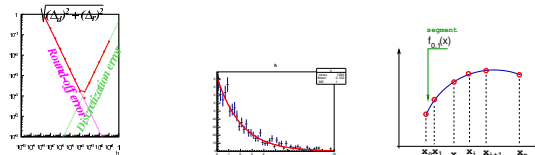




Computational Physics

numerical methods with C++ (and UNIX)



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Numerical methods

✓ System of linear equations

- ▶ Gauss elimination
- ▶ LU decomposition
- ▶ Gauss-Seidel method

✓ Interpolation

- ▶ Lagrange interpolation
- ▶ Newton method
- ▶ Neville method
- ▶ Cubic spline

✓ Numerical derivatives

- ▶ First derivative $O(h^2)$, $O(h^4)$
- ▶ Second derivative $O(h^2)$, $O(h^4)$
- ▶ Derivative by interpolation

✓ Numerical integration

- ▶ Newton-Cotes: trapezoidal and Simpson rules
- ▶ Gaussian quadrature

✓ Monte-Carlo methods



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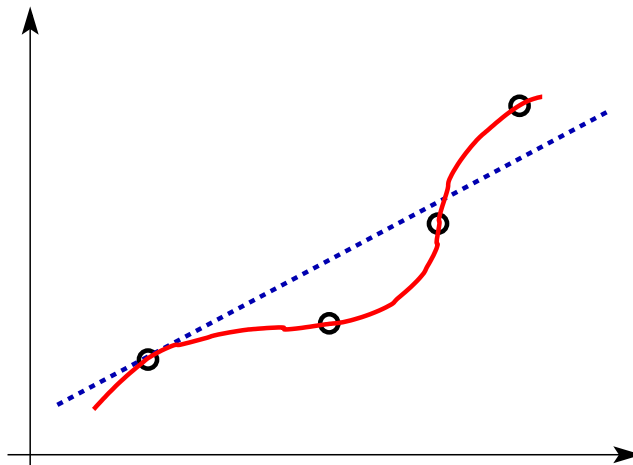
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✓ Monte-Carlo methods



Data interpolation

- ✓ Having a set of discrete data points (x_i, y_i) , **data interpolation** is the way of getting a continuous description passing through the data points





Lagrange interpolation

- ✓ Lagrange interpolation relies on the fact that in a finite interval a function $f(x)$ can always be represented by a polynomial $P(x)$
- ✓ **Linear interpolation:** polynomial of **degree one** passing through data points (x_1, y_1) and (x_2, y_2)

$$P(x) = P_0 + P_1 x$$

System to be solved:

$$\begin{cases} y_1 = P_0 + P_1 x_1 \\ y_2 = P_0 + P_1 x_2 \end{cases} \Rightarrow \begin{pmatrix} 1 & x_1 \\ 1 & x_2 \end{pmatrix} \begin{pmatrix} P_0 \\ P_1 \end{pmatrix} = \begin{pmatrix} y_1 \\ y_2 \end{pmatrix}$$

$$\begin{cases} P_1 = \frac{y_2 - y_1}{x_2 - x_1} \\ P_0 = y_2 - P_1 x_2 \end{cases} \quad P(x) = P_0 + P_1 x = y_1 \frac{x - x_2}{x_1 - x_2} + y_2 \frac{x - x_1}{x_2 - x_1}$$



Lagrange interpolation (cont.)

- ✓ **second-degree polynomial interpolation:** polynomial of **degree two** passing through data points (x_1, y_1) , (x_2, y_2) and (x_3, y_3)

$$P(x) = P_0 + P_1 x + P_2 x^2$$

System to be solved:

$$\begin{cases} y_1 = P_0 + P_1 x_1 + P_2 x_1^2 \\ y_2 = P_0 + P_1 x_2 + P_2 x_2^2 \\ y_3 = P_0 + P_1 x_3 + P_2 x_3^2 \end{cases} \Rightarrow \begin{pmatrix} 1 & x_1 & x_1^2 \\ 1 & x_2 & x_2^2 \\ 1 & x_3 & x_3^2 \end{pmatrix} \begin{pmatrix} P_0 \\ P_1 \\ P_2 \end{pmatrix} = \begin{pmatrix} y_1 \\ y_2 \\ y_3 \end{pmatrix}$$

$$P(x) = y_1 \frac{(x - x_2)(x - x_3)}{(x_1 - x_2)(x_1 - x_3)} + y_2 \frac{(x - x_1)(x - x_3)}{(x_2 - x_1)(x_2 - x_3)} + y_3 \frac{(x - x_1)(x - x_2)}{(x_3 - x_1)(x_3 - x_2)}$$



- $$P(x) = P_0 + P_1x + P_2x^2 + \cdots + P_nx^n$$

$$\begin{aligned} P_n(x) &= \sum_{i=0}^n y_i \ell_i(x) \\ &= y_0 \ell_0(x) + y_1 \ell_1(x) + \\ &\quad \cdots + y_n \ell_n(x) \end{aligned}$$

```
// n = polynomial degree

// n+1 = nb of data points

// x,y = abscissa and values

    double x[n+1], y[n+1];

// loop on data points (0...n)

    for (int i=0; i<n+1; i++) {

// we need a second loop for
// the product

        for (...) {

            }

        }

    }
```



```

-----
| class DataPoints |
-----
/ \      / \
|      |          ...
|      |          virtual double Interpolate(double x);
|      |          virtual void Draw();
|      |          virtual void Print();
|      |          protected:
|      |          int N; //nb data points
|      |          double *x, *y; //x and y values
|      |          };
|      |
|      |
|      |
|      |
|      |
|      -----
|      | class LagrangeInterpol |
|      -----
|
.
.
.
(other interpolation
classes)

class LagrangeInterpol : public DataPoints {
public:
    ...
    double Interpolate(double x);
    void Draw();
    void Print();
private:
    ? //specific data to class
};

```



Newton method

- ✓ The Newton method provides a better computational procedure to get an interpolating polynomial of degree n passing through $(n + 1)$ data points

$$x_i = x_0, x_1, \dots, x_n$$

$$y_i = y_0, y_1, \dots, y_n$$

$$a_i = a_0, a_1, \dots, a_n$$

$$P(x) = a_0 + a_1(x - x_0) + a_2(x - x_0)(x - x_1) + \dots + a_n(x - x_0)(x - x_1) \dots (x - x_{n-1})$$

- ✓ This polynomial can be written in an efficient computational way:

$$P(x) = a_0 + (x - x_0) [a_1 + (x - x_1) [a_2 + (x - x_2) [\dots [a_{n-1} + (x - x_{n-1})a_n] \dots]]$$

- ✓ The coefficients are determined by imposing the polynomial to pass through the data points:

$$(x_0, y_0) : y_0 = a_0$$

$$(x_1, y_1) : y_1 = a_0 + a_1(x_1 - x_0)$$

$$(x_2, y_2) : y_2 = a_0 + a_1(x_2 - x_0) + a_2(x_2 - x_0)(x_2 - x_1)$$

$$\vdots$$

$$(x_n, y_n) : y_n = a_0 + a_1(x_n - x_0) + \dots + a_n(x_n - x_0)(x_n - x_1) \dots (x_n - x_{n-1})$$



Newton method

- ✓ **Coefficients:**

$$a_0 = y_0$$

$$a_1 = \frac{y_1 - y_0}{x_1 - x_0} \equiv \nabla y_1$$

$$a_2 = \nabla^2 y_2$$

$$a_3 = \nabla^3 y_3$$

$$a_4 = \nabla^4 y_4$$

$$\vdots$$

$$a_n = \nabla^n y_n$$

Divided differences:

$$\nabla y_i = \frac{y_i - y_0}{x_i - x_0} \quad (i = 1, 2, \dots, n)$$

$$\nabla^2 y_i = \frac{\nabla y_i - \nabla y_1}{x_i - x_1} \quad (i = 2, 3, \dots, n)$$

$$\nabla^2 y_i = \frac{\nabla^2 y_i - \nabla^2 y_2}{x_i - x_2} \quad (i = 3, 4, \dots, n)$$

$$\vdots$$

$$\vdots$$

$$\nabla^n y_n = \frac{\nabla^{n-1} y_n - \nabla^{n-1} y_{n-1}}{x_n - x_{n-1}}$$

The diagonal terms of the table are the coefficients of the polynomial

	0th	1st	2nd	3rd	4th
x_0	y_0				
x_1	y_1	∇y_1			
x_2	y_2	∇y_2	$\nabla^2 y_2$		
x_3	y_3	∇y_3	$\nabla^2 y_3$	$\nabla^3 y_3$	
x_4	y_4	∇y_4	$\nabla^2 y_4$	$\nabla^3 y_4$	$\nabla^4 y_4$

Computing the interpolated value at x with the polynomial computed in a recursive way:

$$P_0(x) = a_n$$

$$P_1(x) = a_{n-1} + (x - x_{n-1})P_0(x)$$

$$P_2(x) = a_{n-2} + (x - x_{n-2})P_1(x)$$

$$\vdots$$

$$P_k(x) = a_{n-k} + (x - x_{n-k})P_{k-1}(x) \quad (k = 1, 2, \dots, n)$$



Newton method: algorithm

Coefficients:

```
// degree n polynomial
// n+1 data points
//
// For computing the coefficients
// we can use a one-dimensional
// array a[n+1]
//

1) make array a[n+1];

2) copy contents of Y[] data to array a[]

3) compute divided differences and
   store them in the one dimensional
   array a[]

   loop on k=1; k<n+1; k++

       loop on i=k; i<n+1; i++

           a[i] = (a[i] - a[k-1]) /
                 (x[i] - x[k-1])
```

Polynomial:

```
// degree n polynomial
// n+1 data points
//
// For computing the polynomial at
// a point x
// we use the recurrence existing
// after factorizing the polynomial
//
// We assume having already the
// coefficients
// computed in the array a[n+1]
//

1) init the last polynomial P

   P = a[n];

2) loop on k=1; k<n+1; k++

   P = a[n-k] + (x - x[n-k]) * P
```



Neville method

- ✓ The Neville algorithm is still better by computing standards for finding the n degree polynomial because does not require to a computation in two steps
- ✓ It uses linear interpolations between successive iterations: one point needed at 0th order, two points at 1st order, three points at 2nd order, ..., $n + 1$ points at n th order

0th order: $P_0[x_0] = y_0, \dots P_n[x_n] = y_n$

1st order (linear): $P_1[x_0, x_1] = C_0 + C_1 x = \frac{y_1(x-x_0)-y_0(x-x_1)}{x_1-x_0} = \frac{(x-x_0) P[x_1]-(x-x_1) P[x_0]}{x_1-x_0}$

2nd order: $P_2[x_0, x_1, x_2] = \frac{(x-x_2) P[x_0, x_1]-(x-x_0) P[x_1, x_2]}{x_0-x_2}$

3rd order: $P_3[x_0, x_1, x_2, x_3] = \frac{(x-x_3) P[x_0, x_1, x_2]-(x-x_0) P[x_1, x_2, x_3]}{x_0-x_3}$

...

...

x values	0th order	1st order	2nd order	3rd order	...order
x_0	$P_0(x_0) = y_0$				
x_1	$P_0(x_1) = y_1$	$P_1[x_0, x_1]$			
x_2	$P_0(x_2) = y_2$	$P_1[x_1, x_2]$	$P_2[x_0, x_1, x_2]$		
x_3	$P_0(x_3) = y_3$	$P_1[x_2, x_3]$	$P_2[x_1, x_2, x_3]$	$P_3[x_0, x_1, x_2, x_3]$	
x_4	$P_0(x_4) = y_4$	$P_1[x_3, x_4]$	$P_2[x_2, x_3, x_4]$	$P_3[x_1, x_2, x_3, x_4]$	
...		
x_n	$P_0(x_n) = y_n$	$P_1[x_{n-1}, x_n]$	$P_2[x_{n-2}, x_{n-1}, x_n]$	$P_3[x_{n-3}, x_{n-2}, x_{n-1}, x_n]$	



Neville method: algorithm?

- 1) We can try to work with only one array (1-dim) $y[]$ containing the 0th order polynomials passing by the values
- 2) loop on the order of the polynomials:
 $i=0, i < n+1$
- 3) loop on every column to compute the different polynomials
- 4) the interpolant calculated at the coordinate x , corresponds to the last value



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