

3 OCTAVE/MATLAB Problem 1.3

Least-Squares Tracking of a Time-Varying System

3.1 Task a)

For the first task a function called **ls_filter()** shall be implemented which computes the *least-squares* optimum filter coefficients $\mathbf{c}_{LS}[n] = \operatorname{argmin}_{\mathbf{c}} J_{LS}(\mathbf{c}, n)$ according to the cost function

$$J_{LS}(\mathbf{c}, n) = \sum_{k=0}^n |e[k]|^2 \quad (1)$$

We know from the problem class:

$$\mathbf{c}_{LS} = (\mathbf{X}^T \cdot \mathbf{X})^{-1} \cdot \mathbf{X}^T \cdot \mathbf{d} \quad (2)$$

The Matlab files can be found in the **.zip** file. The code is also appended to this PDF sheet (Section 3.5)

3.2 Task b)

For the second task the 'unknown' system filter coefficients shall be plotted. These are:

$$\mathbf{h}[n] = \begin{bmatrix} -1 \\ 2 - 0.97^n \\ 0.3 \cdot \cos(\theta n) \end{bmatrix} \quad (3)$$

where $\theta = \frac{3\pi}{1000}$ for $n \in [0, 999]$

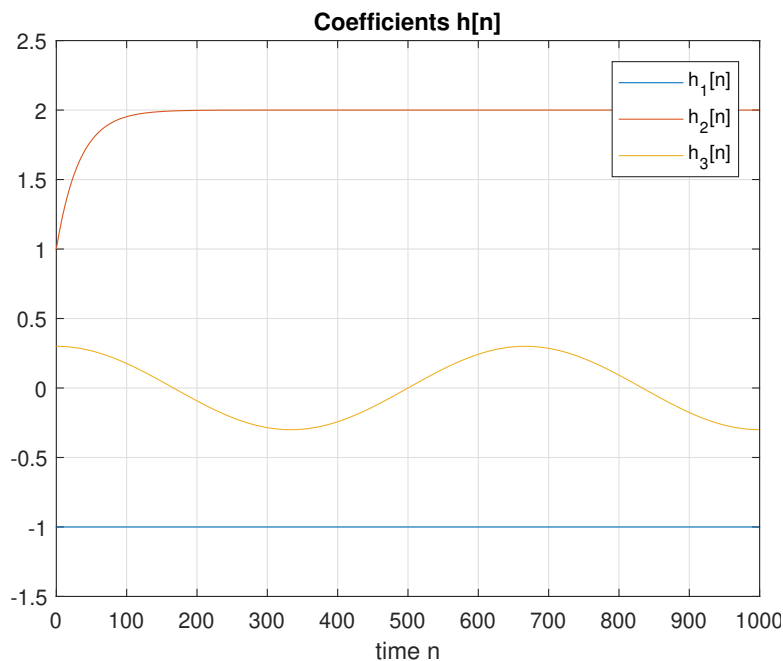


Figure 1: Coefficients for $\mathbf{h}[n]$

3.3 Task c)

In this task the coefficients were calculated using the *least-square* filter programmed in task a). The input signal $x[n]$ is a stationary white noise process with variance $\sigma_x^2 = 1$ (White noise processes always have a zero mean value). To compute the filter coefficients $\mathbf{c}[n]$ the error $e[n]$ has to be computed first. Therefore the values of $d[n]$ are needed and can be computed from the 'unknown' system $\mathbf{h}[n]$ (that is why we know the coefficients of $\mathbf{h}[n]$. In real life we would simply measure the data of $d[n]$). The noise $w[n] = 0$ this time and will be considered in task d).

The adaptive filter should have $N = 3$ coefficients (the filter order i.e. the amount of delay elements is $N - 1 = 2$).

$$\mathbf{c}[n] = \begin{bmatrix} c_1[n] \\ c_2[n] \\ c_3[n] \end{bmatrix} \quad (4)$$

The filter coefficients were computed by segmenting the input signal $x[n]$ and calling the function `ls_filter()`. This should be done for segment lengths $M = \{20, 50\}$.

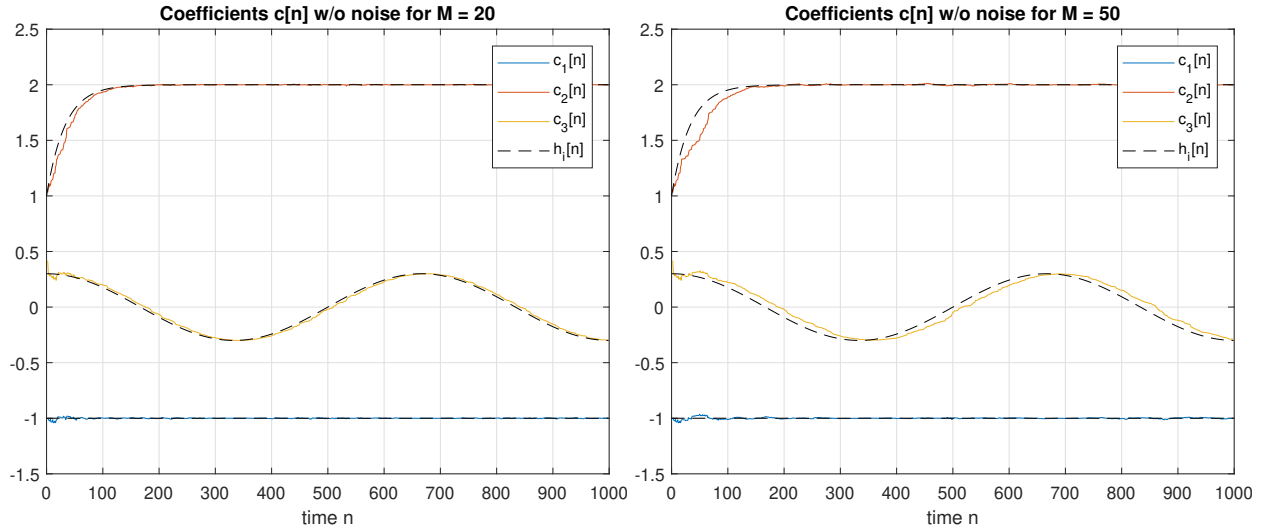


Figure 2: Coefficients for $\mathbf{c}[n]$ *without* noise and segment length $M = 20$ and $M = 50$

As we can see, the values for $\mathbf{h}[n]$ (black striped lines) correspond quite well with the coefficients which were calculated through the *least-square* sense. The first values of $\mathbf{c}[n]$ vary a lot from the correct solution due to assumption that $x[n] = 0$ for $n < 0$ and therefore the first $\mathbf{c}[n]$ get computed with only a few entries which are not equal to 0.

Between the two plots with different segment length M is only little difference. The initial deviation is longer due to the longer segment length and the plot gets a little bit more shifted (increases with segment length).

3.4 Task d)

Last but not least some noise $w[n]$ with variance $\sigma_w = 0.02$ disturbs the 'measured' data $d[n]$. Again the coefficients were calculated and plotted.

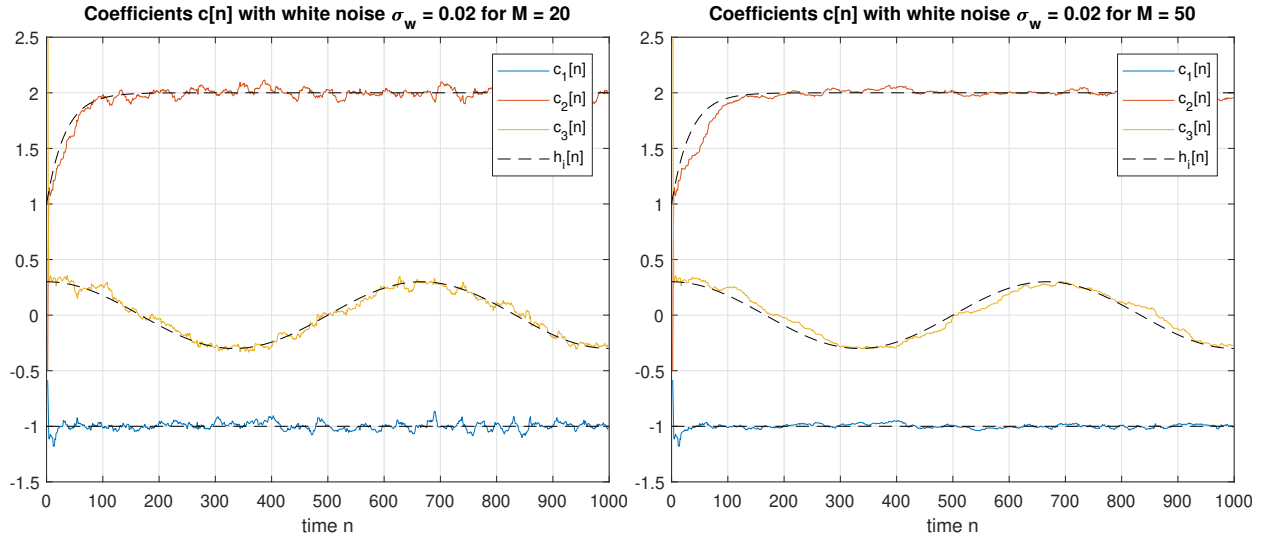


Figure 3: Coefficients for $c[n]$ with noise and segment length $M = 20$ and $M = 50$

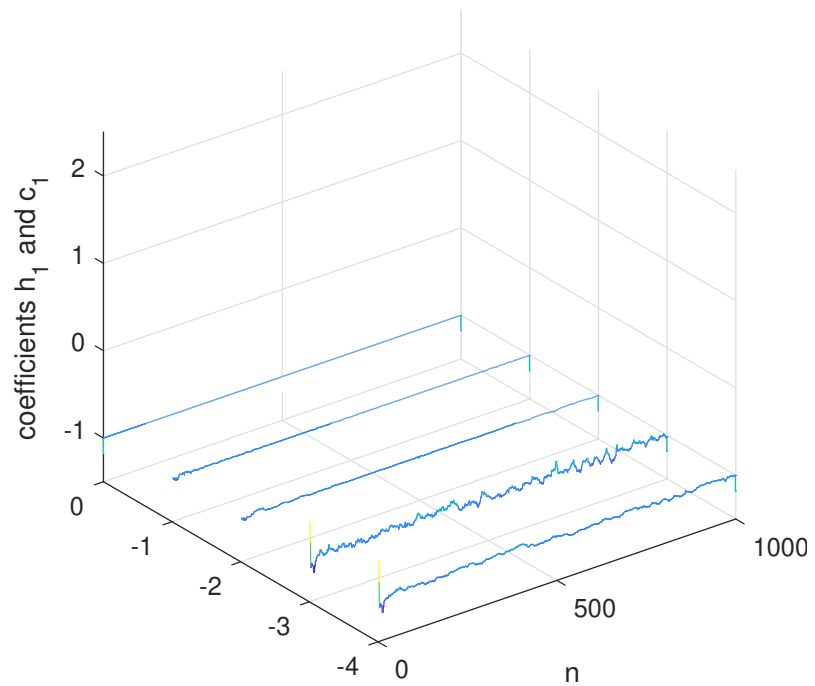
Between $M = 20$ and $M = 50$ is a lot of difference now. The additional noise $w[n]$ causes the coefficients to fluctuate. The plot with the lower segment length $M = 20$ fluctuates way more than the plot with segment length $M = 50$. Due to the property of white noise the effect it has should cancel out for infinite long observation. Since we only look at finite length of data we, some effects of the white noise still can be seen. To further decreases the noise the segment length can be increased.

WATERFALL PLOTS

Order of the lines, starting from top to bottom:

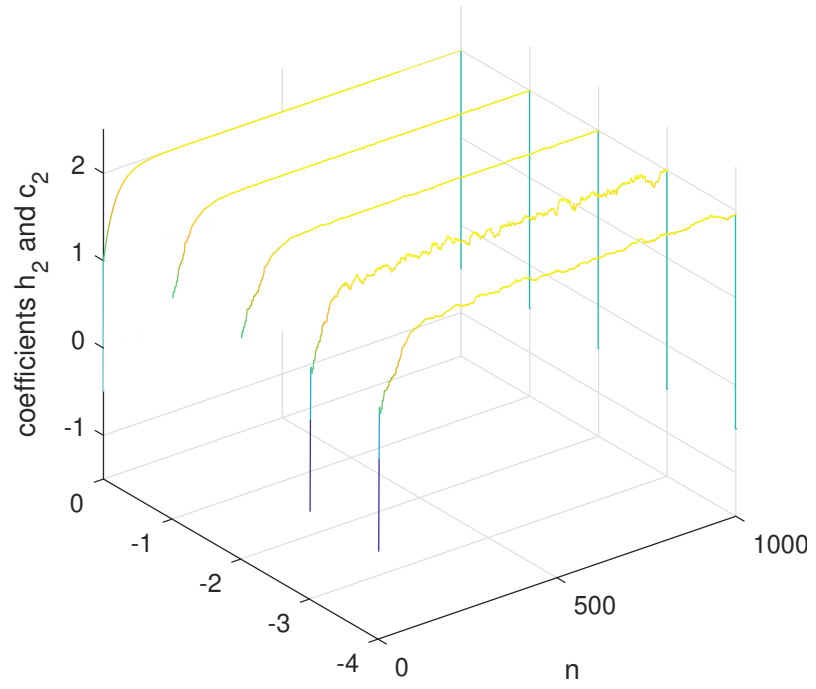
1. $h_i[n]$
2. $c_i[n]$ *without* measurement noise and for segment length $M = 20$
3. $c_i[n]$ *without* measurement noise and for segment length $M = 50$
4. $c_i[n]$ *with* measurement noise and for segment length $M = 20$
5. $c_i[n]$ *with* measurement noise and for segment length $M = 50$

where i denotes the i -th coefficient



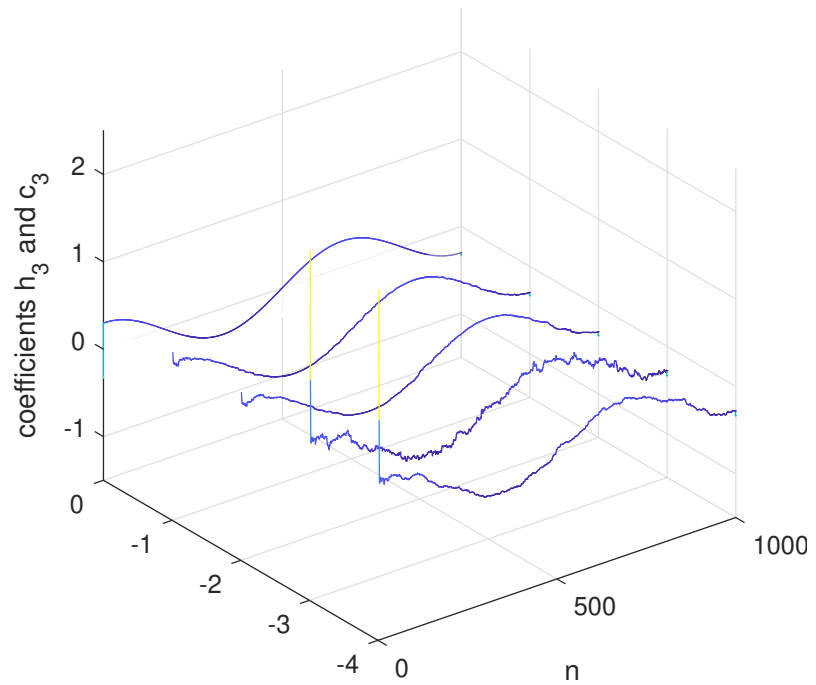
$h \mid M = 20 \ \& \ w = 0 \mid M = 50 \ \& \ w = 0 \mid M = 20 \ \& \ w \sim 0 \mid M = 50 \ \& \ w \sim 0;$

Figure 4: waterfall coefficients $i = 1$



$h \mid M = 20 \ \& \ w = 0 \mid M = 50 \ \& \ w = 0 \mid M = 20 \ \& \ w \sim 0 \mid M = 50 \ \& \ w \sim 0;$

Figure 5: waterfall coefficients $i = 2$



$h \mid M = 20 \ \& \ w = 0 \mid M = 50 \ \& \ w = 0 \mid M = 20 \ \& \ w \sim 0 \mid M = 50 \ \& \ w \sim 0;$

Figure 6: waterfall coefficients $i = 3$

3.5 Matlab Code

Main File

```
1 close all
2 clear all
3 clc
4
5 %suppress: "Warning: Matrix is singular to working precision."
6 id = 'MATLAB:singularMatrix';
7 warning('off',id)
8
9 %suppress: "Warning: Directory already exists."
10 id = 'MATLAB:MKDIR:DirectoryExists';
11 warning('off',id)
12
13 mkdir 'Figures' %create Figures folder
14
15
16
17 %-----
18 % a)
19 N = 2;
20 d = [-3; -5; 0]; %values from example 1.1
21 x = [-1; -1; 1];
22
23
24
25
26 k = N-1; %Pad with N - 1 zeroes for the values of x[-1], x[-2], ... x[-N+1]
27 x_pad = [zeros(k,1); x];
28 d_pad = [zeros(k,1); d];
29
30
31 c = ls_filter(x_pad, d_pad, N)
32
33 %"Probe"
34 d_hat = conv(x,c);
35 d_hat = d_hat(1:length(x)) %delete the values which assume that
36 %                          x[n] = 0 for n > length(x)
37
38
39
40 %-----
41 % b)
42 theta = 3*pi/1000;
43 n = 0:999;
44
45 h = [-1*ones(1,length(n)); 2-0.97.^n; 0.3*cos(theta*n)];
46 %h = h1[0], h1[1], ..., h1[n];
47 %     h2[0], h2[1], ..., h2[n];
48 %     h3[0], h3[1], ..., h3[n];
49
50
51 figure
52 plot(n,h)
53 legend('h_1[n]', 'h_2[n]', 'h_3[n]')
54 grid on
```

```

55     ylim([-1.5 2.5])
56     title('Coefficients h[n]')
57     xlabel('time n')
58
59     saveas(gcf, 'Figures/Coefficients_h', 'epsc')
60
61 %-----
62 % c) and d)
63 N = 3; %3 filter coefficients in h and c
64
65 x = randn(1, length(n)).'; %x[n] = 0 for n < 0 (or 1 in matlab)
66
67 d = vector_conv(x, h);
68 % d_test = vector_conv2(x, h)
69 % e_test = d - d_test
70 counter = 1;
71 for jj = 1:2
72
73     if jj == 1
74         w = 0;
75     else
76         %create white gaussian noise and change variance
77         w = transpose(randn(1, length(n)))./(1/sqrt(0.02));
78     end
79
80     d = d + w;% add noise after filter h
81
82
83     for M = [20, 50]
84         x_pad = [zeros(M-1,1); x]; %pad with M-1 zeros; x[n] = 0 for n < 0;
85         d_pad = [zeros(M-1,1); d]; %and pad d too for the newly created values
            of x[n]
86
87         c = zeros(N, length(n));
88         for ii = n %ii is counts through the time n
89             c(:, ii+1) = ls_filter(x_pad(ii+1:M+ii), d_pad(ii+1:M+ii), N);
90         end
91
92
93         if w == 0
94             text = ['Coefficients c[n] w/o noise for M = ' num2str(M)];
95             text_saveas = ['Coefficients_c_without_noise_M=' num2str(M)];
96         else
97             text = ['Coefficients c[n] with white noise \sigma_w = ' num2str(
                round(var(w),2)) ' for M = ' num2str(M)];
98             text_saveas = ['Coefficients_c_with_noise_M=' num2str(M)];
99         end
100
101         figure
102         plot(n, c)
103         hold on
104         plot(n, h, '—k')
105         legend('c_1[n]', 'c_2[n]', 'c_3[n]', 'h_i[n]')
106         grid on
107         title(text)
108         xlabel('time n')
109         ylim([-1.5, 2.5]) %due to some singularities, the first values
110 %             of c can get quite big -> ruins the plot ->

```

```

111 %                                limit it
112         saveas(gcf,['Figures/' text_saveas] , 'eps') %eps to save the eps
           in colour
113
114         c_plot(:, :, counter) = c;
115         counter = counter + 1;
116     end %for M
117
118 end %for jj
119
120 kk = 1;
121 X = [1; 1; 1; 1; 1]* n;
122 Y = [0; -1; -2; -3; -4]*ones(1, length(n));
123
124 for kk = 1:3
125     Z = [h(kk, :); c_plot(kk, :, 1); c_plot(kk, :, 2); c_plot(kk, :, 3); c_plot(kk, :, 4)];
126     %c_plot(kk, :, 1) has the coefficients for M = 20, w = 0;
127     %c_plot(kk, :, 2) has the coefficients for M = 50, w = 0;
128     %c_plot(kk, :, 3) has the coefficients for M = 20, w ~= 0;
129     %c_plot(kk, :, 4) has the coefficients for M = 50, w ~= 0;
130
131     text_saveas = ['waterfall_coefficients_' num2str(kk)];
132
133 figure
134     waterfall(X, Y, Z)
135     zlim([-1.5, 2.5])
136     xlabel('n')
137     zlabel(['coefficients h_' num2str(kk) ' and c_' num2str(kk)])
138     ylabel('h | M = 20 & w = 0 | M = 50 & w = 0 | M = 20 & w ~= 0 | M = 50 &
           w ~= 0;')
139     saveas(gcf,['Figures/' text_saveas] , 'eps') %eps to save the eps in colour
140 end
141
142
143 %create a placeholder function to overwrite the saveas function
144 function saveas(~, ~, ~)
145     disp('Figure not saved')
146 end
147
148
149 %seen from plots:
150 %w[n] = 0:
151 %theres only little difference between M = 20 and M = 50 and
152 %the values of c correspond quite well to the values of h,
153 %although the plot is a bit shifted(increases with segment
154 %length M). The first values of c are also not quite the same
155 %as h, due to the assumption that x[n] = 0, for n < 0 and the
156 %first c[n] gets computed with only one entry which is not 0;
157 %
158 %w[n] ~= 0:
159 %between M = 20 and M = 50 is a lot of difference now. Probably
160 %due to the higher amount of samples, the noise cancels out
161 %(and would fully cancel for M -> infty(because white noise),
162 %but then the adaptiveness of system would get lost -> c[n] would
163 %get constant if every x[n] is taken into account)

```


Function ls_filter()

```
1 function c = ls_filter(x, d, N)
2 %computes the filter coefficients c for one time instance n, where
3 %corresponds to the last entry of x
4
5 % x is the input signal saved as col vector
6 % d is the reference signal saved as col vector
7 % N is the order of the filter i.e. the amount of coefficients in c
8
9 % Matrix X to compute the coefficients at time instance n
10 % X = [x[n-M+N], x[n-M+N-1], ..., x[n-M];
11 %       x[n-M+N+1], x[n-M+N], ..., x[n-M+1];
12 %       ...
13 %       x[n], x[n-1] ..., x[n-N+1] ];
14
15 % Make sure x and d are col vectors
16 % x = x(:);
17 % d = d(:);
18
19 if isrow(x) || isrow(d)
20     error('vector x or vector d is not a column vector')
21 end
22
23 M = length(x); %segment length
24
25 X = zeros(M-N+1,N); %create placeholder for entries of X
26
27 for ii = 0:N-1 %for order N coefficients of c we need N cols
28     X(:,ii+1) = x( (end-M+N)-ii:end-ii ); %end corresponds to current time n
29 end %to include M-N+1 we have to subtract
30     -M+N
31 c = (X.' * X)^-1 * X.' * d(end-M+N:end); %pseudo inverse; multiply with
32     segmented d
33 end
```

Function vector_conv()

```
1 function y = vector_conv(x, h)
2 %calculates the convolution sum defined in Adaptive System UE
3 %
4 %x has to be a column vector in form of
5 % x = x[0];
6 %     x[1];
7 %     ...;
8 %     x[n-1]
9 %
10 %where n is the time variable
11
12
13 %h has to be matrix in the form of
14 % h = h1[0], h1[1], ..., h1[n-1];
15 %     h2[0], h2[1], ..., h2[n-1];
16 %     h3[0], h3[1], ..., h3[n-1]
17
18
19 x = x(:); %make sure that x is a col vector
20
21
22 N = size(h,1);
23
24
25 t = 1;
26 n = 0:length(x)-1;
27 x_zero_pad = [zeros(N-1,1); x]; %puts zeros for time x[-1], x[-2], ... x[-N+1]
28
29 y = zeros(length(n),1);
30 for n_shift = n + N
31     x_tap_input = x_zero_pad(n_shift:-1:n_shift-N+1);
32     y(t) = h(:,t)' * x_tap_input; % ' is hermitian transposed
33     t = t+1;
34 end
35
36
37 end
```

Function vector_conv2()

```
1 function y = vector_conv2(x, h)
2 % computes the convolution of x and the coefficients of h at time
3 % instance n for every n
4
5 y = zeros(length(x),1);
6
7 for n = 1:length(x)
8     temp = conv(x,h(:,n));
9     y(n) = temp(n);
10 end
11
12 end
```