# 1 Interpolation and Polynomial Approximation

Equations which interpolate given a set of points.

## 1.1 Lagrange Polynomials

Given a set of points x, and the output of their unknown function f, an interpolating function P of order n can be constructed.

$$P(x) = \sum_{k=0}^{n} \left[ f(x_k) \prod_{i=0, i \neq k}^{n} \frac{x - x_i}{x_k - x_i} \right]$$

Lagrange Polynomials are effective for the general case, but prone to round-off error.

```
function [P] = lagrange(x, y, p)
       n = length(y);
       P = zeros(1, length(p));
       for k=1:n
           L = ones(1, length(p));
           for j = [1:k-1 \ k+1:n]
                L = L .* ((p - x(j)) ./ (x(k) - x(j)));
10
           P = P + L.*y(k);
11
12
       end
13
       fprintf('Point %.2f, Value %.8f \n', [p; P]);
14
15
  end
16
```

#### 1.2 Neville's Method

Given a set of points x, and the output of their unknown function f, an interpolating polynomial can be recursively constructed.

$$Q_{i,j} = \frac{(x - x_{i-j})Q_{i,j-1} - (x - x_i)Q_{i-1,j-1}}{x_i - x_{i-j}}$$

Neville's Method reduces the computations required for interpolants and therefore reduces the round-off error from lagrange.

```
\begin{array}{lll} & \text{function} & [P] = \text{neville}(x,\;y,\;p) \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ &
```

#### 1.3 Newton Divided Differences

Given a set of points x and the output of their unknwn function f. A divided difference formula is recursively applied.

$$f[x_i, ..., x_{i+k}] = \frac{f[x_{i+1}, ..., x_{i+k}] - f[x_i, ..., x_{i+k-1}]}{x_{i+k} - x_i}$$

These functions are then used to find the interpolating polynomial  $P_n$ .

$$P_n(x) = f[x_0] + \sum_{k=1}^{n} \left[ f[x_0, ..., x_k] \prod_{i=0}^{k-1} (x - x_i) \right]$$

Divided Differences can have additional points given to it without recalculation of the polynomial.

```
function [P, v] = divided_differences(x, y, p)
       n = length(x);
       D = zeros(n,n);
       D(:,1) = y;
       for j=2:n,
            for k=j:n,
                D(k,j) = (D(k,j-1)-D(k-1,j-1))/(x(k)-x(k-i+1));
            end
       end
10
11
       P = diag(D);
12
       if exist ('p', 'var')
14
           v = P(1);
15
            for i=1:n-1
16
                v = v + P(i+1)*prod(p*ones(1,i)-x(1:i));
17
            end
       end
19
   end
20
```

#### 1.4 Newton Equally Spaced Differences

Given a set of equally spaced points x and the output of their unkown function f, ordered in increasing order. A forward difference formula is recursively applied.

$$P_n(x) = f(x_0) + \sum_{k=1}^n \binom{s}{k} \Delta^k f(x_0)$$

If the equally spaced points are in decreasing order, then backward differences should be used.

$$P_n(x) = f[x_n] + \sum_{k=1}^{n} (-1)^k {\binom{-s}{k}} \Delta^k f(x_n)$$

The Divided Difference algorithm can then be applied.

#### 1.5 Hermite's Method

Hermite Polynomials H agree on the value of the function and its derivative at the points given.

Given a set of points x and the values of their function f and its derivative f'. Using divided differences and creating virtual nodes z to represent the derivative values, the function can be interpolated.

```
function [P, v] = hermite(x, y, yp, p)
       n = length(x);
       z = zeros(2*n,1);
       Q = zeros(2*n, 2*n);
       for i=1:n
            z(2*i-1) = x(i);
            z(2*i) = x(i);
            Q(2*i-1,1) = y(i);
            Q(2*i,1) = y(i);
11
            Q(2*i,2) = yp(i);
12
            if i~=1
13
                Q(2*i-1,2) = (Q(2*i-1,1)-Q(2*i-2,1)) / (z(2*i-1)-z(2*i-2));
14
15
       end
16
17
       for i = 2:2*n-1
18
           for j=2:i
19
                Q(i+1,j+1) = (Q(i+1,j)-Q(i,j))/(z(i+1)-z(i-j+1));
20
            end
21
       end
22
23
       P = diag(Q);
24
25
       if exist('p', 'var')
26
            v = P(1);
27
            for i = 1:2*n-1
28
                v = v + P(i+1)*prod(p*ones(i,1)-z(1:i));
            end
30
       end
31
   end
32
```

### 1.6 Natural Cubic Splines

Splines break the domain into piecewise portions, using a different polynomial for each portion. They must be continuous to the second derivative across pieces.

Given a set of points x and the values of their function f. A spline  $S_j$  is constructed.

$$S_j(x) = a_j + b_j(x - x_j) + c_j(x - x_j)^2 + d_j(x - x_j)^3$$

A natural spline will be linear at its bounds a, b. That is, S''(a) = 0 and S'(b) = 0. A system of equations is constructed and solved.

```
function [S] = natural_cubic_splines (x, y, p)
```

```
n = length(x) - 1;
3
       m = n - 1;
4
       A = y;
5
       h = zeros(1, n);
       alpha = h;
       for i=1:n
9
            h(i) = x(i+1)-x(i);
10
       end
11
12
       for i=1:m
13
            alpha(i+1) = 3.0*(A(i+2)*h(i)-A(i+1)*(x(i+2)-x(i))+A(i)*h(i+1))/(h(i+1))
14
                +1)*h(i);
       end
15
16
       1 = zeros(1, n+1);
17
       mu = 1;
18
       z = 1;
19
       1(1) = 1;
20
       mu(1) = 0;
21
       z(1) = 0;
22
23
       for i=1:m
24
            1(i+1) = 2*(x(i+2)-x(i))-h(i)*mu(i);
25
            mu(i+1) = h(i+1)/l(i+1);
26
            z(i+1) = (alpha(i+1)-h(i)*z(i))/l(i+1);
27
       end
28
29
       l(end) = 1;
30
       z(end) = 0;
31
32
       B = zeros(1,n+1);
33
       C = zeros(1, n+1);
34
       D = zeros(1,n+1);
35
       C(end) = z(end);
36
37
       for i = 0:m
38
            j = m-i;
            C(j+1) = z(j+1)-mu(j+1)*C(j+2);
40
            B(j+1) = (A(j+2)-A(j+1))/h(j+1)-h(j+1)*(C(j+2)+2.0*C(j+1))/3.0;
            D(j+1) = (C(j+2)-C(j+1))/(3.0*h(j+1));
42
       end
43
44
       S = [A; B; C; D]';
45
   end
46
```

# 1.7 Clamped Cubic Splines

Given a set of points x, the values of their function f, and the values of their first derivative at the endpoints. A spline  $S_i$  is constructed. A spline is constructed with the additional endpoint constraint.

```
function [S] = clamped_cubic_splines (x, y, yp)
n = length(x)-1;
```

```
m = n - 1;
4
       A = y;
5
       h = zeros(1, m+1);
6
       for i=1:n
           h(i) = x(i+1)-x(i);
       end
10
11
       alpha = zeros(1, n+1);
12
       alpha(1) = 3.0*(A(2)-A(1))/h(1)-3.0*yp(1);
13
       alpha(end) = 3.0*yp(2) -3.0*(A(n+1)-A(n))/h(n);
14
15
       for i=1:m
16
            alpha(i+1) = 3.0*(A(i+2)*h(i)-A(i+1)*(x(i+2)-x(i))+A(i)*h(i+1))/(h(i+1))
17
               +1)*h(i);
       end
18
19
       1 = zeros(1, n+1);
20
       mu = zeros(1,n+1);
21
       z = zeros(1,n+1);
22
       1(1) = 1;
23
       mu(1) = 0;
24
       z(1) = 0;
25
26
       for i=1:m
27
            1(i+1) = 2*(x(i+2)-x(i))-h(i)*mu(i);
28
           mu(i+1) = h(i+1)/l(i+1);
29
           z(i+1) = (alpha(i+1)-h(i)*z(i))/l(i+1);
30
       end
31
32
       1(n+1) = h(n)*(2-mu(n));
33
       z(n+1) = (alpha(n+1)-h(n)*z(n))/l(n+1);
34
35
       B = zeros(1,n+1);
36
       C = zeros(1,n+1);
37
       D = zeros(1,n+1);
38
       C(n+1) = z(n+1);
39
40
       for i = 1:n
41
           j = n-i;
42
           C(j+1) = z(j+1)-mu(j+1)*C(j+2);
43
           B(j+1) = (A(j+2)-A(j+1))/h(j+1)-h(j+1)*(C(j+2)+2.0*C(j+1))/3.0;
44
           D(j+1) = (C(j+2)-C(j+1))/(3.0*h(j+1));
45
       end
46
47
       S = [A; B; C; D]';
48
  end
49
```