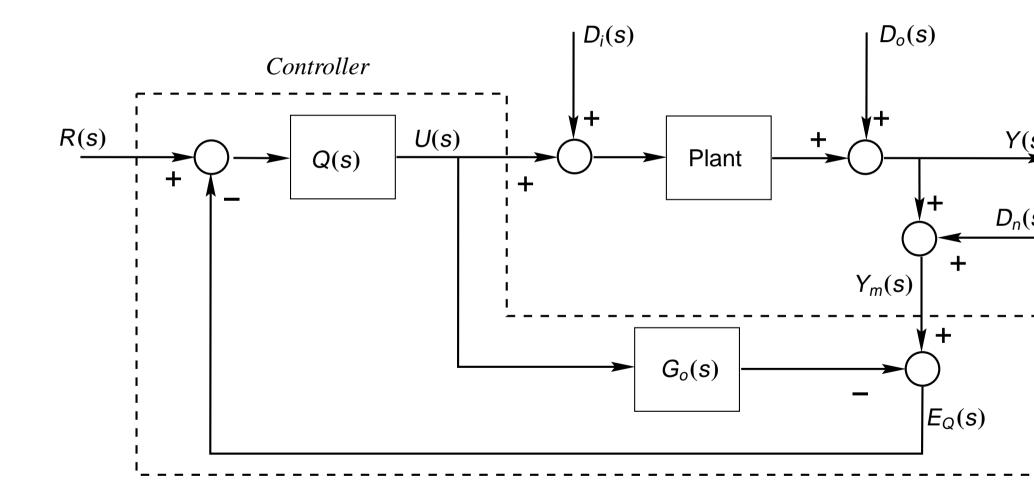
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  - the recognition that delay characteristics cannot be inverted.
- ▶ The structure of the traditional Smith controller can be obtained from the scheme shown on the next slide, which is a particular case of the general scheme for affine parameterisation given earlier.



Using earlier results we know the above configuration describes all stabilising controllers. All we need do is choose Q(s) to be a stable proper transfer function.

#### Youla's parametrisation of all stabilising controllers for stable plants



Design of Q(s) for systems with a time delay

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- Since the delay has no causal inverse, we seek an approximate inverse for  $\overline{G_o}(s)$ .
- ▶ This can be achieved directly. Alternatively, one can use the idea of feedback to generate a stable inverse. Thus we might conceive of evaluating *Q*(*s*) by

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

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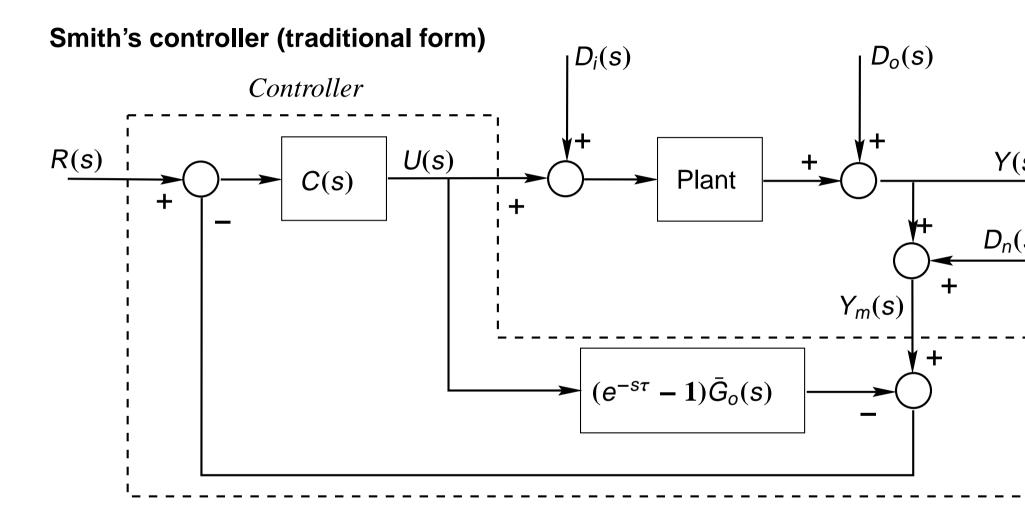
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If we use the above idea to choose Q(s); i.e. put

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

then we can redraw the controller as on the next slide.



In this form, we see that the design of C(s) can essentially be based on the nondelayed model. This is precisely the form of the traditional Smith controller.

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- ▶ This meant that all we needed to do was to choose Q(s) to ensure closed loop stability.
- We next examine a methodology, using affine parameterisation, to remove undesirable closed loop poles.

 $\blacktriangleright$  The idea of the Q(s) parametrisation remains valid since

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We also recall the following expressions for the sensitivity functions

$$T_o(s) = Q(s)G_o(s)$$

$$S_o(s) = 1 - Q(s)G_o(s)$$

$$S_{io}(s) = (1 - Q(s)G_o(s))G_o(s)$$

$$S_{uo}(s) = Q(s)$$

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- In practice we need to draw a distinction between stable poles and desirable poles.
- ▶ For example, a lightly damped resonant pair might well be stable but is probably undesirable.
- ▶ Say the open loop plant contains some undesirable (including unstable) poles. The only way to remove poles from the complementary sensitivity is to choose Q(s) to contain these poles as zeros.

▶ This results in cancellation of these poles from the product  $Q(s)G_o(s)$  and hence from  $S_o(s)$  and  $T_o(s)$ .

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- ▶ However, the cancelled poles may still appear as poles of the nominal input sensitivity  $S_{io}(s)$ , depending on the zeros of  $\mathbf{1} Q(s)G_o(s)$ , i.e. the zeros of  $S_o(s)$ .

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- ▶ To eliminate these poles from  $S_{io}(s)$  we need to also ensure that the offending poles are also zeros of  $(1 Q(s)G_o(s))$ .

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- The above statements represent a set of additional constraints on Q(s) to ensure closed loop stability.
- ▶ The result is summarised in the following Lemma:

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**Lemma.** (Interpolation constraints to avoid undesirable poles). Consider a nominal feedback control loop with one d.o.f. and assume  $G_o(s)$  contains undesirable (including unstable) open loop poles. We then have

1. Each of the sensitivity functions  $T_o(s)$ ,  $S_o(s)$ ,  $S_{io}(s)$  and  $S_{uo}(s)$  will have no undesirable poles if and only if: when the controller C(s) is expressed as:

$$Q(s) = \frac{C(s)}{1 + G_0(s)C(s)}.$$

Then Q(s) must satisfy the following (so called) Interpolation constraints:

- (a) Q(s) is proper, stable and has only desirable poles.
- (b) Any undesirable poles of  $G_o(s)$  are zeros of Q(s) with, at least, the same multiplicity as  $G_o(s)$ .
- (c) Any undesirable poles of  $G_o(s)$  are zeros of  $\mathbf{1} Q(s)G_o(s)$ , with at least the same multiplicity as  $G_o(s)$ .

Lemma. (cont.)

2. When conditions (b) and (c) are satisfied, then all resultant unstable pole-zero cancellations in C(s) should be performed analytically prior to implementation.

#### **PID Design Revisited**

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- In chemical processes, however, disturbances are frequently better modelled as occurring at the input to the system.
- We then recall that, the input disturbance response  $Y_d(s)$  is given by

$$Y_d(s) = S_{io}(s)D_i(s)$$

$$S_{io}(s) = S_o(s)G_o(s)$$

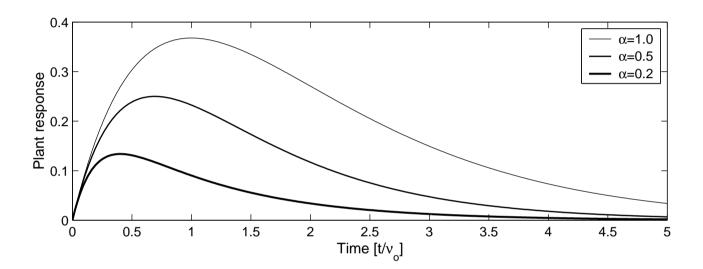
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- Thus the transient component in the input disturbance response will have a mode associated with that pole.
- ▶ The following slide shows the input disturbance response for the PI controller designed earlier via the affine parametrisation.

Input disturbance rejection with plant pole cancellation, for different values of  $\alpha$ .



Note that changing  $\alpha$  changes the magnitude of the response but the slow transient remains since this is dominated by the open loop plant as is evident from:

$$Y_d(s) = S_{io}(s)D_i(s)$$

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- As shown earlier, the only way to remove the pole from  $S_{io}(s)$  is to choose  $F_Q(s)$  in such a way that the offending pole is a zero of  $S_o(s) = 1 Q(s)G_o(s)$ , i.e. we require:

$$S_o(-a) = 0 \Longrightarrow T_o(-a) = F_Q(-a) = 1$$

where 
$$a \triangleq \frac{1}{v_o}$$

**Lemma.** Consider the plant model and Youla's parametrisation of all stabilising controllers for stable plants where  $Q(s) = (G_o(s))^{-1} F_Q(s)$ . Then a PI controller which does not cancel the plant pole, is obtained as

$$C(s) = K_P + \frac{K_I}{s}$$

where

$$K_P = \frac{2\psi_{cl}\omega_{cl}v_o - 1}{K_o}$$

$$K_l = \frac{v_o\omega_{cl}^2}{K_o}$$

and where  $\psi_{cl}$  and  $\omega_{cl}$  are chosen to obtain a closed loop characteristic polynomial given by:

$$A_{cl}(s) = \left(\frac{s}{\omega_{cl}}\right)^2 + 2\psi_{cl}\left(\frac{s}{\omega_{cl}}\right) + 1$$



▶ The proof of the above result is given in the book Control System Design (Goodwin et. al.).

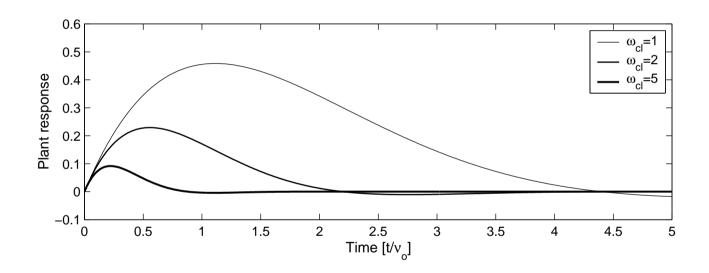
- ▶ The proof of the above result is given in the book Control System Design (Goodwin et. al.).
- It suffices to say that the key idea is to ensure that the *slow* open loop pole at  $\alpha = \frac{1}{v_o}$  is cancelled in the transfer function  $S_o(s) = 1 G_o(s)Q(s)$ ; i.e.

$$S_o(-a) = 0 \Longrightarrow T_o(-a) = F_O(-a) = 1$$

where 
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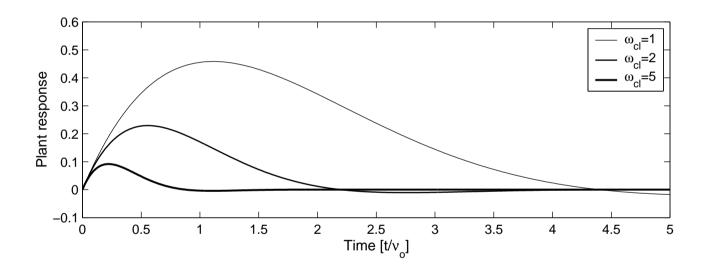
We repeat the simulation presented earlier where  $\alpha = \frac{1}{v_o}$  remained in the input disturbance rejection response.

Input disturbance rejection without plant pole cancellation.



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Input disturbance rejection without plant pole cancellation.



We see now that changing the design variable  $\alpha$  not only changes the size of the response but it also changes the nature of the transient.

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- We found that extra interpolation constraints on Q(s) were needed to eliminate undesirable poles from the input sensitivity  $S_{io}(s)$ .
- In the design examples presented to date we have chosen Q(s) to explicitly account for these interpolation constraints.
- ▶ However, this is a tedious task and one is lead to ask the following question: Can we reparameterise C(s) in such a way that the interpolation constraints given in the Lemma are automatically satisfied?

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- In that sense, delays are related to NMP plant zeros, which cannot be stably inverted either.
- A delay of magnitude T, causes similar trade-offs as an unstable zero at s = T/2.
- An early controller conceived to deal with the non-invertibility of delays is the famous Smith-predictor.
- The trade-offs made in the Smith-predictor can be nicely analysed in the affine structure. Indeed, the structures are very similar. Caution should be exercised, however, not to confuse the generic controller representation of the affine parametrisation with the particular synthesis technique of the Smith-predictor.