

NEWTON'S INTERPOLATION

IDEA: Extend the natural basis:

$$1, \quad x - x_0, \quad (x - x_0)(x - x_1), \quad \dots, \\ \prod_{j=0}^{n-1} (x - x_j)$$

DEFINITION NEWTON'S INTERPOLATING
POLYNOMIAL

$$P_n(x) = a_0 + a_1(x - x_0) + \dots + a_n \prod_{j=0}^{n-1} (x - x_j)$$

The coefficients are chosen so that $P_n(x_i) = y_i$.

$$P(x_0) = y_0 \Rightarrow a_0 = y_0$$

$$P(x_1) = a_0 + a_1(x_1 - x_0) = y_1$$

$$\Rightarrow a_1 = \frac{y_1 - a_0}{x_1 - x_0}$$

Linear system (lower triangular)

$$\left(\begin{array}{cccccc} 1 & 0 & \dots & & & \\ 1 & x_1 - x_0 & 0 & \dots & & \\ 1 & x_2 - x_0 & (x_2 - x_1)(x_2 - 1) & 0 & \dots & \\ \vdots & & & & & \\ 1 & x_n - x_0 & \dots & & & \\ & & & & & \end{array} \right) \quad \left. \begin{array}{c} 0 \\ \prod_{j=0}^{n-1} (x_n - x_j) \end{array} \right\}$$

DIVIDED DIFFERENCES

Newton: $1, x - x_0, (x - x_0)(x - x_1), \dots, \dots$

$$p(x) = a_0 + a_1(x - x_0) + a_2(x - x_0)(x - x_1) + \dots + a_n \frac{\prod_{j=0}^{n-1} (x - x_j)}{(x - x_0)(x - x_1) \dots (x - x_{n-1})}$$

Natural basis:

$$p(x) = \sum_{k=0}^n c_k x^k; \text{ Connection: } c_n = a_n$$

Definition Divided difference of order k

$$f[x_0, x_1, \dots, x_k] = a_k$$

THEOREM

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

Notice: The recursion terminates,
because

$$f[x_i] = y_i = f_i$$

EXAMPLE $(1, 2), (2, 3), (3, 6)$

$$P_2(x) = x^2 - 2x + 3$$

Newton : $a_0 = \underline{\underline{2}}, a_1 = \underline{\underline{\underline{1}}}, a_2 = \underline{\underline{\underline{\underline{1}}}}$

$$f[x_0] = \underline{\underline{2}}$$

$$f[x_0] = \underline{\underline{3}} \quad f[x_0, x_1] = \frac{3 - 2}{2 - 1} = \underline{\underline{\underline{1}}}$$

$$f[x_1] = \underline{\underline{6}} \quad f[x_1, x_2] = \frac{6 - 3}{3 - 2} = \underline{\underline{\underline{3}}} \quad f[x_0, x_1, x_2] = \frac{3 - 1}{3 - 1} = \underline{\underline{\underline{\underline{1}}}}$$

We get the Newton coefficients !

Why does this work?

One point : $f[x_j] = f_j = y_j$

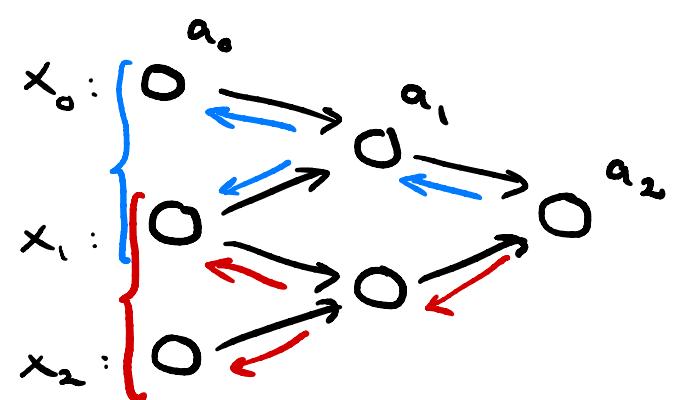
Two points : $f[x_i, x_j] = \frac{f[x_j] - f[x_i]}{x_j - x_i}$

Line spanning two points :

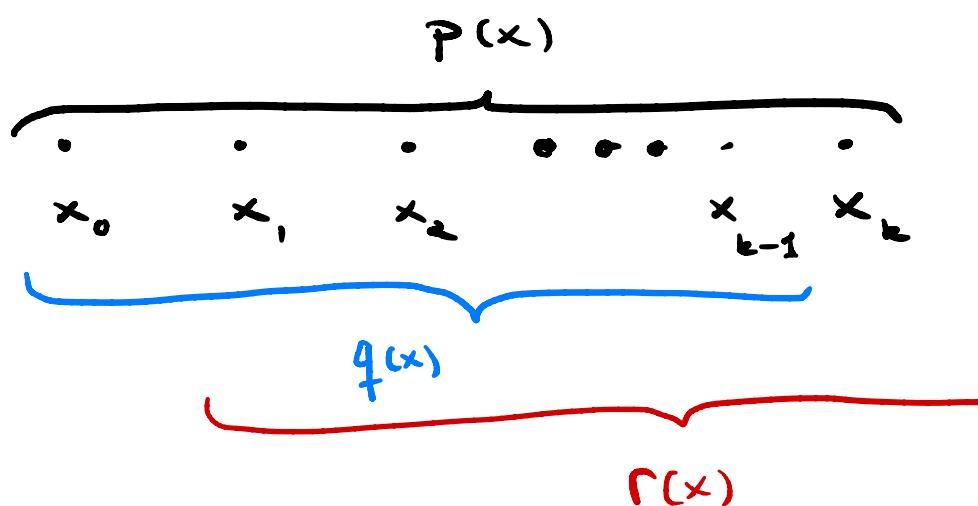
$$y - y_0 = \underbrace{\frac{y_1 - y_0}{x_1 - x_0}}_{(x - x_0)} (x - x_0)$$

Proof

Three interpolating polynomials:



$$\begin{array}{ll} p(x) & \text{degree } k \\ q(x) & " \quad k-1 \\ r(x) & " \quad k-1 \end{array}$$



$$\text{Claim: } p(x) = q(x) + \frac{x - x_0}{x_k - x_0} \underbrace{(r(x) - q(x))}_{=0}$$

$$x_0: p(x_0) = q(x_0) = f_0$$

$$x_1, \dots, x_{k-1}: p(x_i) = q(x_i)$$

$$x_k: p(x_k) = \underline{q(x_k)} + \frac{\underline{x_k - x_0}}{\underline{x_k - x_0}} \underbrace{(r(x_k) - q(x_k))}_{=1} = r(x_k)$$

The highest order term has coeff : $\frac{r(x) - q(x)}{x_k - x_0}$ \square

INTERPOLATION ERROR : $R(x) = f(x) - p(x)$

Assumptions : $f \in C^{n+1}$ ($n+1$ derivatives continuous)

Data : (x_i, y_i) ; $x' \neq x_i$

Auxiliary function :

$$h(x) = f(x) - p(x) - c w(x)$$

$$w(x) = \prod_{j=0}^n (x - x_j)$$

$$c = \frac{f(x') - p(x')}{w(x')}$$

Verify zeros : $x_i \rightarrow h(x_i) = 0$

$$x' : h(x') = f(x') - p(x') - \frac{(f(x') - p(x'))}{w(x')} = 0$$

There are at least $n+2$ zeros.

Rolle's Theorem : $h^{(n+1)}$ will have at least one zero

→ let's call this point ξ

$$h^{(n+1)}(x) = f^{(n+1)}(x) - \underbrace{p^{(n+1)}(x)}_{=0} - c w^{(n+1)}(x)$$

$$= f^{(n+1)}(x) - c(n+1)!$$

$$\Rightarrow h^{(n+1)}(\xi) = f^{(n+1)}(\xi) - c(n+1)! = 0$$

$$\Rightarrow c = \frac{f^{(n+1)}(\xi)}{(n+1)!}$$

Residual at x' :

$$R(x') = \frac{f^{(n+1)}(\xi)}{(n+1)!} \prod_{j=0}^n (x' - x_j)$$

THEOREM

$$R(x) = \frac{f^{(n+1)}(\xi)}{(n+1)!} \prod_{j=0}^n (x - x_j),$$

where $\xi = \xi(x)$.

Notice: Constant c is a divided difference:

$$f[x_0, x_1, \dots, x_n, x] = \frac{1}{(n+1)!} f^{(n+1)}(\xi(x))$$

Error amplification factor : $K(x)$

Measurements : "True value" f_i

"Wrong value" \hat{f}_i

$$p(x) - \hat{p}(x) = \sum_{i=0}^n (f_i - \hat{f}_i) L_i(x)$$

Assuming : $|f_i - \hat{f}_i| < \delta, i=0, \dots, n$

then

$$\begin{aligned} |p(x) - \hat{p}(x)| &\leq \delta \underbrace{\sum_{i=0}^n |L_i(x)|}_{\text{blue bracket}} \\ &= \delta K(x) \end{aligned}$$

$K(x)$ can be used to estimate the accumulated error in interpolation.

Evaluation of polynomials :

Horner's Rule : Complexity is $\Theta(n)$.

$$y = c_n ; y = yx + c_{n-1}$$

$$\Rightarrow y = \sum_{j=0}^n c_j x^j$$