

Exercise A8.13

$A$  is  $(m \times n)$

$b$  is  $m$ -vector  $(m, 1)$

$A$  has linearly independent columns

$$\hat{x}$$
 is the solution to minimize  $\|Ax - b\|^2 = \sum_{i=1}^m (a_i^T x - b_i)^2$

$a_i^T$  is the  $i$ th row of  $A$

$$\hat{y}_k$$
 is the solution to minimize  $\sum_{i \neq k} (a_i^T y - b_i)^2 =$

$$\text{row } k \text{ is removed} \quad = \sum_{i=1}^{k-1} (a_i^T y - b_i)^2 + \sum_{i=k+1}^m (a_i^T y - b_i)^2$$

\* Assume for each  $k=1, \dots, m$ , the matrix formed by removing row  $k$  from  $A$  has linearly independent columns. So solution  $\hat{y}_k$  is unique

a)

$$\text{minimize} \quad \sum_{i=1}^{k-1} (a_i^T y - b_i)^2 + (a_k^T y + z - b_k)^2 + \sum_{i=k+1}^m (a_i^T y - b_i)^2$$

the segment in the middle is our  $k$ th row, we want to remove that so we find the condition that makes this zero.

$$(a_k^T y + z - b_k)^2 = 0 \Rightarrow a_k^T y + z - b_k = 0 \Rightarrow z \text{ is our outlier}$$

$\therefore z = b_k - a_k^T y$ , which agrees with the problem statement

We're left with

$$\text{minimize} \quad \sum_{i=1}^{k-1} (a_i^T y - b_i)^2 + \sum_{i=k+1}^m (a_i^T y - b_i)^2$$

We know because  $A$  is linearly indep columns,  $\hat{y}$  is a unique solution to equations of that form, which is what we have

b)  $\|(A_{e_k})(y z)^T - b\|^2$

$$\therefore \text{normal equation } (A^T)(A_{e_k})(y z)^T = (A^T)b \Rightarrow (A^T A_{e_k})(y z)^T = (A^T b)$$

$$A^T A y + a_k^T z = A^T b \Rightarrow y = (A^T A)^{-1}(A^T b - a_k^T z) = \hat{x} - (A^T A)^{-1} a_k^T z$$

$$a_k^T y + z = b_k \rightarrow a_k^T (\hat{x} - (A^T A)^{-1} a_k^T z) + z = b \rightarrow z = \frac{b_k - a_k^T \hat{x}}{1 - a_k^T (A^T A)^{-1} a_k}$$

$$\hat{y}_k = \hat{x} - \frac{b_k - a_k^T \hat{x}}{1 - a_k^T (A^T A)^{-1} a_k} (A^T A)^{-1} a_k$$

c)  $A = QR$ ,  $\hat{y}_k = \hat{x} - \frac{b_k - a_k^T \hat{x}}{1 - a_k^T R^{-1} R^{-T}} R^{-1} R^{-T} a_k$

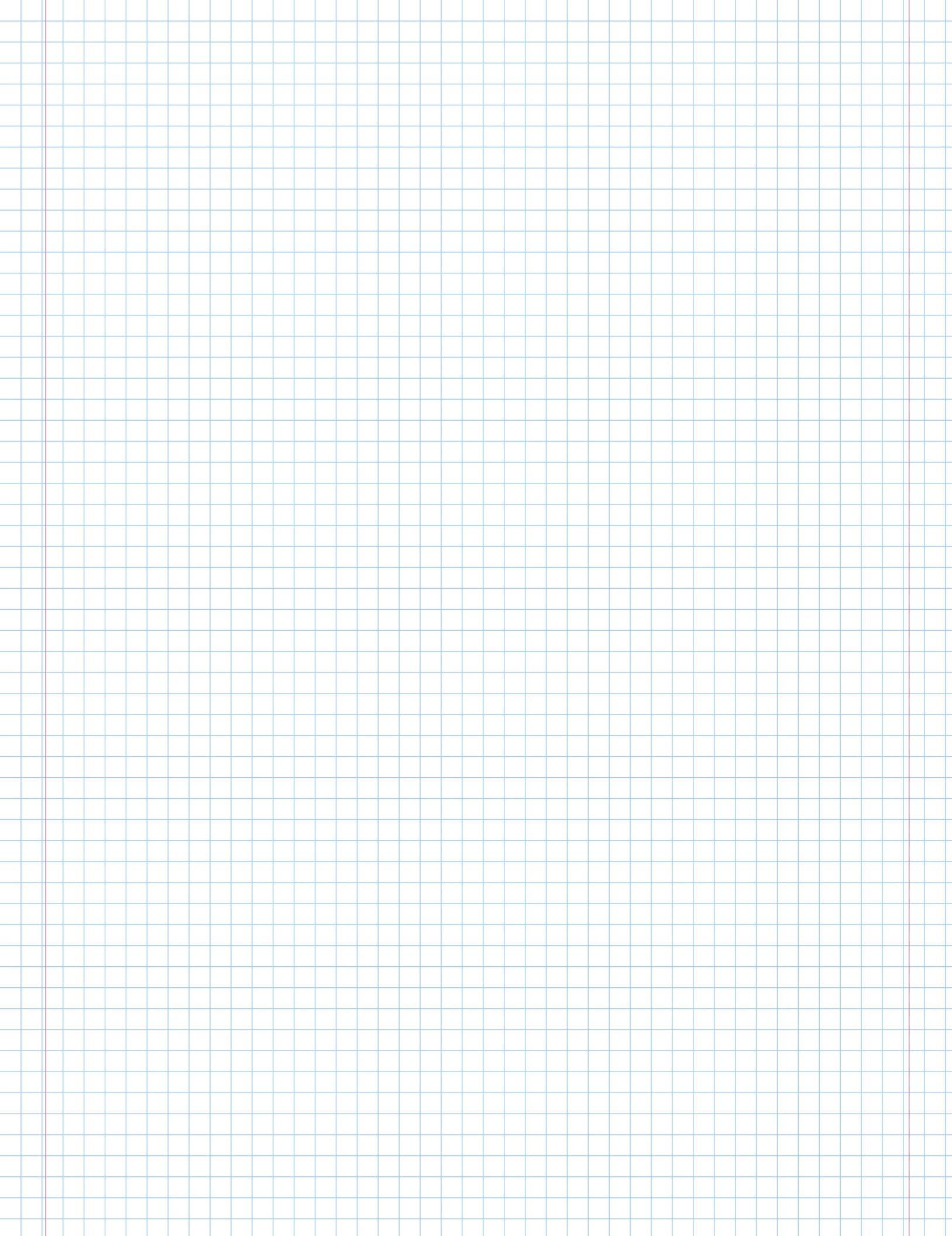
$$A = QR - 2mn^2 \text{ flops}$$

$$a^T b - n^2 \text{ flops}$$

$$\frac{b_k - a_k^T \hat{x}}{1 - a_k^T a_k} - 4mn \text{ flops}$$

$$R^{-1} u_k - mn^2 \text{ flops}$$

$$\frac{\hat{x} - b_k - a_k^T \hat{x}}{1 - a_k^T a_k} n^{-1} u_k - 2mn \text{ flops} \quad \therefore 3mn^2 \text{ flops total}$$



### Problem A9.4

a)

$$\text{minimize } \|Ax - y\|^2 + \lambda(\|D_v x\|^2 + \|D_h x\|^2)$$

$$\hookrightarrow \text{normal equation: } (A^T A + \lambda D_v^T D_v + \lambda D_h^T D_h) x = A^T y$$

$$\text{Assume } A = T(B), D_v = T(E), D_h = T(E^T)$$

$$T(U) = \frac{1}{n^2} \tilde{w}^H \text{diag}(\tilde{w}_u) \tilde{w}$$

Substitute into normal equations

$$\begin{aligned} T(U)^T T(U) &= \left( \frac{1}{n^2} \tilde{w}^H \text{diag}(\tilde{w}_u) \tilde{w} \right)^H \left( \frac{1}{n^2} \tilde{w}^H \text{diag}(\tilde{w}_u) \tilde{w} \right) \\ &= \frac{1}{n^2} \tilde{w}^H \text{diag}(\tilde{w}_u)^H \tilde{w} \cdot \frac{1}{n^2} \tilde{w}^H \text{diag}(\tilde{w}_u) \tilde{w} \\ &= \frac{1}{n^2} \tilde{w}^H \text{diag}(\tilde{w}_u)^H \underbrace{\tilde{w} \tilde{w}^H}_{I_{n^2}} \text{diag}(\tilde{w}_u) \tilde{w} \\ &= \frac{1}{n^2} \tilde{w}^H \text{diag}(\tilde{w}_u)^H \text{diag}(\tilde{w}_u) \tilde{w} \end{aligned}$$

$$\therefore (A^T A + \lambda D_v^T D_v + \lambda D_h^T D_h) x = A^T y$$

$$\hookrightarrow \frac{1}{n^2} \tilde{w}^H \text{diag}(\tilde{w}_b)^H \text{diag}(\tilde{w}_b) \tilde{w} + \frac{1}{n^2} \tilde{w}^H \lambda \text{diag}(\tilde{w}_e)^H \text{diag}(\tilde{w}_e) \tilde{w} + \frac{1}{n^2} \tilde{w}^H \lambda \text{diag}(\tilde{w}_e)^H \text{diag}(\tilde{w}_e) \tilde{w} \\ = \frac{1}{n^2} \tilde{w}^H \text{diag}(\tilde{w}_b)^H \tilde{w} y$$

$$\hookrightarrow \frac{1}{n^2} \tilde{w}^H [\text{diag}(\tilde{w}_b)^H \text{diag}(\tilde{w}_b) + \lambda \text{diag}(\tilde{w}_e)^H \text{diag}(\tilde{w}_e) + \lambda \text{diag}(\tilde{w}_e)^H \text{diag}(\tilde{w}_e)] \tilde{w} = \frac{1}{n^2} \tilde{w}^H \text{diag}(\tilde{w}_b)^H \tilde{w} y$$

$$\hookrightarrow D = \text{diag}(\tilde{w}_b)^H \text{diag}(\tilde{w}_b) + \lambda \text{diag}(\tilde{w}_e)^H \text{diag}(\tilde{w}_e) + \lambda \text{diag}(\tilde{w}_e)^H \text{diag}(\tilde{w}_e)$$

$$\hookrightarrow \frac{1}{n^2} \tilde{w}^H D \tilde{w} x = \frac{1}{n^2} \tilde{w}^H \text{diag}(\tilde{w}_b)^H \tilde{w} y \rightarrow D \tilde{w} x = (\tilde{w}^H)^{-1} (\tilde{w}^H) \text{diag}(\tilde{w}_b)^H \tilde{w} y$$

$$x = \tilde{w}^{-1} D^{-1} \text{diag}(\tilde{w}_b)^H \tilde{w} y$$

%Homework 6, Problem 2 (A9.4)

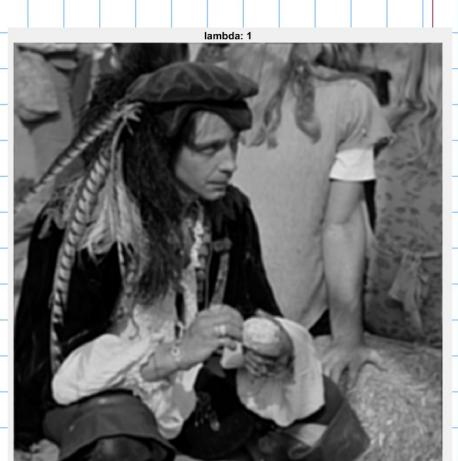
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1 clear all;
2 load('deblur.mat');
3
4 lambda = 10^(-5);
5 [m,n] = size(Y);
6 E = zeros(m,m);
7 E(1,1) = 1;
8 E(m,1) = -1;
9 Afourier = fft2(B);
10 dQ = abs(fft2(E)).^2 + abs(fft2(E')).^2;
11 dW = abs(Afourier).^2 + lambda * dQ ;
12 dE = conj(Afourier).* fft2(Y) ./ dW;
13 X = real(ifft2(dE));
14 imshow(X);
15 title(['lambda: ' num2str(lambda)]);
16
17
18

```



from the samples to the right,  $\lambda > 1$  made the image worse, between  $10^{-4} < \lambda < 10^{-1}$  were good clarity



### Problem 10.1

$$\begin{bmatrix} S_1(t+1) \\ S_2(t+1) \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ 0 & 0.95 \end{bmatrix} \begin{bmatrix} S_1(t) \\ S_2(t) \end{bmatrix} + \begin{bmatrix} 0 \\ 0.1 \end{bmatrix} u(t) \quad t = 0, 1, 2, \dots$$

$$S(t+1) = A S(t) + B u(t)$$

$$t=0 \rightarrow S(1) = A S(0) + B u(0) \quad S(1) = B u(0)$$

$$t=1 \rightarrow S(2) = A S(1) + B u(1) \quad S(2) = A B u(0) + B u(1)$$

$$t=2 \rightarrow S(3) = A S(2) + B u(2) \quad S(3) = A^2 B u(0) + A B u(1) + B u(2)$$

$$t=3 \rightarrow S(4) = A S(3) + B u(3) \quad S(4) = A^3 B u(0) + A^2 B u(1) + A B u(2) + B u(3)$$

$$t=4 \rightarrow S(5) = A S(4) + B u(4) \quad S(5) = A^4 B u(0) + A^3 B u(1) + A^2 B u(2) + A B u(3) + B u(4)$$

$$\vdots$$

$$t=N-1 \rightarrow S(N) = A^{N-1} B u(0) + A^{N-2} B u(1) + \dots + A^1 B u(N-2) + B u(N-1)$$

$$\therefore S(N) = \begin{bmatrix} A^{N-1} B & A^{N-2} \dots AB & B \end{bmatrix} \begin{bmatrix} u(0) \\ u(1) \\ \vdots \\ u(N-2) \\ u(N-1) \end{bmatrix}$$

a) minimize  $\|x\|^2$        $S_1(N)=10$   
 Subject to  $Cx=d$        $S_2(N)=0$

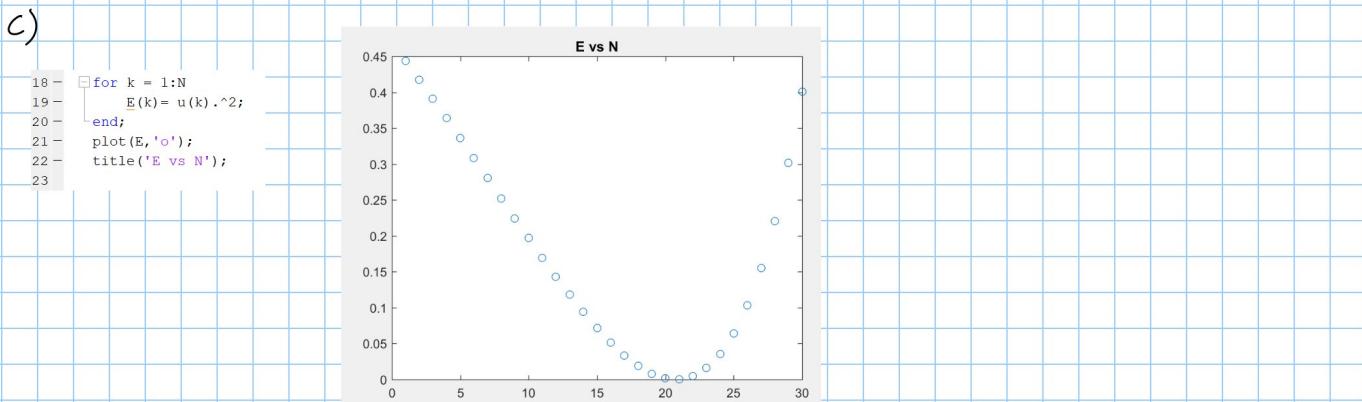
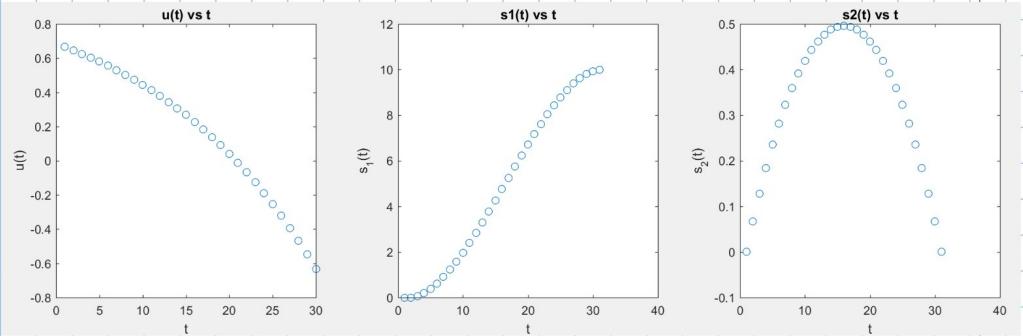
$$E = \sum_{t=0}^{N-1} u(t)^2 \quad C = A, \quad d = S(N) = \begin{bmatrix} S_1(N) = 10 \\ S_2(N) = 0 \end{bmatrix}$$

$$= \text{minimize } \|x\|^2 \quad \therefore x = u(t) = \begin{bmatrix} u(0) \\ u(1) \\ \vdots \\ u(N-2) \\ u(N-1) \end{bmatrix}$$

```

1 % Homework 6, Problem 3
2 clear all;
3
4 N = 30;
5 A = [1 1; 0 0.95];
6 B = [0; 0.1];
7 C = zeros(2,N);
8 C(:,N) = B;
9 for t = N-1:-1:1
10 C(:,t)=A*C(:,t+1);
11 end;
12 d=[10;0];
13 u = C\*(C'*C)\ d;
14
15 s = zeros(2,N+1); s(1,1)
16 for t=1:N
17 s(:,t+1) = A*s(:,t)+B*u(t);
18 end;
19 subplot(1,3,1);
20 plot(u, 'o'); title('u(t) vs t');
21 ylabel('u(t)'); xlabel('t');
22 subplot(1,3,2);
23 plot(s(1,:), 'o'); title('s1(t) vs t');
24 ylabel('s_1(t)'); xlabel('t');
25 subplot(1,3,3);
26 plot(s(2,:), 'o'); title('s2(t) vs t');
27 ylabel('s_2(t)'); xlabel('t');

```



### Problem A10.11

$$\text{minimize} \quad \sum_{i=1}^{50} (p(t_i) - y_i)^2 + \sum_{i=51}^{100} (q(t_i) - y_i)^2 \quad p(o) = q(o)$$

$$p'(o) = q'(o)$$

$$p(x) = \theta_1 + \theta_2 x + \theta_3 x^2 + \theta_4 x^3 \quad q(x) = \theta_5 + \theta_6 x + \theta_7 x^2 + \theta_8 x^3$$

$$x = \theta_1, \dots, \theta_8$$

$$A = \begin{bmatrix} 1 & t_1 & t_1^2 & t_1^3 \\ \vdots & \ddots & & \\ 1 & t_m & t_m^2 & t_m^3 \\ & \ddots & 1 & t_{m+1} & t_{m+1}^2 & t_{m+1}^3 \\ 0 & \vdots & & 1 & t_{2m} & t_{2m}^2 & t_{2m}^3 \end{bmatrix} \quad b = \begin{bmatrix} y_1 \\ \vdots \\ y_m \\ y_{m+1} \\ \vdots \\ y_{2m} \end{bmatrix} \quad C = \begin{bmatrix} 1 & \hat{t} & \hat{t}^2 & \hat{t}^3 & -1 & -\hat{t} & -\hat{t}^2 & -\hat{t}^3 \\ 0 & 1 & 2\hat{t} & 3\hat{t}^2 & 0 & -1 & 2\hat{t} & -3\hat{t} \end{bmatrix} \quad d = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

```

1 %Homework 6, Problem 4
2
3 clear all;
4 splinefit;
5 m = 50;
6 n = 4;
7 y1 = y(1:m);
8 y2 = y(m+1:2*m);
9 t1 = t(1:m);
10 t2 = t(m+1:2*m);
11 A1 = fliplr(vander(t1));
12 A1 = A1(:, 1:n);
13 A2 = fliplr(vander(t2));
14 A2 = A2(:, 1:n);
15
16 A = [ A1, zeros(m,n); zeros(m,n) , A2];
17 b = [y1; y2];
18 c = [1, zeros(1, n-1), -1, zeros(1, n-1);
19 0, 1, zeros(1, n-2), 0, -1, zeros(1, n-2)];
20 d = zeros(2,1);
21 x = [A'*A, c'; c, zeros(2,2)] \ [A'*b; d];
22 pyC1 = x(1:n);
23 pyC2 = x(n+1:2*n);
24
25 disp(pyC1);disp(pyC2);
26

```

```
>> A10_11
0.2841
-0.9025
-1.2503
-0.1317
}
```

} PyC1

```
0.2841
-0.9025
-1.8160
2.5004
```

} PyC2