Numerical Analysis

Aakash Jog

2015-16

Contents

1	Lec	turer Information	3
2	Required Reading		
3		ating Point Representation Loss of Significant Digits in Addition and Subtraction	4
4		des of Approximations Order of Convergence	9
5	Rep	oresentation of Polynomials	10
	5.1	Power series	10
	5.2	Shifted Power Series	12
	5.3	Newton's Form	13
	5.4	Nested Newton's Form	13
	5.5	Properties of Polynomials	13

© (§ (9)

This work is licensed under the Creative Commons Attribution-NonCommercial-ShareAlike 4.0 International License. To view a copy of this license, visit http://creativecommons.org/licenses/by-nc-sa/4.0/.

6	Interpolation				
	6.1	Direct Method	15		
	6.2	Lagrange's Interpolation	16		
	6.3	Hermite Polynomials	20		
	6.4	Newton's Interpolation	20		

1 Lecturer Information

Prof. Nir Sochen

Office: Schreiber 201

Telephone: +972 3-640-8044 E-mail: sochen@post.tau.ac.il

Office Hours: Sundays, 10:00-12:00

2 Required Reading

1. S. D. Conte and C. de Boor, Elementary Numerical Analysis, 1972

3 Floating Point Representation

Exercise 1.

Represent 9.75 in base 2.

Solution 1.

$$9.75 = 8 + 1 + \frac{1}{2} + \frac{1}{4}$$

$$= 2^{3} + 2^{0} + 2^{-1} + 2^{-2}$$

$$= 2^{3} \left(2^{0} + 2^{-3} + 2^{-4} + 2^{-5}\right)$$

$$= \left(2^{11} \left(1 + 0.001 + 0.0001 + 0.00001\right)\right)_{2}$$

$$= \left(2^{11} \left(1.00111\right)\right)_{2}$$

Definition 1 (Double precision floating point representation). A floating point representation which uses 64 bits for representation of a number is called a double precision floating point representation.

The standard form of double precision representation is

$$a = \underbrace{\pm}_{1 \text{ bit } 1 \text{ bit}} \underbrace{1}_{52 \text{ bits}} \times w^{\frac{\pm}{1 \text{ bit } 10 \text{ bits}}}$$

Theorem 1 (Range of double precision floating point representation). The largest number which can be represented with double precision floating point representation is approximately 10^{307} and the smallest number which can be represented is approximately 10^{-307} .

Proof. As the exponent has 10 bits for representation,

$$-(10^{10}-1) \le \text{exponent} \le (10^{10}-1)$$

Therefore,

$$-1023 \le \text{exponent} \le 1023$$

Therefore, the smallest number, in terms of absolute value, which can be represented, is

$$1.\underbrace{0\cdots0}_{52 \text{ bits}} \times 2^{-1024} \approx 10^{-307}$$

Therefore, the smallest number which can be represented is approximately 10^{-307} , and the largest number which can be represented is approximately 10^{307} .

Definition 2 (Overflow). If a result is larger than the largest number which can be represented, it is called overflow.

Definition 3 (Underflow). If a result is smaller than the smallest number which can be represented, it is called underflow.

Definition 4 (Least significant digit).

$$1 = 1.\underbrace{0 \cdots 0}_{52 \text{ zeros}} \times 2^0$$

Let 1_{ε} be the smallest number larger than 1, which can be represented in double precision floating point representation. Therefore,

$$1 = 1.\underbrace{0 \cdots 0}_{51 \text{ zeros}} 1 \times 2^{0}$$
$$= 1 + 2^{-52}$$
$$\approx 1 + 2 \times 10^{-16}$$

Therefore,

$$1 - 1_{\varepsilon} = 2^{-52}$$
$$\approx 2 \times 10^{-16}$$

This number is called the least significant digit, or the machine precision. It is the maximum possible error in representation. It is represented by ε .

Definition 5 (Error). Let the DPFP representation of a number x be \tilde{x} . The absolute error in representation is defined as

absolute error =
$$|x - \tilde{x}|$$

= $0.0 \cdots 01 \times 2^{\text{exponent}}$

The relative error in representation is defined as

$$\delta = \frac{|x - \widetilde{x}|}{x}$$
$$= 0.0 \cdots 01$$
$$< \varepsilon$$

The maximum error, $2^{-52} \approx 2 \times 10^{-16}$, is called the machine precision. In general,

$$\widetilde{x} \star \widetilde{y} = (x \star y) (1 + \delta)$$

where δ is the relative error, ε is the machine precision, $\delta < \varepsilon$, and \star is an operator.

3.1 Loss of Significant Digits in Addition and Subtraction

Exercise 2.

Represent $\pi + \frac{1}{30}$ in base 10 with 4 digits.

Solution 2.

$$\pi \approx 3.14159$$

Approximating by ignoring the last digits,

$$\tilde{\pi} = 3.141$$

Similarly,

$$\frac{\widetilde{1}}{30} = 3.333 \times 10^{-2}$$

Therefore, adding,

$$\widetilde{\pi} + \frac{\widetilde{1}}{30} = 3.141 + 0.03333$$

$$= 3.174$$

Therefore,

$$\delta = \left| \frac{\left(\widetilde{\pi} + \widetilde{\frac{1}{30}}\right) - \left(\pi + \frac{1}{30}\right)}{\pi + \frac{1}{30}} \right|$$
$$= 0.0003$$

Therefore, $\delta < \varepsilon = 0.001$

Exercise 3.

Given

$$a = 1.435234$$

$$b = 1.429111$$

Find the relative error.

Solution 3.

$$a = 1.435234$$

$$b = 1.429111$$

Therefore,

$$a - b = 0.0061234$$

Approximating by ignoring the last digits,

$$\tilde{a}=1.435$$

$$\tilde{b} = 1.429$$

Therefore,

$$\tilde{a} - \tilde{b} = 0.006$$

Therefore,

$$\delta = \left| \frac{(a-b) - \left(\widetilde{a} - \widetilde{b}\right)}{a-b} \right|$$

Therefore,

$$\delta > 10^{-3}$$

$$\delta > \varepsilon$$

Exercise 4.

Solve

$$x^2 + 10^8 x + 1 = 0$$

Solution 4.

$$x = \frac{-10^8 \pm \sqrt{10^{16} - 4}}{2}$$

Therefore,

$$x_{-} \approx -10^{8}$$

Therefore, by Vietta Rules,

$$x_1 x_2 = \frac{c}{a}$$
$$x_1 + x_2 = -\frac{b}{a}$$

Therefore,

$$x_{+}x_{-} = 1$$

$$\therefore x_{+} = \frac{1}{x_{-}}$$

$$\approx -10^{-8}$$

In MATLAB, this can be executed as $x = \mathbf{roots}([1,10^8,1])$ This gives the result

$$x_{+} = -7.45 \times 10^{-9}$$

Therefore, the absolute error is

$$|\tilde{x} - x| = \left| -7.45 \times 10^{-9} - \left(-10^{-8} \right) \right|$$

= 2.55×10^{-9}

Therefore,

$$\delta = \left| \frac{\tilde{x} - x}{x} \right|$$

$$= \left| \frac{2.55 \times 10^{-9}}{10^{-8}} \right|$$

$$= 0.255$$

$$= 25\%$$

The algorithm used by MATLAB is

$$\begin{array}{l} \textbf{if} \ b \geq 0 \ \textbf{then} \\ x_1 = \frac{-b - \sqrt{b^2 - 4ac}}{2a} \\ x_2 = \frac{x}{ax_1} \\ \textbf{else} \\ x_2 = \frac{-b + \sqrt{b^2 - 4ac}}{2a} \\ x_1 = \frac{c}{ax_2} \\ \textbf{end if} \end{array}$$

This is done to avoid subtraction of numbers close to each other, and hence avoid the possible error.

4 Series of Approximations

4.1 Order of Convergence

Definition 6. Let $\{\alpha_n\}_{n=1}^{\infty}$ be a series. $\{\alpha_n\}$ is said to converge to α , denoted as $\alpha_n \to \alpha$, if $\forall \varepsilon > 0$, $\varepsilon \in \mathbb{R}$, $\exists n_0(\varepsilon) \in \mathbb{N}$, such that $\forall n \in \mathbb{N}$, $n > n_0(\varepsilon)$, $|\alpha_n - \alpha| < \varepsilon$.

Usually, the series $\{\alpha_n\}$ is compared to a simpler series such as $\frac{1}{n}, \frac{1}{n^{\beta}}, \dots$

Definition 7. α_n is said to be "big-O" of β_n , and is said to behave like β_n , if $\exists k \in \mathbb{R}, k > 0, \exists n_0 \in \mathbb{N}, n_0 > 0$, such that $\forall n > n_0$,

$$|\alpha_n| \le k|\beta_n|$$

It is denoted as

$$\alpha_n = \mathcal{O}(\beta_n)$$

Definition 8. α_n is said to be "small-O" of β_n if

$$\lim_{n \to \infty} \frac{\alpha_n}{\beta_n} = 0$$

It is denoted as

$$\alpha_n = \mathrm{o}(\beta_n)$$

Exercise 5.

Find the order of convergence of

$$\alpha_n = 2n^3 + 3n^2 + 4n + 5$$

Solution 5.

$$\alpha_n = 2n^3 + 3n^2 + 4n + 5$$

$$\leq (2+3+4+5)n^3$$

$$\therefore \alpha_n < 14n^3$$

Therefore, comparing to the standard form,

$$k = 14$$
$$\beta_n = n^3$$

Therefore, as $\forall n \geq 1, |a_n| \leq 14|\beta_n|$,

$$\alpha_n = \mathrm{O}(\beta_n)$$

Also,

$$\lim_{n \to \infty} \frac{\alpha_n}{\beta_n} = \lim_{n \to \infty} \frac{2n^3 + 2n^2 + 4n + 5}{n^3}$$
$$= 2$$

Therefore, as the limits is not zero,

$$\alpha_n \neq \mathrm{o}(\beta_n)$$

However, $\forall \delta > 0$,

$$\alpha_n = o\left(n^{3+\delta}\right)$$

5 Representation of Polynomials

5.1 Power series

Definition 9 (Power series representation of polynomials).

$$P_n(x) = a_0 + a_1 x + \dots + a_n x^n$$

This representation may lead to loss of significant digits.

Exercise 6.

Let P(x) represent a straight line.

$$P(6000) = \frac{1}{3}$$
$$P(6001) = -\frac{2}{3}$$

If only 5 decimal digits are used, show that there is a loss of significant digits, if the power series representation of the polynomial is used.

Solution 6.

P(x) represents a straight line. Therefore,

$$P(x) = ax + b$$

Therefore,

$$6000a + b = \frac{1}{3}$$
$$6001a + b = -\frac{2}{3}$$

Therefore,

$$\begin{pmatrix} 6000 & 1 \\ 6001 & 1 \end{pmatrix} \begin{pmatrix} a \\ b \end{pmatrix} = \begin{pmatrix} \frac{1}{3} \\ -\frac{2}{3} \end{pmatrix}$$

$$\therefore \begin{pmatrix} a \\ b \end{pmatrix}$$

$$= \frac{1}{|A|} \begin{pmatrix} 1 & -1 \\ -6001 & 6000 \end{pmatrix} \begin{pmatrix} \frac{1}{3} \\ -\frac{2}{3} \end{pmatrix}$$

$$= -\begin{pmatrix} 1 \\ -6000.3 \end{pmatrix}$$

$$= \begin{pmatrix} -1 \\ 6000.3 \end{pmatrix}$$

Therefore,

$$a = -1$$
$$b = 6000.3$$

Therefore,

$$P(x) = -x + 6000.3$$

Substituting 6000 and 6001 in this expression,

$$P(6000) = 0.3$$

 $P(6001) = 0.7$

However, the most accurate values of P(6000) and P(6001), using 5 decimal digits only, should be

$$P(6000) = 0.33333$$

 $P(6001) = -0.66666$

Therefore, there is a loss of significant digits.

5.2 Shifted Power Series

Definition 10 (Shifted power series representation of polynomials).

$$P_n(x) = a_0 + a_1(x - c) + \dots + a_n(x - c)^n$$

This representation is a power series shifted by c. Hence, this representation does not lead to loss of significant digits.

Exercise 7.

Let P(x) be a straight line.

$$P(6000) = \frac{1}{3}$$
$$P(6001) = -\frac{2}{3}$$

If only 5 decimal digits are used, show that there is no loss of significant digits, if the shifted power series representation of the polynomial is used, with c = 6000.

Solution 7.

P(x) represents a straight line. Therefore,

$$P(x) = a(x - 6000) + b$$

Therefore,

$$b = \frac{1}{3}$$

$$a + b = -0.66666$$

$$a = -0.99999$$

Therefore,

$$P(x) = -0.99999(x - 6000) + 0.33333$$

Substituting 6000 and 6001 in this expression,

$$P(6000) = 0.33333$$

 $P(6001) = -0.66666$

Therefore, there is no loss of significant digits, as the values of P(6000) and P(6001) are the most accurate values possible, using 5 decimal digits.

5.3 Newton's Form

Definition 11 (Newton's form of representation of polynomials).

$$P_n(x) = a_0 + a_1(x - c_1) + \dots + a_n(x - c_1) \dots (x - c_n)$$

The number of multiplications needed to calculate $P_n(x)$ is

$$\sum_{i=1}^{n} i = \frac{n(n+1)}{2}$$

The number of additions or subtractions needed to calculate $P_n(x)$ is

$$\sum_{i=1}^{n} i + n = \frac{n(n+1)}{2} + n$$

Therefore, the total number of operations needed to calculate $P_n(x)$ is $O(n^2)$.

5.4 Nested Newton's Form

Definition 12 (Nested Newton's form of representation of polynomials).

$$P_n(x) = a_0 + (x - c_1) \left(a_1 + (x - c_2) \left(a_2 + (x - c_3) \left(\dots \right) \right) \right)$$

The number of multiplications needed to calculate $P_n(x)$ is

$$\sum_{i=1}^{n} 1 = n$$

The number of additions or subtractions needed to calculate $P_n(x)$ is

$$\sum_{i=1}^{n} 2 = 2n$$

Therefore, the total number of operations needed to calculate $P_n(x)$ is big-O of O(n).

5.5 Properties of Polynomials

Theorem 2. For a polynomial in shifted power series form,

$$P_n(x) = P_n(c) + (x - c)q_{n-1}(x)$$

Proof.

$$P_n(x) = a_0 + a_1(x - c) + \dots + a_n(x - c)^n$$

$$= a_0 + (x - c) \left(a_1 + a_2(x - 2) + \dots + a_n(x - c)^{n-1} \right)$$

$$= a_0 + (x - c)q_{n-1}(x)$$

$$= P_n(c) + (x - c)q_{n-1}(x)$$

Theorem 3. If c is a root of $P_n(x)$, i.e., if

$$P_n(c) = 0$$

then

$$P_n(x) = (x - c)q_{n-1}(x)$$

If $c_1 \neq c_2$ are roots of $P_n(x)$, then

$$P_n(x) = (x - c_1)(x - c_2)r_{n-2}(x)$$

Similarly, if $P_n(x)$ has n different roots, then

$$P_n(x) = A(x - c_1) \dots (x - c_n)$$

where $A \in \mathbb{R}$.

If $P_n(x)$ has n+1 different roots, then

$$P_n(x) = A(x - c_1) \dots (x - c_n)(x - c_{n+1})$$

where A = 0.

Theorem 4. If p(x) and q(x) are polynomials of degree at most n, that satisfy

$$p(x_i) = f(x_i)$$

$$q(x_i) = f(x_i)$$

for $i \in \{0, \ldots, n\}$, then

$$p_n(x) \equiv q_n(x)$$

This means that there exists a unique polynomial with degree n which passes through n+1 points, i.e. n+1 points define a unique n degree polynomial.

Proof. Let

$$d_n(x) = p_n(x) - q_n(x)$$

Therefore, $d_n(x)$ is a polynomial of degree at most n, which has n+1 roots. Therefore,

$$d_n(x) \equiv 0$$

Therefore,

$$p_n(x) \equiv q_n(x)$$

6 Interpolation

Theorem 5 (Weierstrass Approximation Theorem). Let $f(x) \in C[a,b]$, i.e. it is continuous on [a,b]. Let $\varepsilon > 0$. Then there exists a polynomial P(x) defined on [a,b], such that $\forall x \in [a,b]$,

$$|f(x) - P(x)| < \varepsilon$$

Definition 13 (Interpolating polynomial). p(x) is said to be the interpolating polynomial of f(x), if for all sample points x_i ,

$$f(x_i) = p(x_i)$$

Theorem 6. Let f(x) such that $\forall i \in \{0, ..., n\}$,

$$f(x_i) = y_i$$

Then, there exists a unique polynomial p(x) of degree at most n, which interpolates f(x) at all sample points x_i .

6.1 Direct Method

Definition 14 (Van der Monde matrix). Let

$$p(x) = \sum_{i=0}^{n} a_i x^i$$

Let

$$f(x_i) = y_i$$

Therefore, as

$$p(x_i) = f(x_i)$$

the constraints are

$$a_0 + a_1 x_0 + \dots + a_n x_0^n = y_0$$

 $a_1 + a_1 x_1 + \dots + a_n x_1^n = y_1$
 \vdots
 $a_n + a_1 x_n + \dots + a_n x_n^n = y_n$

Therefore,

$$\begin{pmatrix} 1 & x_0 & x_0^2 & \dots & x_0^n \\ 1 & x_1 & x_1^2 & \dots & x_1^n \\ 1 & x_2 & x_2^2 & \dots & x_2^n \\ \vdots & \vdots & \vdots & & \vdots \\ 1 & x_n & x_n^2 & \dots & x_n^n \end{pmatrix} \begin{pmatrix} a_0 \\ a_1 \\ a_2 \\ \vdots \\ a_n \end{pmatrix} = \begin{pmatrix} y_0 \\ y_1 \\ y_2 \\ \vdots \\ y_n \end{pmatrix}$$

The matrix

$$V = \begin{pmatrix} 1 & x_0 & x_0^2 & \dots & x_0^n \\ 1 & x_1 & x_1^2 & \dots & x_1^n \\ 1 & x_2 & x_2^2 & \dots & x_2^n \\ \vdots & \vdots & \vdots & & \vdots \\ 1 & x_n & x_n^2 & \dots & x_n^n \end{pmatrix}$$

is called the Van der Monde matrix.

Theorem 7. The Van der Monde matrix is invertible, and hence there exists a unique matrix of coefficients a_0, \ldots, a_n , and hence the interpolating polynomial p(x) is unique.

6.2 Lagrange's Interpolation

Definition 15 (Lagrange polynomials). Let

$$L_k(x) = \prod_{i=0; i \neq k}^{n} (x - x_i)$$

Therefore,

$$L_k(x_i) = \begin{cases} 0 & ; & i \neq k \\ 1 & ; & i = k \end{cases}$$

Let

$$l_k(x) = \frac{L_k(x)}{L_k(x_k)}$$

Therefore,

$$l_k(x_i) = \begin{cases} 0 & ; & i \neq k \\ 1 & ; & i = k \end{cases}$$

The polynomials $l_i(x)$ are called Lagrange polynomials.

Theorem 8. Let

$$p_n(x) = \sum_{i=0}^n f(x_i)l_i(x)$$

where $l_i(x)$ are Lagrange polynomials. Then, $p_n(x)$ is the interpolating polynomial of f(x).

Exercise 8.

Which polynomial of degree 2 interpolates the below data?

$$\begin{array}{c|cc} x & f(x) \\ \hline 1 & 1 \\ 2 & 3 \\ 3 & 7 \\ \end{array}$$

Solution 8.

$$L_k(x) = \prod_{i=0: i \neq k}^{n} (x - x_i)$$

Therefore,

$$L_1(x) = (x-2)(x-3)$$

$$L_2(x) = (x-1)(x-3)$$

$$L_3(x) = (x-1)(x-2)$$

Therefore,

$$L_1(1) = (1-2)(1-3)$$

$$= 2$$

$$L_2(2) = (2-1)(2-3)$$

$$= -1$$

$$L_3(3) = (3-1)(3-2)$$

$$= 2$$

Therefore,

$$l_k(x) = \frac{L_k(x)}{L_k(x_k)}$$

Therefore,

$$l_1(x) = \frac{L_1(x)}{L_1(1)}$$

$$= \frac{1}{2}(x-2)(x-3)$$

$$l_2(x) = \frac{L_2(x)}{L_2(1)}$$

$$= -(x-1)(x-3)$$

$$l_3(x) = \frac{L_3(x)}{L_3(1)}$$

$$= \frac{1}{2}(x-1)(x-2)$$

Therefore,

$$p_2(x) = \sum_i f(x_i)l_i(x)$$

= $\frac{1}{2}(x-2)(x-3) - 3(x-1)(x-3) + \frac{7}{2}(x-1)(x-2)$

Exercise 9.

Given

$$k(z) = \int_{0}^{\frac{\pi}{2}} \frac{\mathrm{d}x}{\sqrt{1 - (\sin z)^{2} (\sin x)^{2}}}$$

and

$$k(1) = 1.5709$$

$$k(4) = 1.5727$$

k(6) = 1.5751

approximate k(3.5).

Solution 9.

$$l_k(x) = \frac{\prod_{i=0; i \neq k}^{n} (x - x_i)}{\prod_{i=0; i \neq k}^{n} (x_k - x_i)}$$

Therefore,

$$l_1(x) = \frac{(x-4)(x-6)}{(1-4)(1-6)}$$
$$l_4(x) = \frac{(x-1)(x-6)}{(4-1)(4-6)}$$
$$l_6(x) = \frac{(x-1)(x-4)}{(6-1)(6-4)}$$

Therefore,

$$l_1(3.5) = \frac{(3.5 - 4)(3.5 - 6)}{(1 - 4)(1 - 6)}$$

$$= 0.08333$$

$$l_4(3.5) = \frac{(3.5 - 1)(3.5 - 6)}{(4 - 1)(4 - 6)}$$

$$= 1.04167$$

$$l_6(3.5) = \frac{(3.5 - 1)(3.5 - 4)}{(6 - 1)(6 - 4)}$$

$$= -0.125$$

Therefore,

$$p_2(x) = \sum f(x_i)l_k(x)$$

$$\therefore p_2(3.5) = \sum f(x_1)l_k(3.5)$$

$$= (1.5709)(0.08333) + (1.5727)(1.04167) + (1.5751)(-0.125)$$

$$= 1.57225$$

6.3 Hermite Polynomials

Definition 16. Let the given data be of the form $(x_i, f(x_i), f'(x_i))$, where i = 0, ..., n.

 H_{2n+1} is called the Hermite polynomial of f(x).

For H_{2n+1} to be the interpolation polynomial of f(x), the constraints are

$$H_{2n+1}(x_i) = f(x_i)$$

 $H'_{2n+1}(x_i) = f'(x_i)$

Therefore, the number of constraints are 2n + 2. Hence, the polynomial is of degree at most 2n + 1.

Theorem 9. Let

$$H_{2n+1}(x) = \sum_{i=0}^{n} f(x_i)\psi_{n,i}(x) + \sum_{i=0}^{n} f'(x_i)\varphi_{n,i}(x)$$

Let

$$\delta_{ij} = \begin{cases} 0 & ; & i \neq j \\ 1 & ; & i = j \end{cases}$$

If the polynomials ψ and φ satisfy

$$\psi_{n,i}(x_j) = \delta_{ij}$$

$$\psi_{n,i}'(x_j) = 0$$

$$\varphi_{n,i}(x_j) = 0$$

$$\varphi'_{n,i}(x_j) = \delta_{ij}$$

then the polynomial H_{2n+1} is the interpolation polynomial of f(x).

6.4 Newton's Interpolation

Definition 17 (Newton's polynomial). The polynomial

$$p_n(x) = \sum_{i=0}^n A_i \prod_{j=0}^{i-1} (x - x_j)$$

is called Newton's polynomial.

Theorem 10. If $p_k(x)$, constructed based on x_1, \ldots, x_k is known, then $p_{k+1}(x)$, based on x_1, \ldots, x_{k+1} can be constructed as

 $\forall i = 0, \dots, k,$

 $(x_i - x_i) = 0.$ Therefore, if i = j,

 $(x_i - x_j) = 0.$ Therefore,

 $\prod (x_i - x_j) = 0$

$$p_{k+1}(x) = p_k(x) + A_{k+1}(x - x_0) \dots (x - x_k)$$

Proof. For $i = 0, \ldots, k$,

$$p_{k+1}(x_i) = p_k(x_i) + A_{k+1} \prod_{j=0}^{k} (x_i - x_j)$$
$$= p_k(x_i) + 0$$

For i = k + 1,

$$p_{k+1}(x_{k+1}) = p_k(x_{k+1}) + A_{k+1} \prod_{j=0}^{k} (x_{k+1} - x_j)$$
$$= f(x_{k+1})$$

where A_{k+1} can be calculated using $p_k(x_{k+1})$ and $f(x_{k+1})$. Therefore,

For n = 1,

$$p_0(x) = A_0$$
$$= f(x_0)$$

For n=2,

$$p_1(x) = p_0(x) + A_1(x - x_0)$$

= $f(x_0) - A_1(x - x_0)$
= $f(x_1)$

Therefore,

$$A_1 = \frac{f(x_1) - f(x_0)}{x_1 - x_0}$$
$$= f[x_0, x_1]$$

For n = 3,

$$p_2(x) = p_1(x) + A_2(x - x_0)(x - x_1)$$

$$= f(x_0) + f[x_0, x_1](x - x_0)$$

$$= f(x_0) + f[x_0, x_1](x - x_0) + A_2(x - x_0)(x - x_1)$$

$$= f(x_2)$$

Therefore,

$$A_2 = \frac{1}{(x_2 - x_0)(x_2 - x_1)} (f(x_2) - f(x_0) - f[x_0, x_1](x_2 - x_0))$$

= $f[x_0, x_1, x_2]$

and so on. In general,

$$A_k = f[x_0, \dots, x_k]$$

Definition 18 (Divided difference).

$$f[x_0, \dots, x_k] = \frac{f[x_1, \dots, x_k] - f[x_0, \dots, x_{k-1}]}{x_k - x_0}$$
$$f[x_0] = f(x_0)$$

is called the kth order divided difference of f(x).

Exercise 10.

Given

$$k(z) = \int_{0}^{\frac{\pi}{2}} \frac{\mathrm{d}x}{\sqrt{1 - (\sin z)^{2} (\sin x)^{2}}}$$

and

$$k(1) = 1.5709$$

$$k(4) = 1.5727$$

$$k(6) = 1.5751$$

approximate k(3.5).

Solution 10.

For the first order divided differences,

$$k[x_i] = k(x_i)$$

Therefore,

$$k[1] = k(1)$$

$$= 1.5709$$

$$k[4] = k(4)$$

$$= 1.5727$$

$$k[6] = k(6)$$

$$= 1.5751$$

For the second order divided differences,

$$k[x_i, x_j] = \frac{k[i] - k[j]}{i - i}$$

Therefore,

$$k[1,4] = \frac{k[1] - k[4]}{1 - 4}$$

$$= \frac{1.5727 - 1.5709}{3}$$

$$k[4,6] = \frac{k[4] - k[6]}{4 - 6}$$

$$= \frac{1.5751 - 1.5727}{2}$$

For the third order divided differences,

$$k[x_i, x_j, x_k] = \frac{k[i, j] - k[j, k]}{i - k}$$

Therefore,

$$k[1,4,6] = \frac{k[1,4] - k[4,6]}{1-6}$$

Hence,

$$A_0 = k[1]$$

 $A_1 = k[1, 4]$
 $A_2 = k[1, 4, 6]$