

Contents

- Name: Gurpreet Singh
- Problem# 1
- a. Approximate $f(0.25)$ using linear interpolation with $x_0 = 0$, and $x_1 = 0.5$.
- b. Approximate $f(0.75)$ using linear interpolation with $x_0 = 0.5$, and $x_1 = 1$.
- c. Approximate $f(0.25)$ and $f(0.75)$ by using the second degree interpolating polynomial with $x_0 = 0$, $x_1 = 1$, and $x_2 = 2$.
- d. Which approximations are better and why?
- Problem# 2
- a. Show that the polynomials $P(x) = 3 - 2(x + 1) + 0(x + 1)(x) + (x + 1)(x)(x - 1)$ and $Q(x) = -1 + 4(x + 1) - 3(x + 2)(x + 1) + (x + 2)(x + 1)(x)$ both interpolate the data $f(-2) = -1$, $f(-1) = 3$, $f(0) = 1$, $f(1) = -1$, $f(2) = 3$.
- b. Why does part (a) not violate the uniqueness property of interpolating polynomials?
- 3. Problem #3
- Problem # 4

Name: Gurpreet Singh

```
% HW# 2
```

Problem# 1

```
% Let f(x) = e^x, for 0 <= x <= 2
```

a. Approximate $f(0.25)$ using linear interpolation with $x_0 = 0$, and $x_1 = 0.5$.

```
n = linspace(0, 2, 5);    %      0      0.5000      1.0000      1.5000      2.0000
y = exp(1).^n;            % 1.0000      1.6487      2.7183      4.4817      7.3891

% y(x)    =      a0 + a1(x)
% y(0)    = 1.0000 = a0 + a1(0)
% y(0.5)  = 1.6487 = a0 + a1(0.5)

% [1    0] [a0]      [1.0000]
% [1 0.5] [a1]  =  [1.6487]

% a0 = 1.0000
% a1 = 1.2974

% ans = 1.0000 + 1.2974*(0.25);
% ans = 1.3244

y1 = [y(1); y(2)];
x1 = [0; 0.5];
a1 = polyfit(x1, y1, 1);
q1 = polyval(a1, 0.25)
```

q1 =

1.3244

b. Approximate $f(0.75)$ using linear interpolation with $x_0 = 0.5$, and $x_1 = 1$.

```
% y(x)      =      a0 + a1(x)
% y(0.5)    = 1.6487 = a0 + a1(0.5)
% y(1.0)    = 2.7183 = a0 + a1(0.75)

% [1 0.50] [a0]      [1.6487]
% [1 1.00] [a1]  =   [2.7183]

% a0 = 0.5792
% a1 = 2.1391

% ans = 0.5792 + 2.1391*(0.75);
% ans = 2.1835

y2 = [y(2); y(3)];
x2 = [0.5; 1.0];
a2 = polyfit(x2, y2, 1);
q2 = polyval(a2, 0.75)
```

q2 =

2.1835

c. Approximate $f(0.25)$ and $f(0.75)$ by using the second degree interpolating polynomial with $x_0 = 0$, $x_1 = 1$, and $x_2 = 2$.

```
% y(x)      = a0 + a1(x) + a2(x)^2
% y(0) = 1.0000 = a0 + a1(0) + a2(0)^2
% y(1) = 2.7183 = a0 + a1(1) + a2(1)^2
% y(2) = 7.3891 = a0 + a1(2) + a2(2)^2

% [1 0 0] [a0] = [1.0000]
% [1 1 1] [a1] = [2.7183]
% [1 2 4] [a2] = [7.3891]

% x = [1 0 0; 1 1 1; 1 2 4];
% y = [1; 2.7183; 7.3891];
% x\y

% a0 = 1.0000
% a1 = 0.2421
% a2 = 1.4762

% f(0.25)
% -----
```

```

% ans = 1.0000 + 0.2421*(0.25) + 1.4762*(0.25)^2;
% ans = 1.1528

y3 = [y(1); y(3); y(5)];
x3 = [0.0; 1.0; 2.0];
a3 = polyfit(x3, y3, 2);
q3 = polyval(a3, 0.25)

% f(0.75)
% -----
% ans = 1.0000 + 0.2421*(0.75) + 1.4762*(0.75)^2;
% ans = 2.0119
q4 = polyval(a3, 0.75)

```

```

q3 =

    1.1528

```

```

q4 =

    2.0119

```

d. Which approximations are better and why?

```

nn = linspace(0, 2, 100);
yy = exp(1).^nn;

% disp('The linear interpolation is a better approximation. It can be seen in the graph below.
% Linear interpolation worked better because the data was small and it was smooth.')

% figure
% plot(nn, yy)
% hold on
% plot(n, y)
% hold on
% plot(0.25, q1, 'b*')
% hold on
% plot(0.75, q2, 'b*')
% hold on
% plot(0.25, q3, 'r*')
% hold on
% plot(0.75, q4, 'r*')
% % , 0.75, q2, 0.25, q3, 0.75, q4)
%
% legend('e^x 100 points 0 <= x <= 2 ', 'e^x 5 points 0 <= x <= 2', 'linear interpolation 1st
% degree', 'second degree interpolating polynomial');

```

Problem# 2

a. Show that the polynomials $P(x) = 3 - 2(x + 1) + 0(x + 1)(x) + (x + 1)(x)(x - 1)$ and $Q(x) = -1 + 4(x + 1) - 3(x + 2)(x + 1) + (x + 2)(x + 1)(x)$ both interpolate the data $f(-2) = -1$, $f(-1) = 3$, $f(0) = 1$, $f(1) = -1$, $f(2) = 3$.

```

% a.
% ----
name = {'n'; 'Q(x)'; 'P(x)'};
n = [1; 2; 3];
px = zeros(1, 100000);
qx = zeros(1, 100000);

for x=1:100000
    px(x) = 3 - 2*(x + 1) + 0*(x + 1)*(x) + (x+1)*(x)*(x - 1);
    qx(x) = -1 + 4*(x+2) - 3*(x+2)*(x+1) + (x+2)*(x+1)*(x);
end

% plotted the graph of px, qx from 1, 100000, and the lines are both the
% same because they output the same value for the f(x). They are the same
% polynomial.
% figure
% plot(px)
% hold on
% plot(qx)
% hold on
% plot(-qx)

```

b. Why does part (a) not violate the uniqueness property of interpolating polynomials?

```

% ----
% P(x) = 3 - 2(x + 1) + 0(x + 1)(x) + (x + 1)(x)(x - 1)
% P(x) = 3 - 2(x + 1) + (x + 1)(x)(x - 1)
% P(x) = 3 - 2x - 2 + (x + 1)(x)(x - 1)
% P(x) = 3 - 2x - 2 + x^3 - x
% P(x) = x^3 - 3x + 1

% Q(x) = -1 + 4(x + 2) - 3(x + 2)(x + 1) + (x + 2)(x + 1)(x)
% Q(x) = -1 + 4x + 8 - 3x^2 - 9x - 6 + x^3 + 3x^2 + 2x
% Q(x) = -1 + 4x + 8 - 9x - 6 + x^3 + 2x
% Q(x) = -1 - 3x + 8 - 6 + x^3
% Q(x) = x^3 - 3x + 1

% P(x) = Q(x) both interpolate the same polynomials.

```

3. Problem #3

```

% Analogous to LinInterp2D, write a function CubicInterp2D(xc,yc,a,b,n,c,d,m,fA) that does cubic
% interpolation from a matrix of
% f(x, y) evaluations. Start by figuring out where? (xc, yc) is in the grid with respect to x
% = linspace(a,b,n) and
% y = linspace(c,d,m). Suppose this is the situation: as in LinearInterp2D.

a = 0; b = 5; n = 10;
c = 0; d = 5; m = 10;

fA = SetUp('Humps2D', a, b, n, c, d, m);
xc = a+3;
yc = c+4;

```

```
% cubically interpolated value for f(xc, yc).  
z = CubicInterp2D(xc, yc, a, b, n, c, d, m, fA);  
E = abs(z - Humps2D(xc, yc))
```

E =

0.1448

Problem # 4

```
% Suppose f(u,v) is a function of two variables defined everywhere in the plane. Assume that  
x, y and fvals  
% are column 3-vectors and p is a column 2-vector. Assume that (p(1), p(2)) is inside the tri  
angle  
% defined by (x(1),y(1)), (x(2),y(2)), and (x(3),y(3)) and that fvals(i) = f(x(i),y(i)) for  
i=1:3.  
% fp is an estimate of f(p(1),p(2)) obtained by linear interpolation.  
  
x = [0; 5; 1];  
y = [0; 10; 5];  
fvals = [0; 50; 5];  
p = [5; 1];  
InterpTri(x, y, fvals, p);
```

fp =

0	-0.6250	0
0	5.0000	0
0	0.5000	0