Controller Implementation

Recall we have shown

$$U(s) = H(s)E(s) = \left[C_o + \sum \frac{C_i}{s - a_i}\right]E(s)$$

where a_i are poles of H(s), and $c_0, c_0, c_2, ...$ are PFE coefs, with $c_0 = \phi$ if $\rho(H) > \phi$ (H(s) has more poles than zeros).

Let
$$X_i(s) = (s-a_i) E(s)$$

then
$$u(t) = c_0 e(t) + \sum c_i x_i(t)$$

where e(+)= Ya(+)-Y(+), and Xi(+) are solins of

$$\dot{x}_i(t) = a_i x_i(t) + c(t)$$

The discrete time stepping under which the computer and Sensor/actuator electronics operate mean that U/1) will be computed only at integer multiples of the sample interval, Ts.

Let
$$u_K = u(t_K) = u(K_{\overline{b}})$$
, and $e_K = e(t_K) = e(K_{\overline{b}})$

From above:

$$U_K = C_0 e_K + \sum C_i \chi_i(t_K)$$

We need to Know how to audicate $x_i(t_K)$, i.e. sol'n

$$\dot{x}_i(t) = a_i x_i(t) + e(t)$$
 evaluated at $t = t_K$

focus on just a single one of these equis, since they are identical except for coets a_i :

$$\dot{x}(t) = \alpha x(t) + \varepsilon(t)$$
 (Let $\varepsilon(t) = \varepsilon(t)$ here, to avoid confusion with $\varepsilon^{\alpha t}$)

Assume E(+) is a step of size Eo, and x(0) = xo

Then
$$X(s) = \frac{x_0}{5-a} + \frac{\xi_0}{5(5-a)}$$

Note: e(+) will not generally be a step, even if ye(+) is!

But, the above is a useful intermediate result, as we will see Next.

Here

$$X(t) = e^{\alpha t} \chi_0 + \left(\frac{-1}{\alpha}\right) (1 - e^{\alpha t}) \varepsilon_0$$

Let
$$\alpha = e^{\alpha T}$$
, $\beta = (1-\alpha)/(-\alpha)$

Let
$$X(T_s) = X_t = X_t + \beta E_0 = X(t_t)$$

Now, how does this help generally?

Sampling of output at discrete times t_K=KT, means that error e(t) will have a staircase graph



i.e. e(t) will be constant with level E_K on the interval $t_K \leqslant t \leqslant t_{K+1}$.

Note that at to, e(+) does look like a step.

So, it is true for first sample interval that

$$X(t_1) = x_1 = \alpha X_0 + \beta E_0$$
 (as above)

But what about subsequent time steps??

Exploit time invariance: when solving constant coeff DE's, the time called zero is arbitrary. All that matters is the initial cond'n and the time elopsed since initial time.

So, to get sol'n for wext sample time to, we can "reset"

The zero time to t,, and use X(t,) as initial cond'n.

New

Then, from new zero time t=ti, error c(t) looks like a step of height

(initially)

 $X(t_2) = X_2 = \propto X_1 + \beta E_1$ by same logic as above

We can repeat this trick for all subsequent tk:

$$X_{K+1} = \alpha X_K + \beta E_K$$

where
$$x_{K} = x(t_{K}) = x(KT_{S})$$

$$\alpha = e^{\alpha T_s}$$
, $\beta = (1-\alpha)/(-\alpha)$

We have thus shown that:

$$X(t_{\kappa+1}) = \propto \times (t_{\kappa}) + \beta e(t_{\kappa})$$

is an iterative algorithm for generating the exact Sol'n for X(t) at each of the sample times t_K , given the staircase structure of e(t).

So generally:

$$u(t_k) = u_k = c_e(t_k) + \sum_i c_i x_i(t_k)$$

Where <u>each</u> xi(tx) is computed iteratively using

$$X_i(t_{\kappa+1}) = \prec_i X_i(t_{\kappa}) + \beta_i e(t_{\kappa})$$

where
$$\alpha_i = \exp[\alpha_i T_i]$$
, $\beta_i = \left[\frac{1-\alpha_i}{(-\alpha_i)}\right]$

and Ts is the sample interval.

Real-time implementation

$$X_i(t_{\kappa+1}) = \prec_i X_i(t_{\kappa}) + \beta_i e(t_{\kappa})$$

and
$$\Rightarrow \alpha_i = \exp[\alpha_i T_s]$$
, $\beta_i = \left[\frac{1-\alpha_i}{(-\alpha_i)}\right]$

Matlab code

```
function u=control(yd,y)

% define K (number!)
K=...
% compute u
```

```
% compute u
e = yd-y;
u = K*e;
```

H(s) with I pole

```
function u=control(yd,y)

% define c0, c1, alpha, beta (as numbers!)

c0=...

c1=...

alpha = exp P_*T here, and

alpha=...

beta=...

Beta = (1-alpha)/(-P_*)

% compute u

e = yd-y;

u = c0*e+c1*x;
```

% update x x = alpha*x+beta*e;

H(s) with I pole

function u=control(yd,y)

```
% define c0, c1, alpha, beta c0=...
c1=...
alpha=...
beta=...
```

% compute u
e = yd-y;
u = c0*e+c1*x;

% update x x = alpha*x+beta*e;

end

Unfortunately, won't work as written!

We need the function to "remember" the values of x between calls.

Remember: Matlab functions (like C/C++ functions) have their own, private workspace (storage) for their variables, which is separate from the main workspace (main function).

Local variables in functions are cleared after the function runs.

Can prevent this clearing by declaring the variable to be "persistent" in Motlab ("Static" in C/C++).

function u=control(yd,y)

```
persistent x
% define c0, c1, alpha, beta
c0 = ...
c1 = ...
alpha=...
beta=...
% compute u
e = yd-y;
u = c0*e+c1*x;
% update x
x = alpha*x+beta*e;
```

H(s) = K (s-ze) = Co + (s-pe)

```
function u=control(yd,y)
persistent x
% define c0, c1, alpha, beta
c0 = ...
c1 = ...
alpha=...
beta=...
% compute u
e = yd-y;
u = c0*e+c1*x;
% update x
x = alpha*x+beta*e;
                           Still won't work!
end
```

X needs to be initialized, but only the 1st time the function is Called.

Matlub initializes a persistent variable as an empty array the first time the function is run

We can test for this, and set initial value of x to our pleasing: "Isempty" function for test

Note: Simplest to initialize x to zero, unless there is a compelling reason not to (very rare)

function u=control(yd,y)

persistent x

```
if isempty(x) initialize × x=0; end
```

% define c0, c1, alpha, beta c0=... c1=... alpha=... beta=...



```
% compute u
e = yd-y;
u = c0*e+c1*x;
```

% update x x = alpha*x+beta*e; All our mathematical analysis
Ultimately boils down to 4 "magic
numbers" that we plug into
this standard template.