

Themes:

Linear equations, least squares, non-linear equations; Error estimates, polynomials
 Sensitivity analysis:

EXS 3.8

Compute sensitivity of (a and) b with respect to k_1 , k_2 , a_{tot} and b_{tot}

Eliminate a and form the polynomial equation. Solve by roots, compute sensitivities db/dk_1 , db/dk_2 , $db/datot$ and $db/dbtot$, by analytic differentiation of roots wrt coefficients and by numerical differencing.

```
% EXS 3.8
clear all
format compact
k1 = 0.12; k2 = 0.22;
atot = 0.60; btot = 1.0;
% a = atot/(1+k1b+k2b^2) = (btot-b)/(k1 b + 2k2 b^2)
% atot(k1 b + 2k2 b^2)-(btot-b)(1+k1b+k2b^2) = 0
% - btot +
% b(atot k1 +1 -btot k1) +
% b^2 (2 k2 atot -btot k2 + k1) +
% b^3 k2
c = [k2,(2*k2*atot-btot*k2+k1),(atot*k1+1-btot*k1),-btot ];
z = roots(c);
i = find(imag(z)==0); % there is only one real root
x = real(z(i))
x0 = x;
%
% Analytical differentiation
%=====
% dp/dz* dz/dk + dp/dk = 0
% dz/dk = -dp/dk / dp/dz;
dc = polyder(c);
dpdz = polyval(dc,x);
%
dcdk1 = [0, 1 , atot-btot, 0 ];
dcdk2 = [1, 2*atot-btot, 0 , 0 ];
dcdatot = [0, 2*k2 , k1 , 0 ];
dcdbtot = [0, -k2 , -k1 , -1 ];
%
% dp/dk:
dpdk1 = polyval(dcdk1 ,x);
dpdk2 = polyval(dcdk2 ,x);
dpdatot = polyval(dcdatot,x);
dpdbtot = polyval(dcdbtot,x);
%
% dz/dk:
dzdk1 = -dpdk1 /dpdz
dzdk2 = -dpdk2 /dpdz
dzdatot = -dpdatot/dpdz
dzdbtot = -dpdbtot/dpdz
%
% Experimental evaluation; change parameters by h
%=====
h = 1e-4; % how large?
par0 = [k1 k2 atot btot];
npar = length(par0);
dxdp = zeros(4,1);
for k = 1:npar
  par = par0;
```

```

par(k) = par(k)+h;
k1    = par(1);
k2    = par(2);
atot = par(3);
btot = par(4);
c = [k2,(2*k2*atot-btot*k2+k1),(atot*k1+1-btot*k1),-btot ];
z = roots(c);
i = find(imag(z)==0);
x = real(z(i));
dxdp(k) = (x-x0)/h;
end
dxdp

```

EXS 4.2

EXS 4.6 – note that normal equations become diagonal. mention FFT för periodic data.

EXS 4.13

Different linear formulations:

1. $\ln x_j - \ln x_0 + k t_j = 0$ (the obvious)

Compute residual vector r and estimate measurement error variance squared as
 $r^T r / \sqrt{m-n}$ $n = \#$ measurements (5), $m = \#$ parameters (2)

The diff. eqn implies $x(t_{k+1})/x(t_k) = \exp(-k(t_{k+1} - t_k))$ so

Note equidistant table $t_{k+1} - t_k = h$

2. $\ln x_{k+1} - \ln x_k = -k h$

Solve normal equations exactly (one unknown!):

$$n k_{\text{hat}} = -1/h \sum (\ln x_{k+1} - \ln x_k) = 1/h \ln (x_N/x_0)$$

Is this a good idea (only x_0 and x_4 are used)

To use more data,

3. $\ln x_4 - \ln x_2 = -k 2 h$,

$$\ln x_3 - \ln x_1 = -k 2 h$$

Solve normal eq's exactly again.

$$4 k h = \ln ((x_4 x_3) / (x_2 x_1))$$

Difference?

Now use differential equation $dx/dt = -kx$ and use difference approx.

4. $(x_{k+1} - x_{k-1})/(2h) + k x_k = 0 \quad k = 1,2,3$

$$2h k = \sum((x_{k+1} - x_{k-1}) * x_k) / \sum(x_k^2)$$

..exact solution ...discuss error sources

Interpolation by polynomials

EXS 5.3 – note linearity of polynomial wrt data.

- mention matlab `polyfit`, `polyval`, `polyder`
- show matlab code for evaluating the Newton form of polynomial

like Horner's scheme.

- Show also how to compute derivative from the recursion.
- Challenge – compute coeffs in “standard” polynomial from (x, c) for Newton

Exempel (finns inte i PP) (sorry, in swedish)

I Lab 3 får ni programmera polynom-interpolation med Newtons ansats ovan. Men interpolationen kan också göras med polyfit, dock med polynomet representerat på annat sätt. Man kan kontrollera sitt program mot polyfit genom att konvertera Newton-koefficienterna c_j till monom-koefficienterna a_j

$$p_n(x) = \sum_{i=1}^{n+1} a_i x^{n+1-i} = \sum_{i=0}^n c_i \prod_{k=1}^i (x - x_k)$$

Antag att vi representerar polynom med koefficienterna $\{c_i\}$, $i = 0, \dots, n$ och $\{x_i\}$, $i = 1, \dots, n+1$. Ge en algoritm för beräkning av $p_n(x)$ som använder n multiplikationer, och en algoritm för beräkning av derivatan $d p_n / dx$.

Lösning

Om man programmerar formeln ovan som den står går det åt ungefär $n^2/2$ multiplikationer. Men produkten $c_k(x-x_1)(x-x_2)\dots(x-x_k)$ kan ju skrivas $c_k t_k$ där t_k räknas ut rekursivt i slingan som adderar termerna, t ex så här (obs. vi lagrar c_k i $c(k+1)$). Interpolationspunkterna ligger i X .

```
t = ones(size(x)); p = 0;
for k = 1:n+1
    p = p + c(k)*t;
    t = t.*(x-x(k));
end
```

Programmet refererar till $X(n+1)$ (men den används aldrig) så därför låter vi $X(n+1)$ ingå i datastrukturen.

Man kan spara ytterligare en multiplikation per varv genom att evaluera högstgradstermen först,

$$\begin{aligned} p_n(x) &= p_0, \text{ där} \\ p_n &= c_n \\ p_{k-1} &= c_{k-1} + (x-x_k) p_k, \quad k = n, n-1, \dots, 1 \end{aligned}$$

Derivatan kan beräknas samtidigt genom att man ”deriverar” rekursionsformeln. Inför $p'_n(x) = q_0$, där q_k är derivatan av mellanresultaten p_k , $k = n, n-1, \dots$

$$\begin{aligned} p_n &= c_n, \quad q_n = 0 \\ p_{k-1} &= c_{k-1} + (x-x_k) p_k, \\ q_{k-1} &= p_k + (x-x_k) q_k, \quad k = n, n-1, \dots, 1 \end{aligned}$$