

# AE 370: HW1

Linyi Hou

February 6, 2020

## Problem 1

The following are examples of invalid inner products:

$$(u, v) = u(x) + v(x), \forall u, v \in C[1, 2]$$

Not an inner product because  $(u, v)$  returns a function and not a scalar.

$$(u, v) = uv - 1, \forall u, v \in \mathbb{R}$$

Not an inner product because  $(0, 0) = -1 < 0$

$$(u, v) = u^v, \forall u, v \in \mathbb{Z}$$

Not an inner product because  $(\alpha u + \beta v, w) = (\alpha u + \beta v)^w \neq \alpha u^w + \beta v^w = \alpha(u, w) + \beta(v, w)$

## Problem 2

### 2.a

Let  $\mathbb{B}$  be a basis for subspace  $W$  of the vector space  $S = \{f(x) \mid x \in [a, b]\}$ . Then,

$$f_a(x) = \sum_{i=1}^{n+1} c_i b_i(x)$$

Consider  $n + 1$  coefficients  $\{c_1, c_2, \dots, c_{n+1}\}$ , which require  $n + 1$  equations to solve.

If we take  $n + 1$  points on  $[a, b]$ :  $\{x_1, x_2, \dots, x_{n+1}\}$ , then

$$f(x_j) = \sum_{i=1}^{n+1} c_i b_i(x_j)$$

Expanding into a linear system  $Ac = f$ :

$$\begin{bmatrix} b_1(x_1) & b_2(x_1) & \cdots \\ b_2(x_1) & \ddots & \vdots \\ b_{n+1}(x_{n+1}) & \cdots & b_{n+1}(x_{n+1}) \end{bmatrix} \begin{bmatrix} 1 \\ c_2 \\ \vdots \\ c_{n+1} \end{bmatrix} = \begin{bmatrix} f(x_1) \\ f(x_2) \\ \vdots \\ f(x_{n+1}) \end{bmatrix}$$

## 2.b, 2.c

\*See MATLAB figures in Appendix A and code in Appendix B.

The condition number is a metric of sensitivity to error for matrix systems. The condition number remains at 1 for the Lagrange basis approximation for all values of  $n$  while the condition number for the monomial basis grows exponentially. Hence Lagrange is a better approximation method due to its insensitivity to error at large sample sizes.

## 2.d

Chebyshev points rely on being able to solve the linear system  $Ac = f$  accurately to produce good results. Using the Lagrange basis, the  $c$  matrix is easy to compute since it simply contains the values of the sample points  $f(x_i)$ . Hence the  $A$  matrix is simply the identity matrix, which is invertible in the operation  $c = A^{-1} * f$ .

However, in the monomial basis, computing the inverse of the  $A$  matrix becomes difficult since sample points are close for large sample sizes, making certain rows of the  $A$  matrix almost linearly dependent. This causes the  $A$  matrix to be close to singular/non-invertible. A computer program with finite accuracy will thus be prone to error for large samples.

## Problem 3

\*See MATLAB figures in Appendix A and code in Appendix B.

## Appendix A: Figures

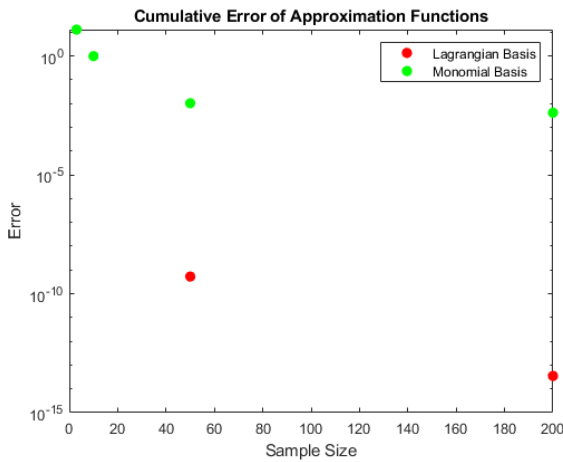


Figure 1: Problem 2.b Error Plots

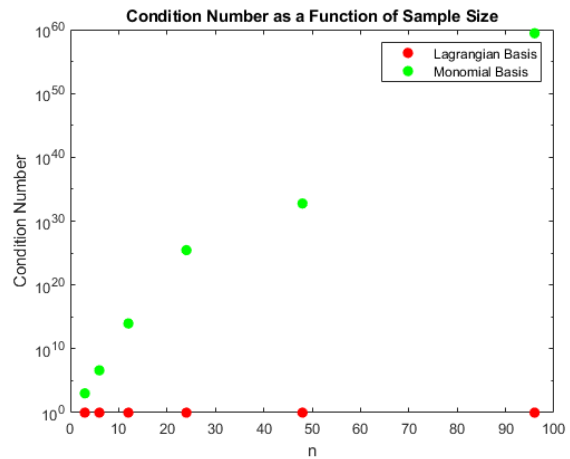


Figure 2: Problem 2.c Condition Numbers

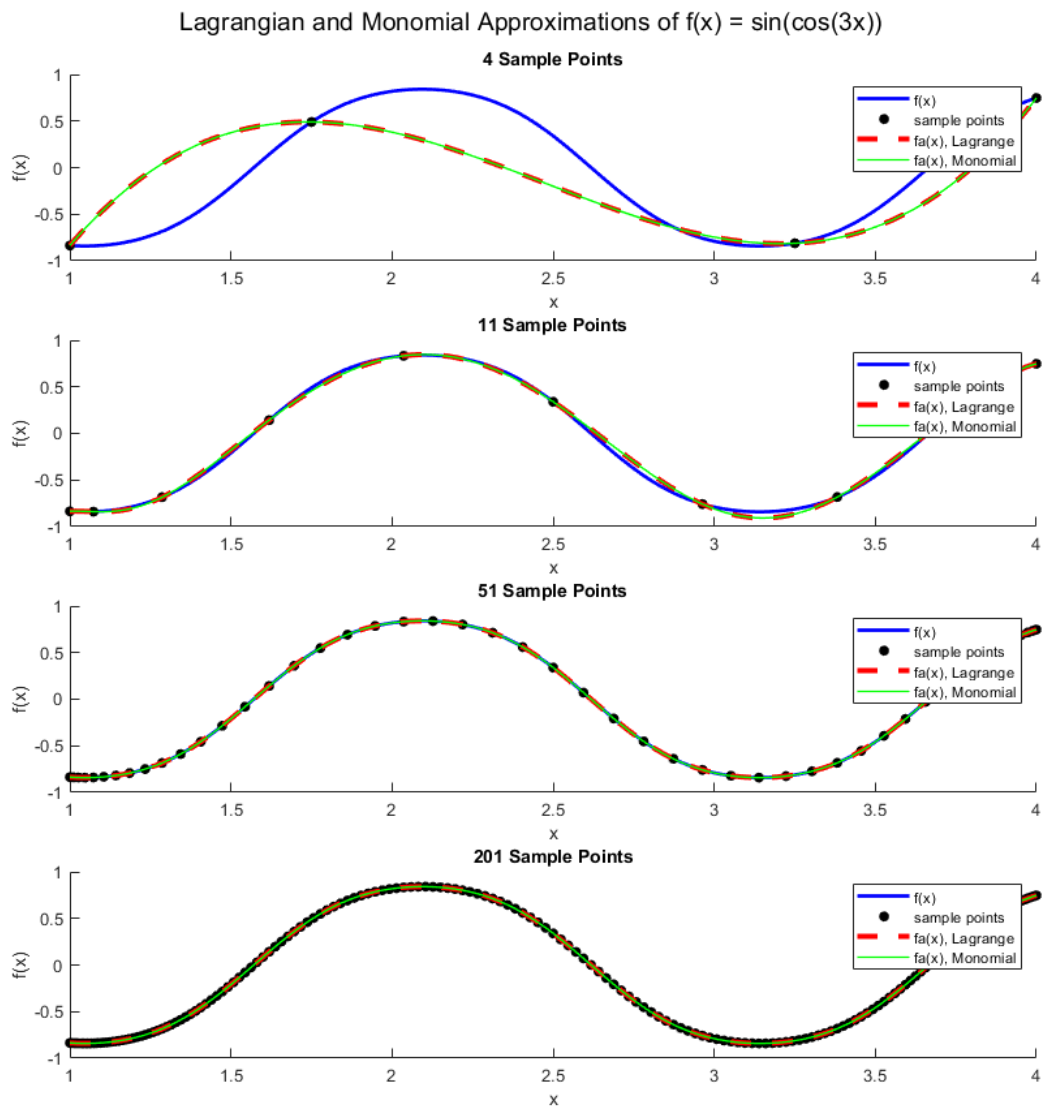


Figure 3: Problem 2.b Function Plots

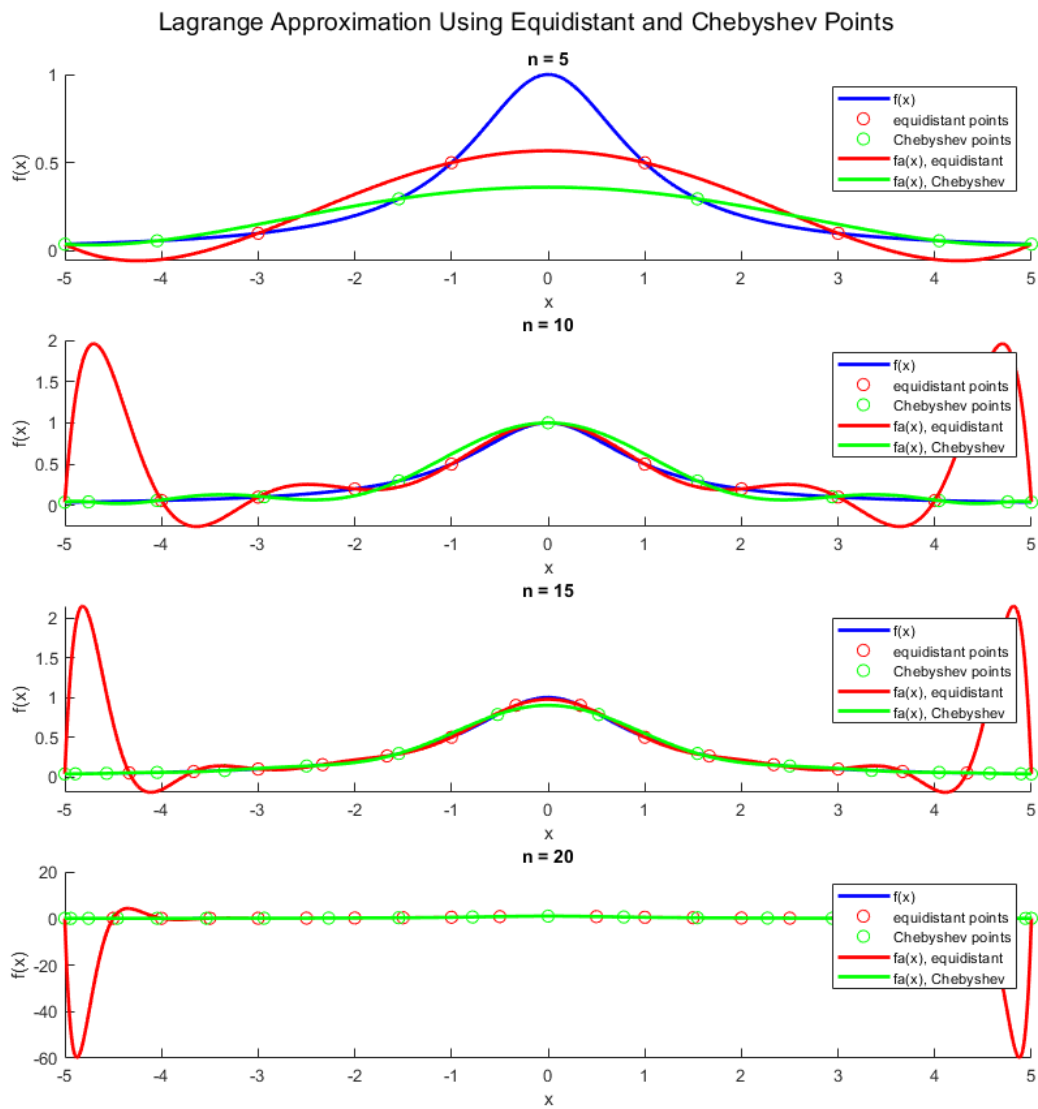


Figure 4: Problem 3.a, 3.b Function Plots

## Appendix B: Code

### MATLAB Code for Problem 2

```
1 % -----
2 % Lagrange basis
3
4 n_vect = [3,10,50,200];
5 f      = @(x) sin(cos(3*x));
6 intv   = [1,4];
7 err_L  = zeros(size(n_vect));
8
9 for i = 1:length(n_vect)
10     n = n_vect(i);
11
12     xp = (intv(1)+intv(2))/2 + (intv(1)-intv(2))/2*cos((0:n)*pi/n);
13     fp = f(xp);
14     dp = fp;
15     A = eye(n+1);
16
17     xx = linspace(intv(1),intv(2),1000);
18     pn_L = 0;
19     % L = @(x) 0;
20     for j = 1:n+1
21         Li = @(x) 1;
22
23         for k = 1:n+1
24             if k ~= j
25                 Li = @(x) (x-xp(k)) ./ (xp(j)-xp(k)) .* Li(x);
26             end
27         end
28     % L = @(x) L(x) + dp(j) .* Li(x);
29     pn_L = pn_L + dp(j) * Li(xx);
30 end
31
32 figure(1)
33 subplot(4,1,i)
34 hold on
35 plot(xx,f(xx),'b-','lineWidth',2)
36 plot(xp,f(xp),'k.','markersize',20)
```

```

37     plot(xx,pn_L,'r—','linewidth',3)
38     title([num2str(n+1) ' Sample Points'])
39     xlabel('x')
40     ylabel('f(x)')
41     err_L(i) = norm(f(xx)-pn_L);
42 end
43
44 hold off
45
46 % -----
47 % Monomial basis
48
49 err_M = zeros(size(n_vect));
50
51 for i = 1:length(n_vect)
52     n = n_vect(i);
53
54     xp = (intv(1)+intv(2))/2 + (intv(1)-intv(2))/2*cos((0:n)*pi/n);
55     fp = f(xp);
56
57     A = zeros(n+1,n+1);
58     for j = 1:n+1
59         A(:,j) = xp' .^ (j-1);
60     end
61     dp = A \ (fp');
62
63     xx = linspace(intv(1),intv(2),1000);
64     M = @(x) 0;
65     for j = 1:n+1
66         M = @(x) dp(j) .* x.^(j-1) + M(x);
67     end
68     pn_M = M(xx);
69
70     figure(1)
71     hold on
72     subplot(4,1,i)
73     plot(xx,pn_M,'g—','lineWidth',1)
74     legend('f(x)','sample points','fa(x), Lagrange','fa(x), Monomial')
75     title([num2str(n+1) ' Sample Points'])
76     xlabel('x')

```

```

77     ylabel('f(x)')
78     err_M(i) = norm(f(xx)-pn_M);
79 end
80
81 sgtitle('Lagrangian and Monomial Approximations of f(x) = sin(cos(3x))')
82
83 hold off
84
85 figure(100)
86 semilogy(n_vect,err_L,'r.','markersize',24,'linewidth',2), hold on
87 semilogy(n_vect,err_M,'g.','markersize',24,'linewidth',2), hold off
88 legend('Lagrangian Basis','Monomial Basis')
89 title('Cumulative Error of Approximation Functions')
90 xlabel('Sample Size')
91 ylabel('Error')
92
93
94 % -----
95 % Part c
96
97 n_vect = [3,6,12,24,48,96];
98 intv   = [1,4];
99 condn_L = zeros(length(n_vect),1);
100 condn_M = condn_L;
101
102 % Lagrangian Basis
103 for i = 1:length(n_vect)
104     n = n_vect(i);
105     A = eye(n+1);
106     condn_L(i) = cond(A);
107 end
108
109 % Monomial Basis
110 for i = 1:length(n_vect)
111     n = n_vect(i);
112     xp = (intv(1)+intv(2))/2 + (intv(1)-intv(2))/2*cos((0:n)*pi/n);
113     A = zeros(n+1,n+1);
114     for j = 1:n+1
115         A(:,j) = xp' .^ (j-1);
116     end

```

```

117     condn_M(i) = cond(A);
118 end
119
120 figure(200)
121 semilogy(n_vect,condn_L,'r.','markersize',24,'linewidth',2), hold on
122 semilogy(n_vect,condn_M,'g.','markersize',24,'linewidth',2), hold off
123 legend('Lagrangian Basis','Monomial Basis')
124 title('Condition Number as a Function of Sample Size')
125 xlabel('n')
126 ylabel('Condition Number')

```

## MATLAB Code for Problem 3

```

1  % -----
2  % Lagrange basis
3
4  n_vect = [5,10,15,20];
5  f      = @(x) 1 ./ (1+x.^2);
6  intv   = [-5,5];
7
8  for i = 1:length(n_vect)
9      n = n_vect(i);
10
11     xp_E = intv(1):(intv(2)-intv(1))/n:intv(2);
12     xp_C = (intv(1)+intv(2))/2 + (intv(1)-intv(2))/2*cos((0:n)*pi/n);
13     fp_E = f(xp_E);
14     fp_C = f(xp_C);
15     dp_E = fp_E;
16     dp_C = fp_C;
17
18     xx = linspace(intv(1),intv(2),1000);
19     pn_E = 0;
20     for j = 1:n+1
21         Li = @(x) 1;
22
23         for k = 1:n+1
24             if k ~= j
25                 Li = @(x) (x-xp_E(k)) ./ (xp_E(j)-xp_E(k)) .* Li(x);
26             end

```



```

27         end
28         pn_E = pn_E + dp_E(j) * Li(xx);
29     end
30
31     pn_C = 0;
32     for j = 1:n+1
33         Li = @(x) 1;
34
35         for k = 1:n+1
36             if k ~= j
37                 Li = @(x) (x-xp_C(k)) ./ (xp_C(j)-xp_C(k)) .* Li(x);
38             end
39         end
40         pn_C = pn_C + dp_C(j) * Li(xx);
41     end
42
43     figure(3)
44     subplot(4,1,i)
45     hold on
46     plot(xx,f(xx),'b-','linewidth',2)
47     plot(xp_E,f(xp_E),'ro','markersize',7)
48     plot(xp_C,f(xp_C),'go','markersize',7)
49     plot(xx,pn_E,'r-','linewidth',2)
50     plot(xx,pn_C,'g-','linewidth',2)
51     legend( 'f(x)',...
52            'equidistant points',...
53            'Chebyshev points',...
54            'fa(x), equidistant',...
55            'fa(x), Chebyshev')
56     title(['n = ' num2str(n)])
57     xlabel('x')
58     ylabel('f(x)')
59
60 end
61
62 sgtitle('Lagrange Approximation Using Equidistant and Chebyshev Points')
63
64 hold off

```