

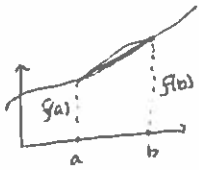
Numerical Analysis

08/29/2017

Discretization & error:

$$I(F) = \int_a^b f(x) dx \approx (b-a) f\left(\frac{a+b}{2}\right) = Q(F)$$

Quadrature Midpoint rule



$$\approx (b-a) \left(\frac{f(a) + f(b)}{2} \right) = \tilde{Q}(F)$$

Trapezoid rule

$$I(F) - \tilde{Q}(F) = \frac{-(b-a)^3}{12} f''(\xi)$$

Error

Notice in each case turn a continuous measurement into a final discrete value.

$$f'(x_0) \approx \frac{f(x_0+h) - f(x_0)}{h}$$

$$\approx \frac{f(x_0) - f(x_0-h)}{h}$$

$$\approx \frac{f(x_0+h/2) - f(x_0-h/2)}{h} = \frac{\delta_h f(x_0)}{h}$$

Notice all these discretization approximate the same thing but they have different 'costs' in terms of computation & error.

$$\left| f'(x_0) - \frac{\delta_h f(x_0)}{h} \right| \leq \frac{|f'''(\xi)| + |f'''(\eta)|}{2} \frac{h^2}{24}$$

where $\xi \in [x_0, x_0 + h/2]$, $\eta \in [x_0 - h/2, x_0]$

Note: \mathbb{R} & \mathbb{C} are not 'computer available'. Only floating point representations to these numbers.

IEEE Standard 754-1985 Binary Floating Point Arithmetic

$$x = \pm s \cdot 2^E$$

\leftarrow exponent
 \leftarrow base, radix
 significance (mantissa)
 $1 \leq s < 2$
 $s = 1 + f$
 $0 \leq f < 1; f = \sum_{i=1}^t f_i 2^{-i}$
 $f_i \in \{0, 1\}$
 binary rep.
 $f = (0.f_1 \dots f_t)_2$
 $E = e - b$
 \leftarrow bias
 $e = \sum_{i=0}^{t-1} e_i 2^i$
 $e_i \in \{0, 1\}$

$$x = (1 - 2^p) (1.f_1 \dots f_t) 2^E$$

p	e ₁	...	e ₀	f ₁	...	f _t
sign	exponent			sign.		

|| e's so e₀, ..., e₀

t bits in mantissa	52
8 bits in exponent	11
E _{max}	1023
E _{min}	-1022
b bias	1023

$$2 \cdot 2^t (E_{\max} - E_{\min} + 1) + 1 = \# \text{ 'j' in } \mathbb{F}$$

\leftarrow sign
 \leftarrow mant. / sig.
 \leftarrow exp
 \leftarrow 0 ie floating point numbers

Inc. exp gives larger range. Inc. mant. gives finer 'grading' of #'s.

Notice all these numbers are rational.

$$x_{\min} = 2^{E_{\min}}$$

$$x_{\max} = (2 - 2^{-t}) 2^{E_{\max}}$$

$$\Delta_E = 2^{E-t}$$

So the numbers not 'evenly' distributed

Rounding

$$fl: \mathbb{R} \rightarrow \mathbb{F}$$

rounding operation

For $x > 0$,

$$fl(x) = \begin{cases} x_-, & \text{if } x \in [x_-, \mu] \text{ or } x = \mu, a_t = 0 \\ x_+, & \text{if } x \in (\mu, x_+] \text{ or if } x = \mu \text{ and } b_t = 0 \end{cases}$$

x_{\pm} left/right neighbor (nearest), i.e.

$$x_+ = \min \{ y \in \mathbb{F} : y \geq x \}$$

$$x_+ = ((1.f_1 \dots f_t)_2 + (0.0 \dots 1)_2) 2^E$$

$$\mu = \frac{1}{2} (x_+ + x_-)$$

08/31/2017

Floating point arithmetic

$$\mathbb{F}(2, t, E_{min}, E_{max})$$

base mant. exp.

$$x \in \mathbb{F} \text{ then } x = \pm s \cdot 2^E$$

$$s = 1 + F; \quad F = (0.f_1 \dots f_t)_2$$

$$s = (1.f_1 \dots f_t)_2$$

p	e_{10}	\dots	e_0	f_1	\dots	f_{t2}
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$$x = \pm s \cdot 2^E = (1 - 2^p)(1.f_1 \dots f_t)_2 2^E$$

$$e = E + b$$

$$t = 52$$

$$p = 11$$

$$E_{max} = 1023$$

$$E_{min} = -1022$$

$$b = 1023$$

Largest $e_{max} = 2^l - 1$

Smallest e_{min}

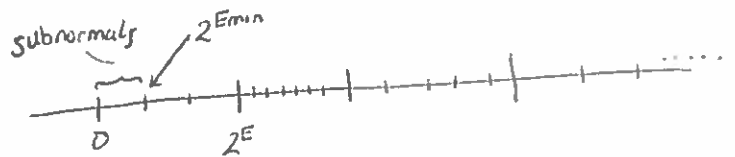
We reserve value (max)

$$E = e_{max} - b \text{ and min value}$$

$$E = e_{min} - b = -b$$

Exponent	Frac.	Fl #
$E = E_{min} - 1$	$F = 0$	± 0
$E = E_{min}$ with	$F \neq 0$	$\pm (0.F)_2 2^{E_{min}}$
$E_{min} < E < E_{max}$	F	$\pm (1.F)_2 2^E$
$E = E_{max} + 1$	$F = 0$	$\pm \infty$
$E = E_{max} + 1$	$F \neq 0$	NaN not a number

Subnormal



$$x \in [2^E, 2^{E+1}) \cap \mathbb{F}$$

$$x = s 2^E$$

Increment between consecutive fl-#'s in $[2^E, 2^{E+1})$ is $\Delta_E = 2^{E-t}$

The increment doubles from $[2^E, 2^{E+1})$ to $[2^{E+1}, 2^{E+2})$

There are 2^t fl-numbers in each $[2^E, 2^{E+1})$

subnormal (not of form $(1-2^p)(1.f_1 \dots f_t)_2 2^E$)

$$x = \pm f 2^{E_{min}}$$

The gap between subnormals is $\Delta = 2^{E_{min} - t}$

Rounding

$$f_l: \mathbb{R} \rightarrow \mathbb{F}$$

$$f_l(x)$$

Certainly want $f_l(f_l(x)) = f_l(x)$
 and also $x < y \rightarrow f_l(x) < f_l(y)$
 and $f_l(x) = x$ for $x \in \mathbb{F}$
 and $f_l(-x) = -f_l(x)$

Suppose $x \in [2^E, 2^{E+1})$

$$x = (1.f_1 \dots f_t)_2 \cdot 2^E$$

possibly infinite. So $x \in \mathbb{R}$

$$x_- = \max \{y \in \mathbb{F} : y \leq x\} = (1.a_1 \dots a_t)_2 \cdot 2^E$$

$$x_+ = \min \{y \in \mathbb{F} : x \leq y\} = (1.b_1 \dots b_t)_2 \cdot 2^E$$

If $x \notin \mathbb{F}$ then one of f_{t+1}, \dots is not zero.

$$x_+ = ((1.f_1 \dots f_t)_2 + (0.0 \dots 1)_2) \cdot 2^E$$

$$x_- = x_+ - 2^{E-t}$$

$$\mu = \frac{1}{2} (x_+ + x_-)$$

$$f_l(x) = \begin{cases} x_- & \text{if } x \in [x_-, \mu) \text{ or } x = \mu \text{ and } a_t = 0 \\ x_+ & \text{if } x \in (\mu, x_+] \text{ or } x = \mu \text{ and } b_t = 0 \end{cases}$$

Relative Error: $x \in [2^E, 2^{E+1})$; $\Delta_E = 2^{E-t}$ gap

$$\frac{|f_l(x) - x|}{|x|} \leq \frac{\frac{1}{2} 2^{E-t}}{2^E} = \frac{1}{2} \cdot 2^{-t} \text{ epsilon, eps}$$

But $(-)_2$ rep. do not have to be 'nice', even for 'nice' numbers:

$$\frac{1}{10} = 0.10 = (0.000110011 \dots)_2$$

$$1, 2^{-53} \in \mathbb{F} \text{ but } 1 + 2^{-53} \notin \mathbb{F}$$

So floating point operations are very complicated.

$x, y \in \mathbb{F}$; $\odot \in \{+, -, \cdot, \div\}$
 denote \boxplus to be \odot but in \mathbb{F} .

$$x \boxplus y = f_l(x \odot y)$$

So the operations in \mathbb{F} - though ugly - work as well as one would hope.

$$f_l(x \odot y) = (x \odot y)(1 + \delta)$$

$$|\delta| \leq \text{eps}$$

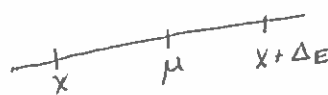
Add & mult. in \mathbb{F} is not associative
 Distributivity does not hold.
 Mult. and div. are not inverse operations.
 Even worse, let h be 'very small'
 $b \boxplus h = b$ "absorption"

Then

$$f_l'(b) := \frac{f_l(b \boxplus h) - f_l(b)}{h} = 0$$

Lemma: (Absorption) $x, y \in \mathbb{F}$, $0 < y < x$
 If $y < \frac{1}{4} 2^{-t} x$ then $f_l(x+y) = x$

PF: $x = s 2^E$; $1 \leq s < 2$
 the next largest \mathbb{F} number larger than x is $x + \Delta_E = s 2^E + 2^{E-t}$



$$\mu = x + \frac{1}{2} 2^{E-t}$$

$$x+y < x + \frac{1}{4} 2^{-t} x < \mu \text{ so } f_l(x+y) = x.$$

09/05/2017

Computational consequences of using
FI arithmetic.

Ex: Consider two mathematically
equivalent algorithms.

$$A_1(a,b) = a^2 - b^2$$

$$A_2(a,b) = (a-b)(a+b)$$

$$\begin{aligned} FI(a^2 - b^2) &= (a^2(1+\epsilon_1) - b^2(1+\epsilon_2))(1+\epsilon_3) \\ &= (a^2 - b^2) \left(\frac{a^2(1+\epsilon_1) - b^2(1+\epsilon_2)}{a^2 - b^2} \right) (1+\epsilon_3) \\ &= (a^2 - b^2) \left(1 + \frac{a^2\epsilon_1 - b^2\epsilon_2}{a^2 - b^2} \right) (1+\epsilon_3) \end{aligned}$$

$|\epsilon_i| \leq \epsilon$. Say $a \approx b$ and $\text{sign } \epsilon_1 = -\text{sign } \epsilon_2$
Then the result is 'bad'.

$$\begin{aligned} FI((a-b)(a+b)) &= ((a-b)(1+\epsilon_1)(a+b)(1+\epsilon_2))(1+\epsilon_3) \\ &= (a^2 - b^2) \underbrace{(1+\epsilon_1)(1+\epsilon_2)(1+\epsilon_3)}_{1+\delta} \end{aligned}$$

$$\text{So } 1+\delta = \epsilon_1 + \epsilon_2 + \epsilon_3 + \epsilon_1\epsilon_2 + \epsilon_1\epsilon_3 + \epsilon_2\epsilon_3$$

$$|\delta| = 3\epsilon + \mathcal{O}(\epsilon^2)$$

So the second choice is clearly better. So in
an algorithm one may see A_2 but ~~not~~ A_1 despite
being the same algebraically.

Ex: Preventing overflow.

$$x = \sqrt{a^2 + b^2}; \quad a = 10^{70}, b = 1$$

$$x = s \sqrt{(a/s)^2 + (b/s)^2}$$

$$\text{with } s = \max(|a|, |b|)$$

$$x = 10^{70} \sqrt{1^2 + (1/10^{70})^2}$$

This version will result in
a number whereas the
first will not.

Ex:

$$1 - \sqrt{1-x} = \frac{x}{1 + \sqrt{1-x}}$$

$$\text{Ex: } 1 - \cos x = 2 \sin^2\left(\frac{x}{2}\right)$$

Polynomial Computations

$$P(x) = a_0 + a_1x + a_2x^2 + \dots + a_nx^n$$

It takes $1, 2, \dots, n-1$ mult. to
compute x^2, x^3, \dots, x^n . Then

$n-1$ mult. for $a_i x^i$, then addition ^{$\hat{n}+1$ tot}
at least computing traditionally.

Horner's scheme:

$$P(x) = (\dots((a_nx - a_{n-1})x + a_{n-2})x + \dots) + a_0$$

This involves far less multiplication

$$b_n = a_n$$

$$b_{n-1} = a_nx + a_{n-1}$$

\vdots

$$b_i = b_{i+1}x + a_i$$

\vdots

$$b_0 = b_1x + a_0$$

The b_i are poly. in x . n mult. &
 n additions

$$P(y) = a_n y^n + a_{n-1} y^{n-1} + \dots + a_1 y = (b_n y^{n-1} + \dots + b_2 y + b_1)(y-x) + b_0$$

Comparing coefficients

$$\begin{aligned} a_n &= b_n \\ a_{n-1} &= -b_n x + b_{n-1} \\ &\vdots \\ a_1 &= -b_2 x + b_1 \\ a_0 &= -b_1 x + b_0 \end{aligned}$$

This gives division.

Notice also...

$$\begin{aligned} b'_{n-1}(x) &= a_n \quad (b'_n = 0) \\ &\vdots \\ b'_i(x) &= \frac{d}{dx} (b_{i+1}x + a_i) \\ &= b'_{i+1}(x) \cdot x + b_{i+1} \quad ; i = n-1, \dots, 0 \end{aligned}$$

$$b'_0(x) = p'(x)$$

So this gives derivatives. Back to division: div. by $x-\alpha$

$$P(x) = Q_\alpha(x)(x-\alpha) + r$$

α	a_n	a_{n-1}	\dots	a_1	a_0
	b_n	b_{n-1}	\dots		

Number in second line is sum of the number above it and number on left mult. by α .

Ex: $P(x) = x^3 - 4x^2 + 3x + 2$

$P(3) = ?$

	1	-4	3	2
3		3	-3	0
	1	-1	0	2

} algorithmic rep. of process at top left of this page

So $P(3) = 2$ and $P(x) = (x^2 - x)(x-3) + 2$

See p. 112

Now say we want to rep. $P(x)$ as...

$$P(x) = \sum_{i=1}^n a_i(x)(x-\alpha)^i$$

Given $P(x)$, find $a_i(\alpha)$.

$$P(x) = Q_{n-1}(x)(x-\alpha) + P(\alpha)$$

Denote $Q_n(\alpha) = P(\alpha)$.

$$Q_{n-1}(x) = Q_{n-2}(x)(x-\alpha) + Q_{n-1}(\alpha)$$

Then...

$$Q_n(x) = Q_0(\alpha)(x-\alpha)^n + Q_1(\alpha)(x-\alpha)^{n-1} + \dots + Q_{n-1}(\alpha)(x-\alpha) + Q_n(\alpha)$$

Hence, $a_i(\alpha) = Q_{n-i}(\alpha)$.

$$P(x) = x^3 - 4x^2 + 3x + 2$$

	1	-4	3	2
3		3	-3	0
	1	-1	0	2
3		3	-6	
	1	-2	-6	
3		3		
	1	-1		

Lagrange Interpolation

Thm: Given distinct x_0, \dots, x_n ($n \geq 0$) and arbitrary numbers f_0, \dots, f_n , there exists a unique polynomial L of degree $\leq n$ such that

$$L(x_i) = f_i ; i=0, \dots, n$$

if $f_i = f(x_i)$.

Cardinal Lag interp. poly. l_i

$l_i(x_j) = \delta_{i,j}$, the Kronecker function.

$$l_i(x) = \frac{(x-x_0) \dots (x-x_{i-1})(x-x_{i+1}) \dots (x-x_n)}{(x_i-x_0) \dots (x_i-x_{i-1})(x_i-x_{i+1}) \dots (x_i-x_n)}$$

Then let $L(x) = \sum_{i=0}^n f_i l_i(x)$

$$\text{Now } L(x_k) = \sum_{i=0}^n f_i l_i(x_k) = f_k$$

This is existence. For uniqueness, suppose $\tilde{L}(x)$ is another such polynomial.

Then $L(x) - \tilde{L}(x)$ is a poly. with at least $n+1$ roots, namely the x_i . So it is identically 0 and uniqueness is satisfied.

$x_i = \text{nodes}$

Notice the polynomial 'depends' on the nodes, in that for numerical accuracy, if the nodes are all very close, this could be an issue.

Suppose

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

$$L(x_i) = f_i$$

Find a_0, \dots, a_n . Well

$$a_0 + a_1 x_i + \dots + a_n x_i^n = f_i$$

So we can write:

$$\begin{pmatrix} 1 & x_0 & x_0^2 & \dots & x_0^n \\ 1 & x_1 & x_1^2 & \dots & x_1^n \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & x_n & x_n^2 & \dots & x_n^n \end{pmatrix} \begin{pmatrix} a_0 \\ a_1 \\ \vdots \\ a_n \end{pmatrix} = \begin{pmatrix} f_0 \\ f_1 \\ \vdots \\ f_n \end{pmatrix}$$

Vandermonde matrix, V

This matrix is often ill conditioned.

So finding the a_i with this are often inaccurate.

So we may want another way of determining the polynomial.

Newton basis:

$$\pi_0(x) = 1 \quad \text{deg } 0$$

$$\pi_1(x) = x - x_0 \quad \text{deg } 1$$

$$\pi_2(x) = (x - x_0)(x - x_1) \quad \text{deg } 2$$

\vdots

$$\pi_n(x) = (x - x_0) \dots (x - x_{n-1}) \quad \text{deg } n$$

We want to express L in Newton's basis.

$$L_{0, \dots, k-1} \in \mathbb{P}_{k-1} \quad \left. \begin{array}{l} \text{poly at most} \\ \text{deg } k-1 \end{array} \right\}$$

$$L_{0, \dots, k} \in \mathbb{P}_k$$

$$L_{0, \dots, k-1}(x_i) = f_i ; i=0, \dots, k-1$$

$$L_{0, \dots, k}(x_i) = f_i ; i=0, \dots, k-1, k$$

$L_{0, \dots, k} - L_{0, \dots, k-1}$ vanishes at x_0, \dots, x_{k-1} . In particular, $x - x_i$ is a factor of the diff for $i = 0, \dots, k-1$. But then it is a multiple of $\pi_k(x)$.

$$L_{0, \dots, k} - L_{0, \dots, k-1} = b_k \pi_k(x)$$

Clearly, b_k (leading coefficient of $L_{0, \dots, k} - L_{0, \dots, k-1}$) = leading coefficient of $L_{0, \dots, k}$.

$$L(x) = \sum_0^k f_i l_k(x) = \sum_0^k f_i \frac{\prod_0^k (x - x_j)}{\prod_0^k (x_i - x_j)}$$

So...

$$b_k = \sum_0^k \frac{f_i}{\prod_{j \neq i} (x_i - x_j)}$$

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Newton's Form

$$\begin{aligned} \pi_0(x) &= 1 \\ \pi_1(x) &= x - x_0 \\ &\vdots \\ \pi_n(x) &= (x - x_0) \dots (x - x_{n-1}) \end{aligned}$$

We would like to express $L_{0, \dots, n}(x)$ in terms of π_k .

$$\begin{aligned} L_{0, \dots, k-1}(x_i) &= f_i; \quad i = 0, \dots, k-1 \\ L_{0, \dots, k}(x_i) &= f_i; \quad i = 0, \dots, k-1, k \end{aligned}$$

$$(L_{0, \dots, k} - L_{0, \dots, k-1})(x_i) = 0; \quad i = 0, \dots, k-1$$

$$\therefore (L_{0, \dots, k} - L_{0, \dots, k-1})(x) = b_k \pi_k(x)$$

$$L_{0, \dots, k}(x) = \sum_0^k f_i \frac{\prod_{j=0}^k (x - x_j)}{\prod_{j \neq i} (x_i - x_j)}$$

$$b_k = \sum_0^k \frac{f_i}{\prod_{j \neq i} (x_i - x_j)}$$

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

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

$$L_0(x) = f(x_0)$$

$$L_{0, \dots, k}(x) = L_{0, \dots, k-1}(x) + f[x_0, \dots, x_k] \pi_k(x)$$

$$L_{0, \dots, n}(x) = f(x_0) + f[x_0, x_1](x - x_0) + \dots + f[x_0, \dots, x_n](x - x_0) \dots (x - x_{n-1})$$

$$L_{0, \dots, n}(x_i) = \sum_0^n b_k \pi_k(x_i) = f_i$$

$$\begin{pmatrix} 1 & 0 & \dots & 0 \\ \pi_1(x_0) & \dots & \dots & 0 \\ \vdots & \dots & \dots & \vdots \\ \pi_n(x_0) & \dots & \pi_n(x_0) & \dots \end{pmatrix} \begin{pmatrix} b_0 \\ b_1 \\ \vdots \\ b_n \end{pmatrix} = \begin{pmatrix} f_0 \\ f_1 \\ \vdots \\ f_n \end{pmatrix}$$

Properties of divided differences

a) Linearity: $f(x) = \alpha g(x) + \beta h(x)$
 $f[x_0, \dots, x_k] = \alpha g[x_0, \dots, x_k] + \beta h[x_0, \dots, x_k]$

b) Commutativity
 $f[x_0, \dots, x_k] = f[x_{\sigma(1)}, \dots, x_{\sigma(n)}]$

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

Define $q \in \mathbb{P}_k$

$$q(x) = \frac{(x-x_0)L_{1,\dots,k}(x) - (x-x_k)L_{0,\dots,k-1}(x)}{x_k - x_0}$$

q interpolates at x_0, \dots, x_k then

$$q \equiv L_{0,\dots,k}$$

Compare leading coefficients.

x_0	f_0		
x_1	f_1	$f[x_0, x_1]$	
x_2	f_2	$f[x_1, x_2]$	$f[x_0, x_1, x_2]$
\vdots	\vdots	\vdots	\vdots
x_n	f_n	$f[x_{n-1}, x_n]$	$f[x_0, \dots, x_n]$

The following identity holds:

$$f(x) = \sum_{k=0}^n f[x_0, \dots, x_k] \pi_k(x) + f[x_0, \dots, x_n, x] \pi_{n+1}(x)$$

Interpolation Remainder:

$$f \in C^{n+1}([a, b]); \{x_0, \dots, x_n\} \subset [a, b]$$

$$L_n = L_{0,\dots,n}$$

$$f(x) - L_n(x) = \frac{f^{(n+1)}(\xi)}{(n+1)!} (x-x_0) \dots (x-x_n)$$

for $\xi \in (a, b)$

$$M_{n+1} = \sup_{0 < x < 1} |f^{(n+1)}(x)| \quad |E(x)| \leq \frac{M_{n+1}}{(n+1)! 2^n}$$

Let $q \in \mathbb{P}_{n+1}$ which interpolates f at x_0, \dots, x_n, x

$$q(t) = L(t) + \frac{f(x) - L_n(x)}{\pi_{n+1}(x)} \pi_{n+1}(t)$$



$$E(t) = f(t) - q(t)$$

$$E(y) = 0 \text{ for } y \in \{x_0, \dots, x_n, x\}$$

$$0 = E^{(n+1)}(\xi) = f^{(n+1)}(\xi) - \frac{f(x) - L_n(x)}{\pi_{n+1}(x)} (n+1)!$$

Remainder Est.

$$x_m = \frac{1}{2} \left((b-a) \cos\left(\frac{(2m+1)\pi}{2(n+1)}\right) + (b+a) \right)$$

$$m = 0, \dots, n$$

If $(a, b) = (-1, 1)$ such nodes of interp in $(-1, 1)$ would be roots of

$$T_{m+1}(x) = \cos((m+1) \arccos x)$$

$$x_m = \cos\left(\frac{(2m+1)\pi}{2(n+1)}\right); m = 0, 1, \dots, n$$

$$\pi_{n+1}(x) = (x-x_0) \dots (x-x_n) = 2^{-n} T_{n+1}(x)$$

$$\max_{-1 < x < 1} |(x-x_0) \dots (x-x_n)| = 2^{-n}$$

Hermite Interpolation

$x_0 \neq x_1$. Want $h_{0,1} \in \mathbb{P}_3$ with...

$$h_{0,1}(x_0) = 0 ; h'_{0,1}(x_0) = 1$$

$$h_{0,1}(x_1) = 0 ; h'_{0,1}(x_1) = 0$$

$$p(x) = (x-x_0)(x-x_1)^2$$

$$p'(x) = (x-x_1)^2 + 2(x-x_0)(x-x_1)$$

$$p'(x_0) = (x_0-x_1)^2$$

$$\therefore \frac{p(x)}{p'(x_0)} = h_{0,1}(x)$$

$$(x-x_0) \left(\frac{x-x_1}{x_0-x_1} \right)^2$$

Find $h_{0,0}(x)$

$$h_{0,0}(x_0) = 1 ; h'_{0,0}(x_0) = 0$$

$$h_{0,0}(x_1) = 0 ; h'_{0,0}(x_1) = 0$$

09/12/2017

$$L_{0,0}(x) = \left(\frac{x-x_1}{x_0-x_1} \right)^2$$

$$h_{0,0}(x) = L_{0,0}(x) - L'_{0,0}(x_0)h_{0,1}(x)$$

$$= \frac{(x-x_1)(2x+x_1-3x_0)}{(x_0-x_1)^3}$$

We can construct $h_{1,1}$ & $h_{1,0}$

$$h_{1,0}(x) = \frac{-(x-x_0)^2(2x+x_0-3x_1)}{(x_1-x_0)^3}$$

$$h_{1,1}(x) = (x-x_1) \left(\frac{x-x_0}{x_1-x_0} \right)^2$$

We may solve the 2pt Hermite interp problem. $H \in \mathbb{P}_3$

$$H^0(x_0) = f_0 ; H'(f_0) = f'(x_0)$$

$$H^0(x_1) = f_1 ; H'(f_1) = f'(x_1)$$

$$H(x) = f(x_0)h_{0,0}(x) + f'(x_0)h_{0,1}(x) + f(x_1)h_{1,0}(x) + f'(x_1)h_{1,1}(x)$$

Newton's form:

$$H(x) = f_0^{[x_0]} + f[x_0, 2](x-x_0) +$$

$$f[x_0, 2, x_1](x-x_0)^2 +$$

$$f[x_0, 2, x_1, 2](x-x_0)^2(x-x_1)$$

$$x_0 \quad f(x_0)$$

$$x_0 \quad f(x_0) \quad f'(x_0)$$

$$x_1 \quad f(x_1) \quad f[x_0, x_1] \quad f[x_0, 2, x_1]$$

$$x_1 \quad f(x_1) \quad f'(x_1) \quad f[x_1, x_1, 2] \quad f[x_0, 2, x_1, 2]$$

Thm: Let x_0, \dots, x_k be distinct &

m_0, \dots, m_k be integers ≥ 1 such that $\sum_{i=0}^k m_i = m+1$. Let f be $\exists f^{(m_i-1)}(x_i)$

exists for $0 \leq i \leq k$ exists, then $\exists!$

poly $h \in \mathbb{P}_m$ with $h^{(e)}(x_i) = f^{(e)}(x_i)$

$i=0, \dots, k ; 0 \leq e \leq m_i-1$

Our system has a unique solution if and only if a homog. system has a unique solution.

$$h^{(\ell)}(x_i) = 0 \quad i = 0, \dots, k \\ \ell = 0, \dots, m_i - 1$$

$$h(x) = h(x_i) + h'(x_i)(x-x_i) + \dots + \\ \frac{1}{(m_i-1)!} h^{(m_i-1)}(x_i)(x-x_i)^{m_i-1} + \\ \frac{1}{m_i!} h^{(m_i)}(x_i + \theta(x-x_i))(x-x_i)^{m_i}$$

$$h(x) = (x-x_i)^{m_i} g(x)$$

$$h(x) = \prod_{i=0}^k (x-x_i)^{m_i} g(x)$$

$$h(x) \equiv 0 \quad \} \quad n+1 \text{ zeros}$$

from counting zeros. $\int_0^1 q(x) = 0$
Then $h \equiv 0$, as 'correct'.

$$\text{Thm: } h(x) = \sum_{i=0}^k \sum_{\ell=0}^{m_i-1} f^{(\ell)}(x_i) h_{i,\ell}(x)$$

where $h_{i,\ell} \in \mathbb{P}_m$ satisfy...

$$h_{i,\ell}^{(m)}(x_j) = \begin{cases} 1, & i=j \text{ and } m=\ell \\ 0, & i \neq j \text{ or } m \neq \ell \end{cases}$$

Define poly:

$$L_{i,\ell}(x) = \frac{(x-x_i)^\ell}{\ell!} \prod_{\substack{j=0 \\ j \neq i}}^k \left(\frac{x-x_j}{x_i-x_j} \right)^{m_j}$$

$$i = 0, \dots, k; \quad \ell = 0, \dots, m_i - 1$$

$$\text{then } h_{i,m_i-1}(x) = L_{i,m_i-1}, \quad i = 0, \dots, k$$

$$h_{i,m}(x) = L_{i,m}(x) - \sum_{\nu=m+1}^{m_i-1} L_{i,m}^{(\nu)}(x_i) h_{i,\nu}(x)$$

$$m = m_i - 2, \dots, 0$$

Consider $L_{i,\ell}(x)$

$$L_{i,\ell}^{(0)}(x_i) = \dots = L_{i,\ell}^{(\ell-1)}(x_i) = 0$$

$$L_{i,\ell}^{(\ell)}(x_i) = 1$$

and at $x = x_j$

$$L_{i,\ell}^{(0)}(x_j) = \dots = L_{i,\ell}^{(m_j-1)}(x_j) = 0 \\ j \neq i$$

Formal definition of divided diff with multiple nodes:

$$a) \quad f[x_0, i] = \frac{f^{(i-1)}(x_0)}{(i-1)!}; \quad i \geq 1$$

$$b) \quad f[x_0, m_0; \dots; x_k, m_k] =$$

$$\frac{f[x_0, m_0-1; \dots; x_k, m_k] - f[x_0, m_0; \dots; x_k, m_k-1]}{x_k - x_0}$$

For any $i, 0 \leq i \leq k$, define...

$$s(i) = \begin{cases} 0, & i = 0 \\ m_0 + \dots + m_{i-1}; & 0 < i \leq k \end{cases}$$

Now every integer p ; $0 \leq p \leq m$
can be represented by

$$p = s(i) + j; \quad 0 \leq i \leq K \\ 0 \leq j \leq m_i - 1$$

$$\pi_0(x) = 1$$

$$\pi_1(x) = \pi_{s(0)+1}(x) = (x - x_0)$$

...

$$\pi_{m_0-1}(x) = \pi_{s(m_0-1)}(x) = (x - x_0)^{m_0-1}$$

$$\pi_{m_0}(x) = \pi_{s(1)+0}(x) = (x - x_0)^{m_0}$$

$$\pi_{s(i)+j}(x) = (x - x_0)^{m_0} \cdots (x - x_{i-1})^{m_i-1} (x - x_i)^j$$

$$0 \leq j \leq m_i - 1$$

$$h_m(x) = \sum_{p=0}^m b_p \pi_p(x) = \sum_{i=0}^K \sum_{j=0}^{m_i-1} b_{s(i)+j} \pi_{s(i)+j}(x)$$

$$b_{s(i)+j} = f[x_0, m_0; \dots; x_{i-1}, m_{i-1}; x_i, j+1]$$

09/14/2017

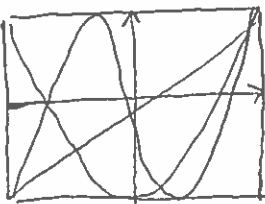
Chebyshev Polynomials

$$T_0(x) = 1$$

$$T_1(x) = x$$

...

$$T_{n+1}(x) = 2x T_n(x) - T_{n-1}(x)$$



This has many expressions

$$* T_n(x) = \cos(n \cos^{-1} x); \quad n \geq 0 \\ -1 < x < 1$$

OR

$$* T_n(x) = \frac{1}{2} (z^n + z^{-n});$$

$$|z| = 1$$

$$x = \operatorname{Re} z$$

$$T_n(x) = \frac{1}{2} \left((x + \sqrt{x^2 - 1})^n + (x - \sqrt{x^2 - 1})^n \right); \quad x \in \mathbb{R}$$

$$T_n(x) = \begin{cases} \cosh(n \operatorname{arccosh} x), & x \geq 1 \\ (-1)^n \cosh(n \operatorname{arccosh}(-x)), & x \leq -1 \end{cases}$$

For the first, use

$$\cos(A+B) = \cos A \cos B - \sin A \sin B$$

Change $n+1$ to $n-1$. Add & rearrange

then let $\theta = \cos^{-1} x$ & $x = \cos \theta$

Show it satisfies same recurrence relations.

$$T_n(t_{n,k}) = 0$$

$$t_{n,k} = \cos\left(\frac{2k-1}{2n} \pi\right); \quad k = 1, \dots, n$$

$$T_n(s_{n,k}) = (-1)^k$$

$$s_{n,k} = \cos\left(\frac{k\pi}{n}\right); \quad k = 0, \dots, n$$

↳ Chebyshev points

Complex Analytic Def:

$$T_n(x) = \cos(n\theta)$$

$$\theta = \arccos x$$

$$z = e^{i\theta}$$

$$\operatorname{Re} z = \frac{z + \bar{z}}{2} = \frac{e^{i\theta} + e^{-i\theta}}{2} = \cos \theta$$

$$\cos(n\theta) = \operatorname{Re}(z^n) = \frac{z^n + z^{-n}}{2} = \frac{z^n + \bar{z}^n}{2}$$

Complex points $\{z_j\}$, $(n+1)$ total



$$z_j = e^{j\frac{\pi}{n}i}; j=0, \dots, n$$

The proj. to x-axis is no longer uniform

$$\operatorname{Re} z_k = \cos\left(\frac{k\pi}{n}\right); k=0, \dots, n$$

Chebyshev points

Minimal Property of T_n

Thm: Let $p \in \mathbb{T}_n$ be monic
then $\|p\|_\infty = \max_{-1 \leq x \leq 1} |p(x)| \geq 2^{1-n}$

$$2^{1-n} = \max_{-1 \leq x \leq 1} \underbrace{|2^{1-n} T_n(x)|}_{\text{monic poly}} \leq \max_{-1 \leq x \leq 1} |p(x)|$$

Orthogonality of T_n

$$T_n(x); -1 < x < 1$$

Consider $L_w^2([-1, 1])$

$$w(x) = \frac{1}{\sqrt{1-x^2}} > 0$$

$$\|f\| = \int_{-1}^1 |f(x)|^2 w(x) dx$$

$$\langle f, g \rangle = \int_{-1}^1 f(x) g(x) w(x) dx$$

$$\langle T_n, T_m \rangle = \begin{cases} 0, & n \neq m \\ \pi & n = m = 0 \\ \frac{\pi}{2} & n = m \neq 0 \end{cases}$$

09/19/2017

Neville's algorithm of poly. interpolation

$$L_{0,1,\dots,n}(x)$$

$$L_{0,1,\dots,n}(x_i) = f_i; i=0, 1, \dots, n$$

$$L_{m_1, \dots, m_r, k}(x_{m_j}) = f_{m_j}; j=0, \dots, k$$

$$L_{0,1}(x) = \frac{x-x_1}{x_0-x_1} f_0 + \frac{x-x_0}{x_1-x_0} f_1$$

Objective:

$$\frac{x-x_1}{x_0-x_1} + \frac{x-x_0}{x_1-x_0} = 1$$

$$L_{1,2}(x) = \frac{x-x_2}{x_1-x_2} f_1 + \frac{x-x_1}{x_2-x_1} f_2$$

$$L_{0,1,2} = \frac{x-x_2}{x_0-x_2} L_{0,1}(x) + \frac{x-x_0}{x_2-x_0} L_{1,2}(x)$$

$$\frac{x_2-x}{x_2-x_0} + \frac{x-x_0}{x_2-x_0} = 1$$

$$L_{0,1,2}(x) = \frac{(x-x_0) L_{1,2}(x) - (x-x_2) L_{0,1}(x)}{x_2-x_0}$$

$$L_{m, \dots, m+k}(x) = \frac{x - x_{m+k}}{x_m - x_{m+k}} L_{m, \dots, m+k-1}(x) + \frac{x - x_m}{x_{m+k} - x_m} L_{m+1, \dots, m+k}(x)$$

For $m \geq 0$ and $n \geq m+k$

$$x_0 = f_0 = L_0(x)$$

$$x_1: f_1 = L_1(x) = L_{0,1}(x)$$

$$x_2: f_2 = L_2(x) = L_{1,2}(x) = L_{0,1,2}(x)$$

$$x_3: f_3 = L_3(x) = L_{2,3}(x) = L_{1,2,3}(x) = L_{0,1,2,3}(x)$$

Bernstein Polynomial

$$(B_n f)(x) = \sum_{k=0}^n f\left(\frac{k}{n}\right) g_{n,k}(x)$$

where $g_{n,k}(x) = \binom{n}{k} x^k (1-x)^{n-k}$

We can also write

$$B_{n,k}(x) = \binom{n}{k} x^k (1-x)^{n-k}$$

$$n=0: B_{0,0}(x) = 1$$

$$n=1: B_{1,0}(x) = 1-x$$

$$B_{1,1}(x) = x$$

$$n=2: B_{2,0}(x) = (1-x)^2$$

$$B_{2,1}(x) = 2x(1-x)$$

$$B_{2,2}(x) = x^2$$

1) $B_{n,k}$ has a zero of mult. $k \in x=0$ and one of mult. $n-k \in x=1$

2) $B_{n,k} \in \mathbb{P}_n$ are nonnegative on $[0,1]$

$$3) B_{n,k}(x) \geq 0 \text{ for } x \in [0,1]$$

$$4) \sum_{k=0}^n B_{n,k}(x) = 1 \text{ for } x \in \mathbb{R}$$

(partition of unity)

5) Symmetry:

$$B_{n,k}(x) = B_{n,n-k}(1-x); k=0, \dots, n$$

$$6) k=0, \dots, n$$

$$B_{n,k}(x) = x B_{n-1,k-1}(x) + (1-x) B_{n-1,k}(x)$$

PF:

$$4) \sum_{k=0}^n \binom{n}{k} x^k (1-x)^{n-k} = (x + (1-x))^n = 1$$

$$5) \binom{n}{k} = \binom{n}{n-k}$$

$$6) \binom{n}{k} = \binom{n-1}{k-1} + \binom{n-1}{k}$$

Max of $B_{n,k}$ occurs at $x = k/n$

$B_{n,0}, \dots, B_{n,n}$ are lin. independent:

$$\sum_{k=0}^n b_k B_{n,k}(x) = 0$$

$$0 = \sum_{k=0}^n b_k B_{n,k}(1) = b_n B_{n,n}(1) \rightarrow b_n = 0$$

= 0 except for $B_{n,n}(x) = 1$
ie when $k=n$

$B_{n,k}(x)$ div. by $1-x$. Divide by $1-x$

then $x=1 \rightarrow b_{n-1} = 0$ and

continue $b_0 = \dots = b_n = 0$

$$B_{n,k}(a,b) = \frac{1}{(b-a)^n} \binom{n}{k} (x-a)^k (b-x)^{n-k}$$

Bernstein Poly for f
 $f \in C[0,1]$.

(consider $B_n f$).

$$(B_n f)(x) = \sum_{k=0}^n f\left(\frac{k}{n}\right) \binom{n}{k} x^k (1-x)^{n-k}$$

Goal $B_n f \xrightarrow{unif} f$ on $[0,1]$

That is, $\|f - B_n(f)\|_\infty \rightarrow 0$

$$\|f\|_\infty = \sup_{0 \leq x \leq 1} |f(x)|$$

Think of an operator:

$$B_n : C[0,1] \rightarrow \mathcal{T}_n \subset C[0,1]$$

Bernstein op. are monotone:

$$f \leq g \Rightarrow B_n f \leq B_n g$$

Thm: (Bohman-Korovkin) If

L_n seq. monotone lin op. on $C[a,b]$

If $\|L_n f - f\|_\infty \rightarrow 0$ for $f = 1, x, x^2$

then true for all $f \in C[a,b]$.

OR

$H_n : C[a,b] \rightarrow C[a,b]$, monotone

$$\|H_n(e^x)(x) - x^r\| \rightarrow 0$$

for $r=0,1,2 \rightarrow$

$$\|H_n(f) - f\| \rightarrow 0$$

$$B_n(1)(x) = \sum_{k=0}^n \binom{n}{k} x^k (1-x)^{n-k} = 1$$

Modify partition of unity formula
 replace n with $n-1$, mult by x

$$\begin{aligned} nx &= \sum_{k=0}^{n-1} n \binom{n-1}{k} x^{k+1} (1-x)^{n-(k+1)} \\ &= \sum_{k=0}^{n-1} (k+1) \binom{n}{k+1} x^{k+1} (1-x)^{n-(k+1)} \\ &= \sum_{s=1}^n s \binom{n}{s} x^s (1-x)^{n-s} \end{aligned}$$

$$= \sum_{k=0}^n k \binom{n}{k} x^k (1-x)^{n-k}$$

Then $nx = \int_0^1$

$$x = \sum_{k=0}^n \frac{k}{n} \binom{n}{k} x^k (1-x)^{n-k}$$

Then $B_n(t^2)(x) \int$

$\int_0^1 B_n$ invariant on constants & lin. pieces. Using above calculation twice

$$\begin{aligned} B_n(t^2)(x) &= \sum_{k=0}^n \left(\frac{k}{n}\right)^2 \binom{n}{k} x^k (1-x)^{n-k} \\ &= \sum_{k=1}^n \left(\frac{k}{n}\right) \binom{n-1}{k-1} x^k (1-x)^{n-k} \\ &= \sum_{k=1}^n \left(\frac{n-1}{n} \frac{k-1}{n-1} + \frac{1}{n}\right) \binom{n-1}{k-1} x^k (1-x)^{n-k} \\ &= \frac{n-1}{n} x^2 \sum_{k=2}^n \binom{n-2}{k-2} x^{k-2} (1-x)^{n-k} + \frac{x}{n} \\ &= \frac{n-1}{n} x^2 + \frac{x}{n} \xrightarrow{unif} x^2 \end{aligned}$$

That is, use above with $n \rightarrow n-1$, mult. by nx and manipulate.

09/21/2017

Bernstein to Bézier Curves

$P_0, \dots, P_n \in \mathbb{R}^d$ control points

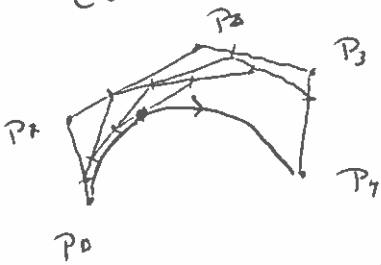
Bézier curve is...

$$C(t) = \sum_{k=0}^n P_k B_{n,k}(t)$$

$t \in [0, 1]$

$C(0) = P_0$

$C(1) = P_n$



$n=2$:

$$B_{0,2}(t) = (1-t)^2$$

$$B_{1,2}(t) = 2(1-t)t$$

$$B_{2,2}(t) = t^2$$

$$C(t) = (1-t)^2 P_0 + 2(1-t)t P_1 + t^2 P_2$$

$C(0) = P_0$; $C(1) = P_2$

$$C(\frac{1}{2}) = \frac{1}{2} \left(\frac{P_0 + P_1}{2} + P_1 + \frac{P_1 + P_2}{2} \right) = \frac{1}{2} \left(\frac{P_0 + P_2}{2} + P_1 \right)$$

$$C(t) = (1-t) \left((1-t)P_0 + tP_1 \right) + t \left((1-t)P_1 + tP_2 \right)$$

$0 \leq t \leq 1$

Convex Comb. of Convex Comb
 \downarrow
 Convex Comb.

This recursion is called
 Casteljau alg.

09/26/2017

Spline Functions

$S_1^0(\Delta)$

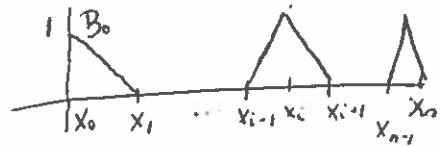
$\Delta = \{x_0, \dots, x_n\} \subset [a, b]$

Consider $f \in S_1^0(\Delta)$ such that
 $f \in C^0([a, b])$

$f|_{[x_i, x_{i+1}]} \in \mathbb{P}_1$

$\dim S_1^0(\Delta) = n+1$

Define basis in $S_1^0(\Delta)$



$B_i(x_j) = \delta_{i,j}$
 Kronecker function

$$B_0(x) = \begin{cases} \frac{x-x_1}{x_0-x_1}, & [x_0, x_1] \\ 0, & \text{otherwise} \end{cases}$$

$$B_i(x) = \begin{cases} \frac{x-x_{i-1}}{x_i-x_{i-1}}, & [x_{i-1}, x_i] \\ \frac{x_{i+1}-x}{x_{i+1}-x_i}, & [x_i, x_{i+1}] \\ 0, & \text{otherwise} \end{cases}$$

$$B_n(x) = \begin{cases} \frac{x-x_{n-1}}{x_n-x_{n-1}}, & [x_n, x_{n+1}] \\ 0, & \text{otherwise} \end{cases}$$

Clearly, these are lin. independent

Given f_0, \dots, f_n ; $f_i = f(x_i)$
 Find $s \in S_1(\Delta)$ such that

$$s(x_i) = f_i$$

$$s|_{[x_{i-1}, x_i]} = f_{i-1} + (x-x_{i-1})f[x_{i-1}, x_i]$$

$$s|_{[x_i, x_{i+1}]} = f_i + (x-x_i)f[x_i, x_{i+1}]$$

$$\text{Let } L(f)(x) = \sum_{i=0}^n f(x_i) B_i(x)$$

$$\|f - L(f)\|_{\infty} \approx ?$$

$$L(f)|_{[x_i, x_{i+1}]} = f(x_i) + (x-x_i)f[x_i, x_{i+1}]$$

$$|f(x) - L(f)(x)| = |(x-x_i)(x-x_{i+1})f[x_i, x_{i+1}, x]|$$

$$\leq \frac{h^2}{2} \sup_{a \leq x \leq b} |f''(\xi)|$$

where $h = \max_i h_i$, $h_i = x_i - x_{i-1}$

Notice then that as # mesh points $\rightarrow \infty$,

$$L(f)(x) \rightarrow f$$

$$\|L(f)\| = \max_{0 \leq i \leq n-1} \sup_{x_i \leq x \leq x_{i+1}} |L(f)(x)|$$

$$= \max_{0 \leq i \leq n-1} |f(x_i)| \leq \|f\|$$

Let $g \in S_1(\Delta)$.

$$L(g) = g$$

$$\|f - L(f)\| \leq \|f - g - L(f) + L(g)\|$$

$$\leq \|f - g\| + \|L(g - f)\|$$

$$\leq \|f - g\| + \|f - g\|$$

$$= 2\|f - g\|$$

$$\inf_{g \in S_1(\Delta)} \|f - g\| \leq \|f - L(f)\| = 2\|f - g\|$$

Piecewise Cubic Splines

$$S_3^1 = \{f: f \in C^1[a, b], f|_{[x_i, x_{i+1}]} \in \mathbb{P}_3\}$$

$h \in S_3^1(\Delta)$. On $[x_i, x_{i+1}]$

poly given by 4 coefficients. So 4n coeff. in total. But we need C^1 , so at nodes the derivatives agree so

$$4n - 2(n-1)$$

$$= 2n + 2$$

parameters

$$f h^{(j)}(x_i) = f^{(j)}(x_i) \quad i=0, \dots, n; j=0, 1$$

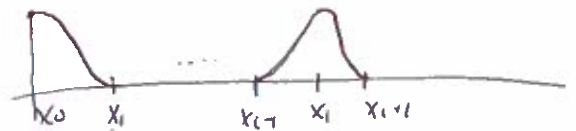
$$h_{i,0}(x_j) = \delta_{ij}$$

$$h_{i,0}'(x_j) = 0$$

$$h_{i,1}(x_j) = 0$$

$$h_{i,1}'(x_j) = \delta_{ij}$$

$$i, j = 0, \dots, n$$



$$h_{i,0}(x) = \begin{cases} \frac{2}{h_{i-1}^3} (x-x_{i-1})^2 (x-x_i - \frac{h_{i-1}}{2}), & [x_{i-1}, x_i] \\ \frac{2}{h_i^3} (x-x_{i-1})^2 (x-x_i + \frac{h_i}{2}), & [x_i, x_{i+1}] \end{cases}$$

$$|f - H(f)(x)| = |(x-x_i)^2 (x-x_{i+1})^2 f[x_i, 2; x_{i+1}, 2]|$$

$$\leq \frac{h_i^2}{4} \max_{a \leq t \leq b} |f^{(4)}(t)|$$

$$\Delta = \{a = x_0 < x_1 < \dots < x_{n-1} < x_n = b\}$$

$$S_m^k(\Delta) = \left\{ s : s|_{[x_i, x_{i+1}]} \in \mathbb{P}_m, s \in C^k[a, b] \right\}$$

$$\dim S_m^k(\Delta) = \underbrace{(m+1)(n)}_{\text{coeff. int.}} - \underbrace{(k+1)(n-1)}_{\text{cont.}}$$

$$= mn + n - kn + k + n + 1$$

$$= mn - kn + k + 1$$

$$= n(m-k) + (k+1)$$

$$s(x_j) = f(x_j) ; j = 0, \dots, n$$

Take $k = m-1$

$$S_m^{m-1}(\Delta)$$

$$\dim S_m^{m-1}(\Delta) = m+n$$

f periodic spline:

$$s^{(\ell)}(a) = s^{(\ell)}(b) ; \ell = 1, \dots, m-1$$

Natural Spline

$$m = 2\ell - 1 ; \ell \geq 2$$

$$s^{(\ell-j)}(a) = s^{(\ell-j)}(b) = 0$$

$$j = 0, 1, \dots, \ell-2$$

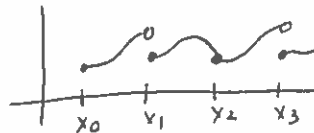
09/28/2017

Splines

$$\Delta = \{a = x_0 < x_1 < \dots < x_{n-1} < x_n = b\}$$

$f: [a, b] \rightarrow \mathbb{R}$ is piecewise poly.

if $f|_{[x_i, x_{i+1}]}$ is a polynomial.



Def: f spline degree m if

$$1) f|_{[x_i, x_{i+1}]} \in \mathbb{P}_m$$

$$2) f \in C^{m-1}([a, b])$$

$$S_m^{m-1}(\Delta) = S_m(\Delta)$$

$$\dim S_m(\Delta) = n(m+1) - m(n-1)$$

$$= m+n$$

want simplest spline. $\Delta = \{x_0\}$.

Let $s \in S_m(\Delta)$. Let $s \equiv 0$ for $x < x_0$. Then $s \in \mathbb{P}_m$ for $x \geq x_0$.

$$s^{(0)}(x_0) = s^{(1)}(x_0) = \dots = s^{(m-1)}(x_0) = 0$$

$$\text{So } s(x) = c(x-x_0)^m$$

Fitting a spline: Consider $S_2(\Delta) \ni s$

$$\Delta = \{0, 1, 2\}$$

$$s(x) = \begin{cases} a_{0,2}x^2 + a_{0,1}x + a_{0,0}, & [0,1] \\ a_{1,2}x^2 + a_{1,1}x + a_{1,0}, & [1,2] \end{cases}$$

Also $s \in C^1(0,2)$.

Cont. e $x=1 \rightarrow$

$$a_{0,2} + a_{0,1} + a_{0,0} = a_{1,2} + a_{1,1} + a_{1,0}$$

Cont. s' e $x=1 \rightarrow$

$$2a_{0,2} + a_{0,1} = 2a_{1,2} + a_{1,1}$$

Need more constraints:

$$S(0) = f_0, S(1) = f_1$$

$$S(2) = f_2; S'(0) = f_0'$$

$$a_{00} = f_0$$

$$a_{12} + a_{11} + a_{10} = f_1$$

$$4a_{12} + 2a_{11} + a_{10} = f_2$$

$$a_{01} = f_0'$$

$$\begin{pmatrix} 1 & 1 & 1 & -1 & -1 & -1 \\ 2 & 1 & 0 & -2 & -1 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 1 & 1 \\ 0 & 0 & 0 & 4 & 2 & 1 \\ 0 & 1 & 0 & 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} a_{02} \\ a_{01} \\ a_{00} \\ a_{12} \\ a_{11} \\ a_{10} \end{pmatrix} = \begin{pmatrix} 0 \\ f_0' \\ f_0 \\ f_1 \\ f_2 \\ f_0' \end{pmatrix}$$

inv.

second possibility: Match value at $\frac{1}{4}, \frac{1}{2}, \frac{3}{4}, \frac{3}{2}$, two on each

$$\text{int. in } \Delta \begin{pmatrix} 1 & 1 & 1 & -1 & -1 & -1 \\ 2 & 1 & 0 & -2 & -1 & 0 \\ 1/16 & 1/4 & 1 & 0 & 0 & 0 \\ 1/4 & 1/2 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 1 & 1 \\ 1/16 & 1/4 & 1 & 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} a_{02} \\ a_{01} \\ a_{00} \\ a_{12} \\ a_{11} \\ a_{10} \end{pmatrix} = \begin{pmatrix} 0 \\ f_0' \\ f_0 \\ f_1 \\ f_2 \\ f_0' \end{pmatrix}$$

Third possibility: Match values at $0, 1/3, 1/4, 1/2$

$$\begin{pmatrix} 1 & 1 & 1 & -1 & -1 & -1 \\ 2 & 1 & 0 & -2 & -1 & 0 \\ 1/16 & 1/4 & 1 & 0 & 0 & 0 \\ 1/4 & 1/2 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 \\ 1/16 & 1/4 & 1 & 0 & 0 & 0 \end{pmatrix} = \begin{pmatrix} \vdots \\ \vdots \\ \vdots \\ \vdots \\ \vdots \\ \vdots \end{pmatrix}$$

non-inv.

Determination of Cubic Spline

$$S''(a) = 0, S''(b) = 0$$

$$S \in \mathcal{S}_3(\Delta)$$

$$\Delta = \{x_0, \dots, x_n\}$$

$$S|_{[x_i, x_{i+1}]} \in \mathbb{P}_3$$

$$4n - 3(n-1) = 4n - 3n + 3 = n + 3$$

We impose

$$S(x_i) = f_i; i = 0, \dots, n$$

$$h_i = x_{i+1} - x_i$$

Call $S''(x_j) = M_j$ moments of S

We show S can be determined in terms of moments. Moments can be computed from interpolating data by solving a tridiagonal system

$$1) S''(x) = M_i \frac{x_{i+1} - x}{h_{i+1}} + M_{i+1} \frac{x - x_i}{h_{i+1}}$$

Integrate twice & det. int. constant from f_i

$$2) S(x) = M_i \frac{(x_{i+1} - x)^3}{6h_{i+1}} + M_{i+1} \frac{(x - x_i)^3}{6h_{i+1}} + A_i (x - x_i) + B_i$$

$$\text{Set } S(x_i) = f_i \\ S(x_{i+1}) = f_{i+1}$$

$$4) f_i = S(x_i) = \frac{M_i h_{i+1}^2}{6} + B_i$$

$$5) f_{i+1} = S(x_{i+1}) = \frac{M_{i+1} h_{i+1}^2}{6} + A_i h_{i+1} + B_i$$

$$B_i = f_i - M_i \frac{h_{i+1}^2}{6}$$

$$6) A_i = \frac{f_{i+1} - f_i}{h_{i+1}} - \frac{h_{i+1}}{6} (M_{i+1} - M_i)$$

$$S(x) = a_i + b_i(x - x_i) + c_i(x - x_i)^2 + d_i(x - x_i)^3$$

$$a_i = f_i$$

$$c_i = M_i/2$$

$$b_i = \frac{f_{i+1} - f_i}{h_{i+1}} - \frac{2M_i + M_{i+1}}{6} \frac{h_{i+1}}{h_{i+1}}$$

Show how to compute M_i ;
 $i = 0, \dots, n$ from f_0, \dots, f_n

$$M_0 = M_n = 0$$

S' is cont. at nodes

$$S'(x_i) = S'(x_i^-) = S'(x_i^+)$$

$$S'(x) = -M_i \frac{(x_{i+1} - x)^2}{2h_{i+1}} + M_{i+1} \frac{(x - x_i)^2}{2h_{i+1}} + \frac{f_{i+1} - f_i}{h_{i+1}} - \frac{h_{i+1}}{6} (M_{i+1} - M_i)$$

$$S'(x_i^+) = -M_i \frac{h_{i+1}}{2} + \frac{f_{i+1} - f_i}{h_{i+1}} - \frac{h_{i+1}}{6} (M_{i+1} - M_i)$$

$$= \frac{f_{i+1} - f_i}{h_{i+1}} - \frac{h_{i+1}}{3} M_i - \frac{h_{i+1}}{6} M_{i+1}$$

$$S'(x_i^-) = \frac{f_i - f_{i-1}}{h_i} + \frac{h_i}{3} M_i + \frac{h_i}{6} M_{i-1}$$

Now:

$$\frac{h_i}{6} M_{i-1} + \frac{h_i + h_{i+1}}{3} M_i + \frac{h_{i+1}}{6} M_{i+1} = \frac{f_{i+1} - f_i}{h_{i+1}} - \frac{f_i - f_{i-1}}{h_i}$$

$$i = 1, \dots, n-1$$

$$\mu_i := \frac{h_i}{h_i + h_{i+1}}$$

$$\lambda_i := \frac{h_{i+1}}{h_i + h_{i+1}}; \quad \mu_i + \lambda_i = 1$$

$$\delta_i := \frac{6}{h_i + h_{i+1}} \left(\frac{f_{i+1} - f_i}{h_{i+1}} - \frac{f_i - f_{i-1}}{h_i} \right)$$

Then...

$$\mu_i M_{i-1} + 2M_i + \lambda_i M_{i+1} = \delta_i$$

$$\text{If } M_0 = M_n = 0$$

$$\lambda_0 = \mu_n = \delta_0 = \delta_n = 0$$

$$\begin{pmatrix} 2 & \lambda_0 & & & 0 \\ \mu_1 & 2 & \lambda_1 & & \\ & \mu_2 & 2 & \dots & \\ & & & \dots & \lambda_{n-1} \\ 0 & & & & 2 \end{pmatrix} \begin{pmatrix} M_0 \\ M_1 \\ \vdots \\ M_n \end{pmatrix} = \begin{pmatrix} \delta_0 \\ \vdots \\ \delta_n \end{pmatrix}$$

Try to think or prove that the above matrix is strictly diagonally dominant.

10/03/2017

Cubic Splines & B-splines
 (absent)

10/05/2017

B-splines

Our def based on divided diff.

$$F[t_0, \dots, t_n] = \sum_{j=0}^n F(t_j) \prod_{\substack{s=0 \\ s \neq j}}^n (t_j - t_s)^{-1}$$

$$\delta^n(t_0, \dots, t_n) F = f[t_0, \dots, t_n]$$

$$F(t, y)$$

$$t_+ = \max\{t, 0\}$$

$$t_+^0(t) = \begin{cases} 0, & t \leq 0 \\ 1, & 0 < t \end{cases}$$



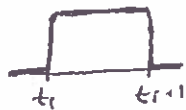
Nodes:

$$\dots < t_{-1} \leq t_0 < t_1 < t_2 < \dots$$

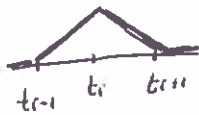
$$\lim_{i \rightarrow \pm \infty} t_i = \pm \infty$$

$$B_i^r(x) = (t_{i+r+1} - t_i) \delta_t^{r+1}(t_i, \dots, t_{i+r+1}) (t-x)_+^r$$

$$B_i^0(x)$$



$$B_i^1(x)$$



Lemma: (Leibniz)

$$F = gh \rightarrow$$

$$F[t_i, \dots, t_{i+k}] = \sum_{r=i}^{i+k} g[t_i, \dots, t_r] h[t_r, \dots, t_{i+k}]$$

Compare to: $F^{(n)} = \sum \binom{n}{k} F^{(n-k)} g^{(k)}$

P₂:

$$G_r(t) = g[t_i] + \sum_{s=i+1}^{i+k} g[t_i, \dots, t_s] (t-t_i) \dots (t-t_{s-1})$$

$$H(t) = h[t_{i+k}] + \sum_{s=i}^{i+k-1} h[t_s, \dots, t_{i+k}] (t-t_{s+1}) \dots (t-t_{i+k})$$

Consider $F(t) = G(t)H(t)$

$$F(t) = \left(\sum_{r=i}^{i+k} a_r b_s \right) \left(\sum_{s=i}^{i+k} a_r b_s \right)$$

$$= \underbrace{\sum_{r \leq s} a_r b_s}_{P_1(t)} + \sum_{r > s} a_r b_s \underbrace{\quad}_{P_2(t)}$$

Examine $P_2(t)$

$$P_2(t_j) = 0 \quad j = i, i+1, \dots, i+k \rightarrow$$

$$\delta^k(t_i, \dots, t_{i+k}) P_2 = 0$$

a_r $r = i, \dots, i+k$ (contains) $(t-t_i) \dots (t-t_{r-1})$
of deg $r+1$

b_s $s \geq r$ (contains) $(t-t_{s+1}) \dots (t-t_{i+k})$
of deg $i+k-s$

$a_r b_s$ is of deg $r+k-s$

when $r > s, r-1 \geq s$ then each term

a_r contains $(t-t_i) \dots (t-t_s)$

$P_1(t)$ must interpolate f at t_i, \dots, t_{i+k}

$$\delta^k(t_i, \dots, t_{i+k})$$

$$\delta^k F = \delta^k P_1 + \delta^k P_2$$

$$\delta^k F = \delta^k P_1$$

Leading coeff. of $P_i(t)$

Leading coefficient of P_i is a sum of leading coeff. in terms of deg X in

$$\sum_{r=i}^{i+k} \text{Arb}_r = \sum_{r=i}^{i+k} g[t_i, \dots, t_r] h[t_r, \dots, t_{i+k}]$$

$$(t-t_i)(t-t_r)(t-t_{i+k})$$

Recurrence (de Boor, Cox)

$$B_i^r(x) = \left(\frac{x-t_i}{t_{i+r}-t_i} \right) B_i^{r-1}(x) + \left(\frac{t_{i+r+1}-x}{t_{i+r+1}-t_{i+1}} \right) B_{i+1}^{r-1}(x)$$

$$(t-x)_t^r = (t-x)(t-x)_t^{r-1}$$

$$g(t) = t-x$$

$$g[t_i] = g(t_i) = t_i - x$$

$$g(t_i, t_{i+1}) = 1$$

$$g[t_i, \dots, t_j] = 0, j > i+1$$

$$\delta_t^{r+1}(t_i, \dots, t_{i+r+1})(t-x)_t^r =$$

$$\delta_t^{r+1} [(t-x)(t-x)_t^{r-1}]$$

$$= g[t_i] \delta_t^{r+1} (t-x)_t^{r-1} + g[t_i, t_{i+1}] \delta_t^r (t-x)_t^{r-1}$$

$$= (t-x) \frac{\delta_t^r(t_{i+1}, \dots, t_{i+r+1})(t-x)_t^{r-1} - \delta_t^r(t_i, \dots, t_{i+r})(t-x)_t^{r-1}}{t_{i+r+1}-t_i} + \delta_t^r (t-x)_t^{r-1}$$

⋮

$$= \frac{x-t_i}{t_{i+r+1}-t_i} \delta_t^r(t_{i+1}, \dots, t_{i+r})(t-x)_t^{r-1} + \frac{t_{i+r+1}-x}{t_{i+r+1}-t_i} \delta_t^r(t_{i+1}, \dots, t_{i+r+1})(t-x)_t^{r-1}$$

Compact Support:

$$B_i^r(x) = 0 \text{ for } x \notin (t_i, t_{i+r+1}) \quad r \geq 0$$

$$\delta_t^{r+1}(t_i, \dots, t_{i+r+1})(t-x)_t^r$$

Compact Support

$$B_i^r(x) = 0 \text{ for } x \notin (t_i, t_{i+r+1}) \quad r \geq 0$$

$$\delta_t^{r+1}(t_i, \dots, t_{i+r+1})(t-x)_t^r$$

$$t_i \leq t \leq t_{i+r+1} < x$$

$$x < t \leq t \leq t_{i+r+1}$$

Positivity in $r+1$ (conject. int.)

$$B_i^r(x) > 0; \quad x \in (x_i, \dots, x_{i+r+1}), \quad r \geq 0$$

10/10/2017

Numerical Integration (Quadrature)

$$I(f) := \int_a^b f(x) dx = \lambda_1 f_1 + \dots + \lambda_n f_n =: Q(f)$$

$$E(f) = I(f) - Q(f) \quad \text{want small}$$

Interpolatory Quadratures

$$I(f) = \int_a^b f(x) dx \quad a, b \text{ finite}$$

$$x_i := a + ih, \quad h = \frac{b-a}{n}; \quad i = 0, \dots, n$$

$$Q(f) = \int_a^b L_n(x) dx; \quad L_n(x) = \sum_{i=0}^n l_i(x) f_i$$

Lag. int. of $f \in C^1$

$$Q(f) = \int_a^b \sum_{i=0}^n l_i(x) f_i dx = \sum_{i=0}^n \underbrace{\int_a^b l_i(x) dx}_{\lambda_i \text{ weights}} f_i$$

Can determine weights by undet. coefficients
 $Q(x^i) = I(x^i); \quad i = 0, 1, \dots, n$

Find weights λ_j of such poly.

$$Q(f) = \sum \lambda_j f_j$$

$$\sum \lambda_j x_j^i = \int_a^b x^i dx$$

$[x_j^i]$ Vandermonde matrix

We say Q is exact of deg $\leq n$
 if $Q(f) = I(f) \quad \forall f \in \mathbb{P}_n$

$$E(f) = I(f) - Q(f)$$

$$= \int_a^b f - L_n dx$$

$$= \int_a^b \frac{f^{(n+1)}(\xi_x)}{(n+1)!} (x-x_0) \dots (x-x_n) dx$$

$\xi_x \in [a, b]$

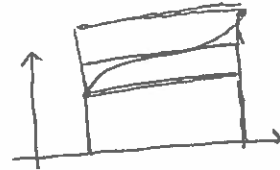
Other (constant) interpolants of f

$$Q(f) = (b-a) f(a) = h f(a)$$

$$Q(f) = (b-a) f(b) = h f(b)$$

$$Q(f) = (b-a) f\left(\frac{a+b}{2}\right) = h f(x_{1/2})$$

$$h = x_1 - x_0 = b - a$$



We will prove:

$$\int_a^b f(x) dx = \underbrace{h f(x_{1/2})}_{Q(f)} + \frac{1}{24} h^3 \overset{2^{nd} \text{ deriv.}}{f''(\mu)}$$

$$L_0(x) = f(x_{1/2})$$

$$f(x) - L_0(x) = f'(\xi_x) (x - x_{1/2})$$

$$E(f) = \int_a^b f'(\xi_x) (x - x_{1/2}) dx$$

$$|E(f)| \leq M_1 \overset{\text{bound}}{\int_a^b |x - x_{1/2}| dx}$$

$$= M_1 \frac{(b-a)^2}{2}$$

$$M_1 = \sup_{a \leq x \leq b} |f'(x)|$$

But cannot infer degree of exactness from this.

$$H(x) = f(x_{1/2}) + f'(x_{1/2})(x - x_{1/2})$$

Hermite interpolant at $x = x_{1/2}$

$$H(x_{1/2}) = f(x_{1/2})$$

$$H'(x_{1/2}) = f'(x_{1/2})$$

$$Q(f) = \mathcal{I}(H) = \underbrace{\mathcal{I}(f(x_{1/2}))}_{\mathcal{I}(L_0)} + \mathcal{I}(f'(x_{1/2})(x - x_{1/2}))$$

$$Q(f) = \mathcal{I}(L_0)$$

$$f(x) - H(x) = \frac{1}{2!} (x - x_{1/2})^2 f^{(2)}(\xi_x)$$

$$E(f) = \mathcal{I}(f - Q) = \int_a^b \frac{1}{2} (x - x_{1/2})^2 f''(\xi_x) dx$$

MVT \mathcal{I} :

Φ, ψ cont. on $[a, b]$. if ψ doesn't change

$\exists \mu$

$$\int_a^b \phi(x) \psi(x) dx = \phi(\mu) \int_a^b \psi(x) dx$$

$$E(f) = \frac{1}{2} f''(\mu) \int_a^b (x - x_{1/2})^2 dx$$

$$= \frac{h^3}{24} f''(\mu)$$

Lin. interpolant of f . Trap rule

$$x_0 = a \quad x_1 = b$$

$$L(x) = \frac{x-b}{a-b} f(a) + \frac{x-a}{b-a} f(b)$$

$$Q(f) = \mathcal{I}(L) = \frac{f(a)}{b-a} \int_a^b (b-x) dx + \frac{f(b)}{b-a} \int_a^b (x-a) dx$$

$$= \frac{f(a)}{b-a} \frac{1}{2} (b-a)^2 + \frac{f(b)}{b-a} \frac{1}{2} (b-a)^2$$

$$Q(f) = \frac{b-a}{2} (f(a) + f(b))$$

$$\mathcal{I}(f) - Q(f) =$$

$$\int_a^b \frac{f''(\xi_x)}{2!} \underbrace{(x-a)(x-b)}_{\text{always neg.}} dx$$

$$= \frac{(b-a)^3}{8} \int_{-1}^1 (t^2 - 1) dt = -\frac{(b-a)^3}{6}$$

$$x = \frac{b-a}{2} t + \frac{a+b}{2}; \quad -1 \leq t \leq 1$$

$$E(f) = \frac{(b-a)^3}{12} f''(\xi)$$

Trap rule exact for \mathbb{P}_1 .

Newton-Cotes

Quad. Interp.: Simpson's Rule

$$x_0 = a, \quad x_1 = \frac{a+b}{2}, \quad x_2 = b$$

$$L_2(x) = \frac{(x-x_1)(x-x_2)}{(x_0-x_1)(x_1-x_2)} f_0 +$$

$$\frac{(x-x_0)(x-x_2)}{(x_1-x_0)(x_1-x_2)} f_1 +$$

$$\frac{(x-x_0)(x-x_1)}{(x_2-x_0)(x_2-x_1)} f_2$$

Integrate $L_2(x)$

$$x - x_1 = s$$

$$x - x_0 = s + h$$

$$x - x_2 = s - h$$

10/12/2017

$$\int_{x_0}^{x_2} (x-x_1)(x-x_2) dx =$$

$$\int_{-h}^h s(s-h) ds = \frac{2}{3} h^3$$

$$\frac{\int_{x_0}^{x_2} (x-x_0)(x-x_2) dx =$$

$$\int_{-h}^h (s-h)(s+h) ds =$$

$$\int_{-h}^h (s^2-h^2) ds =$$

$$\frac{4}{3} h^3$$

$$\int_{x_0}^{x_2} (x-x_0)(x-x_1) dx =$$

$$\int_{-h}^h s(s+h) ds = \frac{2}{3} h^3$$

$$Q(f) = h \left(\frac{1}{3} f_0 + \frac{4}{3} f_1 + \frac{1}{3} f_2 \right)$$

$$= \frac{b-a}{6} (f_0 + 4f_1 + f_2)$$



$$E(f) = I(f - L_2)$$

Define $H_3 \in \mathbb{P}_3$

$$\begin{cases} H_3(x_i) = f_i, & i=0,1,2 \\ H_3'(x_1) = f'(x_1) \end{cases}$$

$$H_3(x) = L_2(x) + K(x-x_0)(x-x_1)(x-x_2)$$

$$H_3'(x_1) = f'(x_1)$$

$$H_3'(x) = L_2'(x) + K(x-x_0)(x-x_2)$$

Simpson's Rule:

$$[a,b] \quad x_0 = a, \quad x_1 = \frac{a+b}{2}, \quad x_2 = b$$

$$Q(f) = \frac{b-a}{6} (f_0 + 4f_1 + f_2)$$

Exact for poly. of deg. ≤ 2

Actually exact for $p \in \mathbb{P}_3$

Smooth function f . Define Hermite

$$H_3 \in \mathbb{P}_3$$

$$H_3(f_i) = f(x_i) \quad i=0,1,2$$

$$H_3'(x_1) = f'(x_1)$$

Newton:

$$H_3(x) = L_2(x) + K(x-x_0)(x-x_1)(x-x_2)$$

Find K

$$L_2'(x_1) = \frac{f_2 - f_0}{2h}$$

$$H_3'(x_1) = \frac{f_2 - f_0}{2h} - Kh^2$$

$$K = \frac{1}{h^2} \left(\frac{f_2 - f_0}{2h} - f'(x_1) \right)$$

$$I(H_3) = I(L_2) + K \int_a^b (x-x_0)(x-x_1)(x-x_2) dx$$

$$= I(L_2) + K \int_{-h}^h (s+h) \underbrace{s(s-h)}_{\substack{s(s^2-h^2) \\ \text{odd}}} ds$$

$$= I(L_2)$$

$$E(f) = I(f - H_3)$$

$$f(x) = H_3(x) + (x-x_0)(x-x_1)^2(x-x_2) \frac{f^{(4)}(\xi)}{4!}$$

$$E(f) = \int_{x_0}^{x_2} (x-x_0)(x-x_1)^2(x-x_2) \frac{f^{(4)}(\xi)}{4!} dx$$

$$= \frac{1}{4!} f^{(4)}(\mu) \int_{-1}^1 \underbrace{(s+h)(s-h)s^2 ds}_{-\frac{4}{15}h^5}$$

$$= -\frac{1}{90} h^5 f^{(4)}(\mu)$$

We need a 'good' set of nodes for quadrature \rightarrow Orth. poly.

$$\langle f, g \rangle = \int_a^b f(x)g(x)w(x) dx$$

$$w(x) > 0 \quad \& \quad x^n w \in L^1(a,b) \quad n \geq 1$$

weight

$$\|f\|_{L^2(a,b)} = \int_a^b |f(x)|^2 w(x) dx$$

Use Gram-Schmidt to produce orth poly.

$$\{1, x, x^2, \dots\} \quad \{p_0, p_1, \dots\}$$

$$p_0(x) = 1; \quad p_1(x) = x - \frac{\langle 1, x \rangle}{\langle 1, 1 \rangle}$$

$$p_n(x) = x^n - \sum_{i=0}^{n-1} \lambda_{i,n} p_i(x)$$

monic $\lambda_{i,n} = \frac{\langle x^n, p_i \rangle}{\langle p_i, p_i \rangle}$

Thm (Triple Recursion) $\exists!$ seq. of poly. $\{p_n\}_{n=0}^{\infty}$ with p_n monic \Rightarrow

$$\langle p_n, q \rangle = 0 \quad \forall q \in \mathbb{P}_{n-1}$$

Satisfying...

$$p_n(x) = (x - \lambda_n) p_{n-1}(x) - \mu_n p_{n-2}(x)$$

$$\lambda_n = \frac{\langle x p_{n-1}, p_{n-1} \rangle}{\|p_{n-1}\|^2}$$

$$\mu_n = \frac{\|p_{n-1}\|^2}{\|p_{n-2}\|^2}$$

$$\langle p_n, q \rangle = 0 \quad \forall q \in \mathbb{P}_{n-1}$$

$$p_n(x) - x p_{n-1}(x) \in \mathbb{P}_{n-1} = \sum_{i=0}^{n-1} a_i p_i(x)$$

For $i \leq n-3$

$$x p_i(x) \in \mathbb{P}_{n-2}$$

$$a_i \langle p_i, p_i \rangle = \langle p_n, p_i \rangle - \langle x p_{n-1}, p_i \rangle$$

$$= - \langle x p_{n-1}, p_i \rangle$$

$$= - \langle p_{n-1}, x p_i \rangle$$

$$= 0$$

$$i=n-1; a_{n-1} = - \frac{\langle x_{p_{n-1}}, p_{n-1} \rangle}{\langle p_{n-1}, p_{n-1} \rangle} = -\lambda_n$$

$$i=n-2; a_{n-2} = \frac{\langle x_{p_{n-1}}, p_{n-2} \rangle}{\langle p_{n-2}, p_{n-2} \rangle}$$

$$\begin{aligned} \langle x_{p_{n-1}}, p_{n-2} \rangle &= \langle p_{n-1}, p_{n-1} - p_{n-1} + x_{p_{n-1}} \rangle \\ &= \langle p_{n-1}, p_{n-1} \rangle + \underbrace{\langle p_{n-1} - p_{n-1} + x_{p_{n-1}} \rangle}_{=0} \end{aligned}$$

Extended prop:

$$\|p_n\| \leq \|S\| \quad \text{S monic in } \mathbb{R}^n$$

$$S = p_n - \sum_{i=0}^{n-1} c_i p_i$$

$$\|S\|^2 = \|p_n\|^2 + \sum_{i=0}^{n-1} |c_i|^2 \|p_i\|^2 \geq \|p_n\|^2$$

p_n has n real roots in (a, b)

Let x_1, \dots, x_k real roots of p_n of odd multiplicity.

If $k=n \rightarrow$ true

$$\text{If } k < n \rightarrow q_k(x) = \prod_{i=1}^k (x - x_i)$$

$$q_k \equiv 1 \quad k=0$$

$$\int_a^b \underbrace{p_n(x) q_k(x)}_{\text{only roots mult.}} w(x) dx = 0$$

Gauss Quadrature

Want quad

$$Q(f) = \sum_{i=1}^k \lambda_i f(x_i)$$

Weights computed from

$$\int_a^b \delta_i(x) w(x) dx = \lambda_i$$

By choosing Gauss nodes, execution of Q could be $2k+1$ inst of k .

If $k=0$ set $p_0(x) \equiv 1$

A quad. Q with $k+1$ nodes

and exact for $2k+1$ poly if

unique called Gauss quad

10/19/2017

Gaussian Quad.

$$I(f) = \int_a^b f(x) w(x) dx$$

$$Q(f) = \sum_{i=0}^k \lambda_i f(x_i)$$

Idea: Choose x_0, \dots, x_k so that

as to raise exactness of Q to

$$2k+1 = 2 \# \text{ nodes} - 1$$

Lemma: A quadr rule Q with nodes x_0, \dots, x_k equal to roots of P_{k+1} and weights $\lambda_0, \dots, \lambda_k$ given

$$\lambda_i = \int_a^b l_i(x) w(x) dx \quad i=0, \dots, k$$

$$l_i(x) = \prod_{\substack{j=0 \\ j \neq i}}^k \frac{x - x_j}{x_i - x_j}$$

Claim exact for $P \in \mathbb{P}_{2k+1}$

$P_{k+1} \in \mathbb{P}_{k+1}$ orthogonal in $L^2_w(a,b)$

First show exact for $P \in \mathbb{P}_k$.

$$Q(f) = \sum_{i=0}^k \lambda_i f(x_i) = \sum_{i=0}^k \int_a^b l_i(x) w(x) dx f(x_i)$$

$$= \int_a^b \underbrace{\sum_{i=0}^k l_i(x) f(x_i)}_{L_k(f)(x)} w(x) dx$$

$$= \int_a^b f(x) \omega(x) dx = I(f)$$

There is a poly. in \mathbb{P}_{k+1} for which

Q is also exact \mathbb{P}_{k+1}

$$Q(p_{k+1}) = 0 = \langle 1, p_{k+1} \rangle$$

$$= \int_a^b p_{k+1}(x) w(x) dx$$

$$= I(p_{k+1})$$

$f \in \mathbb{P}_{2k+1}$

$$f(x) = q(x) p_{k+1} + r(x)$$

$r \in \mathbb{P}_k$

$$P_{k+1}(x) = (x - x_0) \dots (x - x_k)$$

$$f(x_i) = r(x_i); \quad i=0, \dots, k$$

$$\int_a^b f(x) w(x) dx =$$

$$\int_a^b q(x) p_{k+1}(x) w(x) dx$$

$\langle q, p_{k+1} \rangle$

$$+ \int_a^b r(x) w(x) dx$$

$$= Q(r) \sum_{i=0}^k \lambda_i r(x_i)$$

$$= \sum \lambda_i f(x_i) = Q(f)$$

Uniqueness

Suppose Q exact for \mathbb{P}_{2k+1}
so in particular for $q \pi$

$$\pi(x) = \prod x - x_i$$

$q \in \mathbb{P}_k$

$$0 = Q(q \pi) = I(q \pi)$$

$$0 = \int_a^b \pi(x) q(x) w(x) dx$$

$\langle \pi, q \rangle = 0$ and π monic $\rightarrow \pi = p_{k+1}$

$$Q(l_i) = \sum_{j=0}^k \lambda_j l_i(x_j) = \int_a^b \dots$$

The weights of Q are positive
 Q is exact for $l_i^2(x) \in \mathbb{P}_{2k}$

$$\begin{aligned} \lambda_i &= \sum_{j=0}^k \lambda_j l_i^2(x_j) \\ &= Q(l_i^2) \\ &= I(l_i^2) \\ &= \int_a^b l_i^2(x) w(x) dx \end{aligned}$$

Error Gauss quadr.

Let $H \in \mathbb{P}_{2k+1}$

$$\begin{aligned} H^{(l)}(x_i) &= f^{(l)}(x_i) \\ i &= 0, \dots, k \\ l &= 0, 1 \end{aligned}$$

$$f(x) = H(x) + f[x_0, 2; \dots; x_{k+1}, 2, x] \pi^2(x)$$

$$I(f) = I(H) + \int_a^b f[x_0, 2; \dots; x_{k+1}, 2, x] \pi^2(x) w(x) dx$$

$$I(H) = Q(H) = Q(f)$$

$$E(f) = I(f) - Q(f)$$

$$= I(f) - I(H)$$

$$= I(f - H)$$

$$= \int_a^b f[x_0, 2; \dots; x_{k+1}, 2, x] \pi^2(x) w(x) dx$$

$$= f[x_0, 2; \dots; x_{k+1}, 2, \xi] \int_a^b \pi^2(x) w(x) dx$$

$$= \frac{1}{(2k+2)!} f^{(2k+2)}(\xi) \int_a^b \pi^2(x) w(x) dx$$

The weights of Q are positive. Q exact for $l_i^2(x) \in \mathbb{P}_{2k}$

$$0 < \lambda_i = \sum_{j=0}^k \lambda_j l_i^2(x_j)$$

$$= Q(l_i^2)$$

$$= I(l_i^2) = \int_a^b l_i^2(x) w(x) dx$$

Gauss-Lobatto

Degree exactness $2k-1$

$$x_0 = a \quad x_k = b$$

x_1, \dots, x_{k-1} are roots of P_{k-1}

Orth. poly of deg $k-1$ on (a, b)

with weight

$$w(x) = w(x) (x-a)(b-x)$$

$$(-1, 1)$$

$$w(x) = (1-x^2)^{-1/2}$$

Gauss-Chebyshev quad.

$$(0, 1) \quad w(x) \equiv 1$$

Gauss-Legendre

Ex: Gauss-Hermite quad nodes
 x_0, x_1

$$Q(f) = \sum_{i=0}^1 A_i f(x_i)$$

exact for poly. of deg. $2k+1=3$

$$Q(f) \approx I(f) = \int_{-\infty}^{\infty} \underbrace{f(x)}_{w(x)} e^{-x^2} dx$$

$$Q(x^i) = I(x^i); \quad i=0,1,2,3$$

$$\int_{-\infty}^{\infty} e^{-x^2} dx = \sqrt{\pi}$$

$$Q(1) = I(1) = \sqrt{\pi}$$

$$A_0 + A_1 = \sqrt{\pi}$$

$$\int_{-\infty}^{\infty} x e^{-x^2} dx = 0$$

$$\int_{-\infty}^{\infty} x^3 e^{-x^2} dx = 0$$

$$\int_{-\infty}^{\infty} x^2 e^{-x^2} dx =$$

$$-\frac{1}{2} \int_{-\infty}^{\infty} x (e^{-x^2})' dx$$

$$\int_{-\infty}^{\infty} x (e^{-x^2})' dx = \frac{1}{2} \left[x e^{-x^2} \right]_{-\infty}^{\infty}$$

$$-\frac{1}{2} \int_{-\infty}^{\infty} e^{-x^2} dx$$

$$= -\frac{\sqrt{\pi}}{2}$$

$$\begin{cases} A_0 + A_1 = \sqrt{\pi} \\ A_0 x_0 + A_1 x_1 = 0 \\ A_0 x_0^2 + A_1 x_1^2 = \frac{\sqrt{\pi}}{2} \\ A_0 x_0^3 + A_1 x_1^3 = 0 \end{cases}$$

$$\text{Let } \pi(x) = (x-x_0)(x-x_1)$$

$$\text{Find } x_0, x_1$$

$$= x^2 + px + q$$

Want to find p, q not x_0, x_1 .

$$q(1) \quad p(2) \quad 1(3)$$

$$q\sqrt{\pi} = A_0 q + A_1 q$$

$$0 = A_0 x_0 p + A_1 x_1 p$$

$$\frac{\sqrt{\pi}}{2} = A_0 x_0^2 + A_1 x_1^2$$

$$\sqrt{\pi}(q + \frac{1}{2}) = A_0 (x_0^2 + p x_0 + q) + A_1 (x_1^2 + p x_1 + q)$$

$$q = -\frac{1}{2}$$

$$0 = x^2 - \frac{1}{2}$$

$$x = \pm \frac{1}{\sqrt{2}}$$

$$A_0 = A_1 = \frac{\sqrt{\pi}}{2}$$

$$Q(f) = \frac{\sqrt{\pi}}{2} f(-\frac{1}{\sqrt{2}}) + \frac{\sqrt{\pi}}{2} f(\frac{1}{\sqrt{2}})$$

10/24/2017

Convergence of Quad. Formulas

$$Q_k(f) = \sum_{i=0}^k \lambda_{i,k} f(x_{i,k}) \approx \int_a^b f(x) w(x) dx$$

Formula with $k+1$ nodes

$$E_k(f) = \int_a^b f(x) w(x) dx - Q_k(f)$$

$$\lim_{k \rightarrow \infty} E_k(f) \stackrel{?}{=} 0$$

Thm: (Polya) $\lim_{k \rightarrow \infty} E_k(f) = 0$ for any cont. function f iff

a) $\exists M \Rightarrow \forall k \sum_{i=0}^k |\lambda_{ik}| \leq M$

b) $\forall n \lim_{k \rightarrow \infty} E_k(x^n) = 0$

Sufficiency: Assume (a) & (b)

$\exists p \sup_{x \in [a,b]} |f-p(x)| = \|f-p\|_{\infty} \leq \epsilon$

$$|E_k(f)| - |E_k(p)| \leq |E_k(f-p)|$$

$$= \left| \int_a^b (f-p)(x) w(x) dx - \sum_{i=0}^k \lambda_{ik} (f-p)(x_{ik}) \right|$$

$$\leq \epsilon \left[\int_a^b w(x) dx + \sum_{i=0}^k |\lambda_{ik}| \right]$$

$$\leq \epsilon \left[M + \int_a^b w(x) dx \right]$$

$$|E_k(f)| \leq \epsilon \left[M + \int_a^b w(x) dx \right] + \underbrace{|E_k(p)|}_{\rightarrow 0}$$

Euler-MacLaurin Formula:

Study of quad. error in comp. trap.

Trap. Rule:

$$\int_{x_0}^{x_0+h} f(x) dx \approx \frac{h}{2} (f(x_0) + f(x_0+h))$$

Comp Trap.

$$\int_{x_0}^{x_k} f(x) dx \approx T_k(f) = h \left(\frac{1}{2} f(x_0) + f(x_1) + \dots + f(x_{k-1}) + \frac{1}{2} f(x_k) \right)$$

$$x_i = x_0 + ih; i=0, \dots, k$$

Develop the formula for error on $[0,1]$ interpolating by party

$$\int_0^1 f(x) dx = \int_0^1 (x - 1/2) f(x) dx$$

$$= (x - 1/2) f(x) \Big|_0^1 - \int_0^1 (x - 1/2) f'(x) dx$$

$$= \frac{1}{2} (f(0) + f(1)) - \int_0^1 (x - 1/2) f'(x) dx$$

$$B_1(x) = x - 1/2$$

$$B_1'(x) = 1$$

$$\int_0^1 B_1(x) dx = 0$$

$$B_1(x) = \int_0^x dt + B_{1,0}$$

$\rightarrow -1/2 = b_1$

$$\int_0^1 f(x) dx = \frac{1}{2} (f(0) + f(1)) - \int_0^1 \underbrace{B_1(x) f'(x)}_{\text{Error in trap.}} dx$$

$$f(0) = \int_0^1 f(x) dx - \frac{1}{2} (f(1) - f(0)) + \int_0^1 B_1(x) f'(x) dx$$

$$B_2(x) : \frac{1}{2} B_2'(x) = B_1(x); \int_0^1 B_2(x) dx = 0$$

$$B_2(x) = 2 \int_0^x B_1(t) dt + \underbrace{B_{2,0}}_{b_2 = B_2(0) = B_1(0)}$$

$$B_2(x) = x^2 - x + 1/6$$

$$\int_0^1 B_1(x) f'(x) dx = \int_0^1 \frac{1}{2} B_2'(x) f'(x) dx$$

$$= \frac{1}{2} B_2(x) f'(x) \Big|_0^1 - \int_0^1 \frac{1}{2} B_2(x) f''(x) dx$$

$$= \frac{b_2}{2} (f'(1) - f'(0)) - \frac{1}{2} \int_0^1 B_2(x) f''(x) dx$$

2-term Euler - Maclaurin

$$\int_0^1 f(x) dx = \frac{1}{2} (f(0) + f(1)) - \frac{b_2}{2} (f'(1) - f'(0)) + \frac{1}{2} \int_0^1 B_2(x) f''(x) dx$$

$$\begin{cases} B_0(x) \equiv 1 \\ B_{n+1}'(x) = (n+1) B_n(x); n \geq 0 \\ \int_0^1 B_n(x) dx = 0 \end{cases}$$

$$B_{n+1}(x) = (n+1) \int_0^x B_n(t) dt + B_{n+1}(0)$$

$$B_{n+1}'(x) = -(n+1) \int_0^x B_n(t) dt + dx$$

$$b_{n-1} = B_{n-1}(0) = B_{n+1}(1)$$

$$B_n(x) = \sum_{k=0}^n \binom{n}{k} b_{n-k} x^k \quad \text{Taylor formula}$$

$$\sum_{k=0}^n \binom{n+1}{k} B_k(x) = (n+1) x^n$$

$$B_n(x+1) - B_n(x) = n x^{n-1}; n \geq 2$$

$$B_n(1-x) = (-1)^n B_n(x)$$

$$\sum_{k=0}^n \binom{n}{k} b_k = 0; n \geq 2$$

$$b_{2j+1} = 0$$

$$B_n''(x) = n(n-1) B_{n-2}(x)$$

⋮

$$B_n^{(k)}(x) = n(n-1)\dots(n-k+1) B_{n-k}(x)$$

$$B_0(t) = 1$$

$$B_1(t) = t - 1/2$$

$$B_2(t) = t^2 - t + 1/6$$

$$B_3(t) = t^3 - 3/2 t^2 + 1/2 t$$

$$\int_0^1 f(x) dx = \frac{1}{2} (f(0) + f(1))$$

$$- \sum_{j=1}^n \frac{b_{2j}}{(2j)!} (f^{(2j)}(1) - f^{(2j)}(0)) + R_{2n}$$

$$R_{2n} = \frac{1}{(2n)!} \int_0^1 B_{2n+1}(x) f^{(2n+1)}(x) dx$$

$$T(h) = \tau_0 + \tau_1 h^\alpha + \mathcal{O}(h^{2\alpha})$$

General scheme for Richardson extrapolation
 (computable) $\tau_0 = T(0)$ unknown
 error

$$0 < b < 1$$

$$T(h) = \tau_0 + \tau_1 h^\alpha + \mathcal{O}(h^{2\alpha})$$

$$T(bh) = \tau_0 + \tau_1 b^\alpha h^\alpha + \mathcal{O}(b^\alpha h^{2\alpha})$$

$$= \tau_0 + \tau_1 b^\alpha h^\alpha + \mathcal{O}(h^{2\alpha})$$

$$\tau_1 h^\alpha = \frac{T(h) - T(bh)}{1 - b^\alpha} + \mathcal{O}(h^{2\alpha})$$

$$\tau_0 = T(h) + \frac{T(bh) - T(h)}{1 - b^\alpha} + \mathcal{O}(h^{2\alpha})$$

10/26/2017

$$\int_0^1 f(x) dx = \frac{1}{2}(f(0) + f(1)) - \sum_{j=1}^n \frac{b_{2j}}{(2j)!} (f^{(2j-1)}(1) - f^{(2j-1)}(0)) + R_{2n}$$

$$R_{2n} = \frac{1}{(2n)!} \int_0^1 B_{2n}(x) f^{(2n)}(x) dx$$

$$\int_0^N f(x) dx = \frac{1}{2} f(0) + f(1) + \dots + f(N-1) + \frac{1}{2} f(N) + R_{2n}$$

$$\phi(t) = f(a+th)$$

$$\int_a^b f(x) dx = h \int_0^N \phi(t) dt$$

$$\begin{aligned} \int_a^b f(x) dx &= T_n(f) - h \sum_{j=1}^n \frac{b_{2j}}{(2j)!} (\phi^{(2j-1)}(N) - \phi^{(2j-1)}(0)) \\ &\quad - \frac{h}{(2n+1)!} \int_0^N B_{2n+1}(\xi) \phi^{(2n+1)}(\xi) dt \\ &\quad - \frac{h^{2n+1}}{(2n+1)!} \int_a^b B_{2n+1}\left(\frac{x-a}{h}\right) f^{(2n+1)}(x) dx \end{aligned}$$

$$\int_a^b f(x) dx = T_n(f) - \sum_{j=1}^n \frac{b_{2j}}{(2j)!} h^{2j} (f^{(2j-1)}(b) - f^{(2j-1)}(a)) + R_{2n+1}$$

$$R_{2n+1} = -\frac{h^{2n+1}}{(2n+1)!} \int_a^b B_{2n+1}\left(\frac{x-a}{h}\right) f^{(2n+1)}(x) dx$$

If $a=0, b=N, f(x) = x^{m-1}$

$$\sum_{0 \leq k < N} f(k) = \int_0^N x^{m-1} dx + \sum_{k=1}^n \frac{b_k}{k!} (f^{(k-1)}(N) - f^{(k-1)}(0))$$

$$= \frac{1}{m} \sum_{k=0}^m \binom{m}{k} \frac{1}{b^k} N^{m-k}$$

$m=3$

$$\sum_{0 \leq k < N} k^2 = \frac{N^3}{3} - \frac{N^2}{2} + \frac{N}{6}$$

[Early Fourier Analysis - Montgomery]

Approximation

Let V be a normed vector space over \mathbb{R} or over \mathbb{C} .

Let $W \subseteq V$ be a finite dim. vector subspace.

$f \in V$. Find $h^* \in W$ such that

$$\|f - h^*\| \leq \|f - h\| \quad \forall h \in W$$

Ex:

$$V = C[a, b], \quad \|f\| = \sup_{a \leq x \leq b} |f(x)|$$

$$W = \mathbb{T}_n$$

Uniform approximation

Ex:

$$V = L^2(a, b), \quad \|f\| = \langle f, f \rangle^{1/2} = \left(\int_a^b |f(x)|^2 dx \right)^{1/2}$$

Least square approximation

$$V = L^2_p[(0, a)], \quad \text{where } L^2_p[(0, a)] = \{f: (0, a) \rightarrow \mathbb{R}\}$$

$$f(x) = f(x+a) \quad (\text{'a'-periodic})$$

$$\|f\| = \int_0^a |f(x)|^2 dx$$

$$W = \{\text{trig. poly.}\}$$

$$p(x) = \sum_{n=-N}^N c_n e^{i 2\pi n x/a}$$

Orthogonal Projection

$$\langle u, w \rangle = 0 \quad \text{orth. vectors}$$

Lem: $v \in V, w \neq 0. \exists!$ vect.

$$v_{||} \perp v_{\perp} \Rightarrow$$

$$v = v_{||} + v_{\perp}$$

$$v_{||} = c w \quad \text{for some } c \in \mathbb{C}$$

$$\langle v_{\perp}, v_{||} \rangle = 0$$

Unique:

$$\langle v, w \rangle = \langle v_{||}, w \rangle + \langle v_{\perp}, w \rangle$$

$$= \langle c w, w \rangle + \langle v_{\perp}, w \rangle$$

$$= c \langle w, w \rangle$$

$$= c \|w\|^2$$

(why?)

$$c = \frac{\langle v, w \rangle}{\|w\|^2}$$

$$\text{Ex: } \frac{\langle v, w \rangle}{\|w\|^2}; \quad v_{||} = c w, \quad v_{\perp} = v - c w$$

$$v_{\perp} + v_{||} = v$$

$$\langle v_{\perp}, w \rangle = \langle v, w \rangle - c \langle w, w \rangle$$

$$= \langle v, w \rangle - \frac{\langle v, w \rangle}{\|w\|^2} \langle w, w \rangle$$

$$= 0$$

Cauchy-Sch.

$$|\langle v, w \rangle| \leq \|v\| \|w\|$$

$w=0$ eq.

$$w \neq 0 \Rightarrow v = v_{||} + v_{\perp}; \quad v_{||} = c w$$

$$c = \frac{\langle v, w \rangle}{\|w\|^2}$$

$$\begin{aligned} \|v\|^2 &= \|v_{||}\|^2 + \|v_{\perp}\|^2 \\ &\geq \|v_{||}\|^2 = |c|^2 \|w\|^2 \\ &= \frac{|\langle v, w \rangle|^2}{\|w\|^2} \end{aligned}$$

$$\|v^*\|^2 \|w\|^2 \geq |\langle v, w \rangle|^2 \quad \checkmark$$

Lem: $\|u+v\| \leq \|u\| + \|v\|$

Thm: $W \subset V, (V, \|\cdot\|), \dim W < \infty$
for any $f \in V, \exists$ an 'optimal'
element h^*

$$\|f - h^*\| \leq \|f - h\| \quad \forall h \in W$$

Pf: $\|f - h^*\| \leq \|f\| \quad (h=0)$

$$K = \{h \in W : \|f - h\| \leq \|f\|\}$$

K closed in f.d. $\rightarrow K$ compact

want to obtain min.

$$\|f - h\|; h \in K$$

Cont. in h

$$|\|f - (h+g)\| - \|f - h\|| \leq \|g\|$$

$$\|g\| = \|f - (h+g) - (f - h)\| \geq \|f - (h+g)\| - \|f - h\|$$

$$\|g\| = \|f - h - (f - (h+g))\|$$

$\|f - h\|$ attains min on K

$$\|f - h^*\| \text{ min.}$$

h^* may not be unique.

10/31/2017

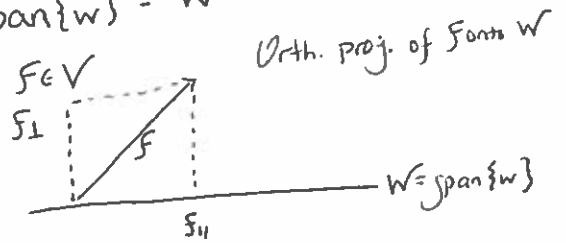
$W \subset V, f \in V$. Find
 $h^* \in W \ni \|f - h^*\| \leq \|f - h\|$
for all $h \in W$.

$$E_W(f) := \|f - h^*\|$$

Error approx f by fun. in W .

Know W fin dim. $\rightarrow \exists h^*$

$$\text{span}\{w\} = W \subset V$$



$$f = f_{||} + f_{\perp}; \quad f_{||} = c w$$

$$c = \frac{\langle f, w \rangle}{\|w\|^2}$$

$f_{||}$ will be our approx.

$$\langle f - f_{||}, w \rangle = 0 \quad \forall w \in W$$

$$\text{Then } \langle f - f_{||}, f_{||} - w \rangle = 0$$

By Pythag.

$$\|f - f_{||}\|^2 + \|f_{||} - w\|^2 = \|f - w\|^2$$

Generally, $\|f - f_{||}\| \leq \|f - w\|$

Let $S \subset W$ be set of best approx to
 $f \in V$. Then S is convex; $h_1, h_2 \in S$
distinct

$$\|f - h_1\| = E_W(f) = \|f - h_2\|$$

Consider $\alpha h_1 + \beta h_2; \alpha, \beta \neq 0; \alpha + \beta = 1$

$$\|f - (\alpha h_1 + \beta h_2)\| = \|\alpha(f - h_1) + \beta(f - h_2)\|$$

$$* f = (\alpha + \beta)f$$

$$\leq \alpha \|f - h_1\| + \beta \|f - h_2\|$$

$$= (\alpha + \beta) E_W(f) = E_W(f)$$

We say that normed space V is strictly convex if $\|x\| \leq r$,
 $x \neq y \Rightarrow \|x+y\| < 2r$

OR
 $\|f\| = \|g\|, f \neq g \Rightarrow \|f+g\| < 2$

Approx in strictly convex V giving a unique solution.

$g_1 \neq g_2 \Rightarrow \|f-g_1\| = \|f-g_2\| = E_w(f) = \inf_{h \in W} \|f-h\|$

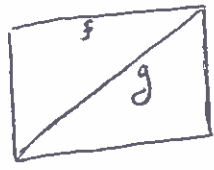
$\|f - \frac{1}{2}(g_1+g_2)\| = \|\frac{1}{2}(f-g_1) + \frac{1}{2}(f-g_2)\|$
 $< E_w(f)$

Contradicts $g_1, g_2 \in E_w(f)$

$V = C[0,1], \|f\| = \sup_{0 \leq x \leq 1} |f(x)|$

$(V, \|\cdot\|)$ is not strictly convex.

$f(t) = 1; g(t) = t$



$\|f\| = \|g\| = 1$ as \sup

$(f+g)(t) = t+1$

$\sup_x |t+1| = 2$

$W \subset V, V$ inner product space

Thm: Given $f \in V$ there exists a unique $f^* \in W$.

$\|f - f^*\| \leq \inf_{w \in W} \|f - w\|$

$E_w(f) = \inf_{w \in W} \|f - w\| = d > 0$

$D_n = \{w \in W : \|f - w\| \leq d + \frac{1}{n}\}$

When $n \rightarrow \infty$, diameter $D_n \rightarrow 0$
 $w_1, w_2 \in D_n$

$\|w_1 + w_2 - 2f\|^2 + \|w_1 - w_2\|^2$
 $= 2(\|w_1 - f\|^2 + \|w_2 - f\|^2)$
 $\leq 4(d^2 + \frac{2d}{n} + \frac{1}{n^2})$
 $= 4d^2 + \frac{8d}{n} + \frac{4}{n^2}$

$\| \frac{w_1 + w_2}{2} - f \| \geq d$

$\|w_1 - w_2\|^2 \leq \frac{8d}{n} + \frac{4}{n^2}$

Choose $\tilde{w}_n \in D_n; n \geq 1$ then

$\lim \tilde{w}_n = f^*$

Orthogonal Proj. Thm: V inner product
 $W \subset V, W$ complete vector subspace

$f \in V \Rightarrow f^* \in W$ is the best approx
of f iff $\langle f - f^*, w \rangle = 0 \forall w \in W$

PF:

Suff: $\langle f - f^*, \tilde{w} \rangle = 0 \forall \tilde{w} \in W$

Set $\tilde{w} = \frac{f^* - w}{\|f^* - w\|}$ for $w \in W$. Given

$\langle f - f^*, f^* - w \rangle = 0$

$\|f - f^*\|^2 + \|f^* - w\|^2 = \|f - w\|^2$

$\|f - f^*\| \leq \|f - w\|$

$\|f - f^*\| \leq \|f - w\| \forall w \in W$

Nec: Show $\langle f - f^*, \tilde{w} \rangle = 0 \forall \tilde{w}$

Let $w_k = f^* + \alpha(w - f^*)$

$\alpha \in \mathbb{C}$, $w \in W$ arbitrary

$$\begin{aligned} \|f - f^*\|^2 &\leq \|f - w_k\|^2 \\ &= \|f - f^* - \alpha(w - f^*)\|^2 \\ &= \|f - f^*\|^2 + \|w - f^*\|^2 \\ &\quad - \alpha \langle w - f^*, f - f^* \rangle \\ &\quad - \bar{\alpha} \langle w - f^*, f - f^* \rangle \end{aligned}$$

But then

$$|\alpha|^2 \|w - f^*\|^2 \geq \alpha \langle w - f^*, f - f^* \rangle + \bar{\alpha} \langle w - f^*, f - f^* \rangle$$

Let $\alpha = |\alpha| e^{i\theta}$. Divide by $|\alpha|$,
Change $\alpha \rightarrow 0$. This gives
 $0 \geq e^{i\theta} z + \bar{e}^{i\theta} z$; $z = \langle w - f^*, f - f^* \rangle$

Set $\theta = 0$, $\text{Re } z \leq 0$
If $\theta = \pi$, $\text{Re } z \geq 0$ } $\text{Re } z = 0$

$\theta = \frac{\pi}{2}, \frac{3\pi}{2}$ $\text{Im } z \geq 0$

If $f^* = \sum_{j=1}^n c_j \phi_j$; $f \in V$

$\langle f - f^*, \phi_j \rangle = 0$

$\langle f - \sum_{j=1}^n c_j \phi_j, \phi_i \rangle$

$\sum c_j \langle \phi_j, \phi_i \rangle = \langle f, \phi_i \rangle$

$$\begin{pmatrix} \langle \phi_1, \phi_1 \rangle & \dots & \langle \phi_1, \phi_n \rangle \\ \vdots & \ddots & \vdots \\ \langle \phi_n, \phi_1 \rangle & \dots & \langle \phi_n, \phi_n \rangle \end{pmatrix} \begin{pmatrix} c_1 \\ \vdots \\ c_n \end{pmatrix} = \begin{pmatrix} \langle f, \phi_1 \rangle \\ \vdots \\ \langle f, \phi_n \rangle \end{pmatrix}$$

G

Gram matrix

$Gc = F$

$G_{i,j} = \langle \phi_i, \phi_j \rangle$
row \uparrow column \downarrow

If $\{\phi\}$ orthogonal

$c_i = \frac{\langle f, \phi_i \rangle}{\langle \phi_i, \phi_i \rangle}$

$f^* = \sum \frac{\langle f, \phi_i \rangle}{\langle \phi_i, \phi_i \rangle} \phi_j$

G symmetric or hermitian
symmetric positive definite:

$\langle Gx, x \rangle > 0$ unless $x = 0$

11/02/2017

Thm (Best L^2 approx)

V inner product space

$W \subset V$ complete subspace
 $f \in V$, $f^* \in W$ is the B.A. \rightarrow best approx
element in W

$\|f - f^*\| \leq \|f - w\| \quad \forall w \in W$

\Leftrightarrow

$\langle f - f^*, w \rangle = 0 \quad \forall w \in W$

If $\dim W = n$ then

$f^* = \sum_{i=1}^n \frac{\langle f, \phi_i \rangle}{\langle \phi_i, \phi_i \rangle} \phi_i$

$\{\phi_i\}_{i=1}^n$ orth. system in W (basis)

If $\dim W = \infty$, then need $\{\phi_i\}$ orthogonal set which is dense in W .
 $\{\phi_i\}$ then a Hilbert basis in W .

$\{e_n(x)\}$ on $L^2(-\pi, \pi)$

$$V = L^2[(0,1)]$$

$$f \in V \quad f(x) = \begin{cases} x & 0 < x < 1/2 \\ 1-x & 1/2 \leq x < 1 \end{cases}$$



$$\phi(x) = \begin{cases} 1, & 0 \leq x \leq 1 \\ 0, & \text{otherwise} \end{cases}$$

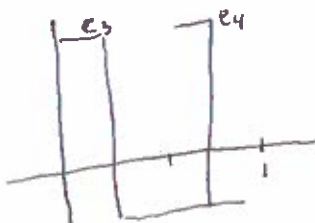
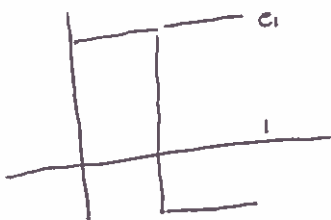
$$\psi(x) = \begin{cases} 1 & 0 \leq x < 1/2 \\ -1 & 1/2 \leq x < 1 \\ 0 & \text{otherwise} \end{cases}$$

$$e_1(x) = \phi(x)$$

$$e_2(x) = \psi(x)$$

$$e_3(x) = \psi(2x)$$

$$e_4(x) = \psi(2x-1)$$



$\{e_1, \dots, e_n\}$ orth. system in $L^2(0,1)$

$$\langle e_1, e_1 \rangle = 1$$

$$\langle e_2, e_2 \rangle = 1$$

$$\langle e_3, e_3 \rangle = 1/2$$

$$\langle e_4, e_4 \rangle = 1/2$$

$$f^* = \sum_{i=1}^4 \frac{\langle f, e_i \rangle}{\langle e_i, e_i \rangle} e_i$$

$$\langle f, e_1 \rangle = 1/2$$

$$\langle f, e_2 \rangle = 0$$

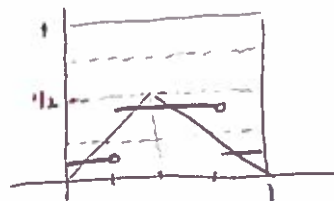
$$\langle f, e_3 \rangle = -1/6$$

$$\langle f, e_4 \rangle = 1/6$$

$$f^* = \frac{1}{4} e_1 - \frac{1}{8} e_3 + \frac{1}{8} e_4$$

$$= \frac{2e_1 - e_3 + e_4}{8}$$

$$f^*(x) = \begin{cases} \frac{1}{8} & 0 \leq x < 1/4 \\ \frac{2}{8} & 1/4 \leq x < 1/2 \\ \frac{3}{8} & 1/2 \leq x < 3/4 \\ \frac{1}{8} & 3/4 \leq x < 1 \\ 0 & \text{otherwise} \end{cases}$$



$$W = \text{span} \{e_1, \dots, e_4\}$$

Finite dim. spaces complete

Example above Haar functions

Linear Least-Square Fitting

A set of data points

$$(t_1, b_1), \dots, (t_m, b_m); t_i \in \mathbb{R}^k, b_i \in \mathbb{R}$$

We want to approx f by a lin. comb. of functions ϕ_1, \dots, ϕ_n (model functions)

$$\sum_{i=1}^m c_i \phi_i; \text{ model of } f.$$

$$f^* = \sum_{i=1}^m c_i \phi_i$$

$$\text{where } \phi_i = \begin{pmatrix} \phi_i(t_1) \\ \vdots \\ \phi_i(t_m) \end{pmatrix}$$

$$b = \begin{pmatrix} b_1 \\ \vdots \\ b_m \end{pmatrix}$$

want

$$\langle f - f^*, \phi_i \rangle = 0$$

$$\langle b, \phi_i \rangle = \langle f^*, \phi_i \rangle$$

$$\sum_{k=1}^m \left(\sum_{j=1}^n c_j \phi_j(t_k) \right) \phi_i(t_k)$$

$$\sum b_k \phi_i(t_k)$$

$$A = \begin{pmatrix} \phi_1(t_1) & \dots & \phi_n(t_1) \\ \vdots & \dots & \vdots \\ \phi_1(t_m) & \dots & \phi_n(t_m) \end{pmatrix}$$

$$A^T A c = A^T b$$

A sym. pos. def. \rightarrow inv.

By solving for c

$$r = b - Ac$$

When c satisfies normal system.
Obtain $c^* \Rightarrow$

$$\|r(c^*)\|_2 \leq \|r(c)\|_2$$

11/07/2017

$f \in V$; $W \subset V$ inner product space
closed subspace.

want $f^* \in W \Rightarrow$

$$\|f - f^*\| \leq \|f - w\| \quad \forall w \in W$$

$$\langle f - f^*, w \rangle = 0 \quad \forall w \in W$$

Thm: $\{\phi_i\}_{i=1}^{\infty}$ countable orthonormal family in $V \Rightarrow S_N f := \sum_{n=1}^N \langle \phi_n, f \rangle \phi_n$

The following are equiv:

1) Fin. lin. comb. $\sum a_i \phi_i$ are dense in V

2. For any $f \in V$

$$\lim_{N \rightarrow \infty} \|f - S_N f\| = 0$$

3. For any $f \in V$

$$\|f\|^2 = \sum_{n=1}^{\infty} |\langle f, \phi_n \rangle|^2$$

4. $\forall f \in V \exists \langle f, \phi_i \rangle = 0 \quad \forall i, f = 0$

\geq
Bessel Inequality

$$\|f\| \geq \|f - \sum_{i=1}^N a_i \phi_i\| \geq \epsilon$$

$$\|f - S_N(f)\|$$

$$2 \rightarrow 3: \|f\|^2 = \|f - S_N f + S_N f\|^2$$

$$= \langle f - S_N f, f - S_N f \rangle$$

$$= \|S_N(f)\|^2 + \|f - S_N f\|^2 + 2 \langle S_N f, f - S_N f \rangle$$

0

$$\|f - \sum \langle f, \phi_k \rangle \phi_k\|^2 \rightarrow \rightarrow 0 \text{ as } N \rightarrow \infty$$

$$= \|f\|^2 - \sum |\langle f, \phi_k \rangle|^2$$

3 → 4: Trivial

4 → 1: \bar{V} set of all fin. lin. comb. $\{\phi_i\}$

WTS \bar{V} dense in V .

$$\langle f, \phi_i \rangle = 0 \rightarrow f = 0 \text{ WTS } \bar{V} \text{ dense}$$

Suppose $\bar{V} \neq V$. $\exists \psi \in V \setminus \bar{V}$

Uniform Approx of Cont. Functions

$f \in C[a, b]$. Find $p \in \mathbb{P}_n$.

$$E_n(f) = \|f - p^*\| \leq \|f - q\| \quad \forall q \in \mathbb{P}_n$$

$$\|f\| = \sup_{a \leq x \leq b} |f(x)|$$

Characterization of $p^* \in \mathbb{P}_n$ given by Chebyshev Thm

Def: Function equioscillates on

$$x_0 < x_1 < \dots < x_k \text{ iff}$$

a) $|e(x_i)| = \|e\|$

b) $e(x_i) = -e(x_{i+1})$; $i=0, \dots, k-1$

Points x_0, \dots, x_k on alternating set for e (alternans)

Thm (Chebyshev) $f \in C[a, b]$. $p \in \mathbb{P}_n$

is B.A.P. iff $f - p$ equioscillates at $n+2$ points in $[a, b]$.

$$\sup \max_{x \in [a, b]} (f(x) - p(x)) = \max(\sup f, \sup g)$$

Lemma: \exists at least 2 distinct points

$$x_1, x_2 \in [a, b] \exists$$

$$|f(x_1) - p(x_1)| = E$$

$$|f(x_2) - p(x_2)| = E$$

$$f(x_2) - p(x_2) = -(f(x_1) - p(x_1))$$

$$\text{Def: } e(x) = (f - p)(x) \text{ (cont.)}$$

let cont. take value E or $-E$

$y = E$ or $y = -E$. WTS both

if not $e(x) > -E \quad \forall x \in [a, b]$

$$\min_{x \in [a, b]} e(x) = m > -E$$

$$c = \frac{E+m}{2} > 0$$

$$q = p + c; \quad f - q = f - p - c = e - c$$

$$-(E-c) = m - c \leq e(x) - c \leq E - c$$

$$\|f - q\| = \sup_{x \in [a, b]} |e(x) - c| \leq E - c < E$$

$f \in C[a, b]$ Find $p_0 \in \mathbb{P}_0$ constant

$$\Rightarrow \|f - p_0\| \leq \|f - c\|; \quad c \in \mathbb{R}$$

$$M = \sup_x f(x) = f(x_2)$$

$$E(c) = \sup_{x \in [a, b]} |f(x) - c|$$

$$|z| = \max(z, -z)$$

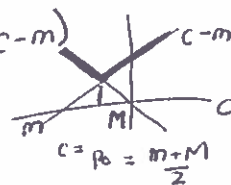
$$|f(x) - c| = \max(f(x) - c, c - f(x))$$

$$\sup |f(x) - c| = \max(\sup f - c, \sup c - f) = \max(M - c, c - m)$$

$$E(p_0) = \|f - p_0\| \text{ func.}$$

$f - p_0$ equioscillation $x_1 < x_2$

$$\frac{M-m}{2} = |f(x_1) - \frac{m+M}{2}| = \|f - \frac{m+M}{2}\|; \quad \begin{matrix} f(x_1) = \frac{m-M}{2} \\ f(x_2) = \frac{M-m}{2} \end{matrix}$$



11/09/2017

Thm: (Int. Value Theorem) f cont. function on $[a, b]$ with $f(a) < f(b)$, then $\forall y$ with $f(a) < y < f(b)$, $\exists x_0 \in [a, b]$ with $f(x_0) = y$.

Pf: (Divide & Conquer) Consider sequence of int. $[a_1, b_1] := [a, b], [a_2, b_2], \dots$ with $f(a_k) < y < f(b_k)$

$[a_2, b_2]$ 1/2 of $[a_1, b_1]$

$f(\frac{a+b}{2}) < y \rightarrow [a_2, b_2] := [\frac{a+b}{2}, b]$

otherwise $[a, \frac{a+b}{2}]$

$$b_k - a_k = 2^{1-k} (b_1 - a_1)$$

$y \in \cap [a_k, b_k]$. Seq. nested closed intv. with diam $[a_k, b_k] \rightarrow 0$. $a_k, b_k \rightarrow x_0$ limit and $f(x_0) = y$. \square

Thm: Let $f: [a, b] \rightarrow \mathbb{R}$ cont. $\exists f(a)f(b) \leq 0 \rightarrow \exists \xi \in [a, b]$ $\exists f(\xi) = 0$

Pf: $f(a) = 0$ or $f(b) = 0$. Done. Then $f(a) > 0 \neq f(b) < 0$ or $f(a) < 0, f(b) > 0$. Now use Int. Value Thm. \square

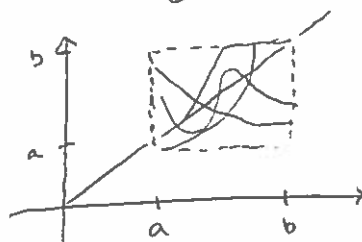
Thm (Brouwer Fixed Pt. Thm)
 $g: [a, b] \rightarrow [a, b]$ cont.
 then $\exists \xi \in [a, b]$ $\exists g(\xi) = \xi$. Fixed point of g .

Pf: $f(x) := x - g(x)$

$$f(a) = a - g(a) \leq 0$$

$$f(b) = b - g(b) \geq 0$$

Opposite signs. $f(a)f(b) \leq 0$. \square



Bijection

$$[a, b] = [a_0, b_0]$$

$$* x_0 := a + \frac{1}{2} (b_0 - a_0)$$

[num. better than $x_0 = \frac{b_0 + a_0}{2}$]

$$f(f(x_0)) = 0 \text{ if not}$$

$$(a_1, b_1) = \begin{cases} (x_0, b_0), & f(x_0)f(a_0) > 0 \\ (a_0, x_0), & f(x_0)f(a_0) < 0 \end{cases}$$

Interpret bijection in interpolation terms:

$L(x)$ affine function $\Rightarrow L(a_k) = -1, L(b_k) = 1$
 (just opposite sign)

$$L(x) = -1 + (x - a_k) \frac{2}{b_k - a_k}$$

Solve $L(x) = 0$ to find x_k (middle of int.)

$$x_k = \frac{1}{2} (a_k + b_k)$$

False position

$$L(a_k) = f(a_k)$$

$$L(b_k) = f(b_k)$$

$$L(x) = f(a_k) + (x - a_k) \frac{f(b_k) - f(a_k)}{b_k - a_k}$$

$$L(x) = 0$$

$$x_k = a_k - \frac{f(a_k)}{f(a_k) - f(b_k)} (b_k - a_k)$$

Simple iteration of $f(x) = 0$ (finding solution)

(converting into fixed pt. eqn)

$$f(x) = 0$$

$$\alpha f(x) = 0$$

$$g(x) = x$$

$$g(x) = x + \alpha f(x)$$

fixed point \rightarrow
 $f(x) = 0$

x_0 arb.

$$g(x_k) = x_{k+1}$$

Sufficient condition for conv. of simple iteration that $g: [a, b] \rightarrow [a, b]$ a contraction mapping.

$$\exists 0 < L < 1 \Rightarrow |g(x) - g(y)| < L|x - y|$$

ie Lipschitz function with constant < 1 .

$$|g(x)| - |g(y)| < |g(x) - g(y)| < L|x - y|$$

$$|g(x)| < |g(y)| + L|x - y|$$



"one bounding g "

Thm (Contraction Mapping Thm)

$g: [a, b] \rightarrow [a, b]$ cont.

g assumed to be a contraction
Then g has unique fixed point in $[a, b]$

Seq $\{x_k\}$ conv. to fixed pt for any starting point $x_0 \in [a, b]$.

Pf: \exists exists by Brouwer's. If Z

$$|z - \eta| = |g(z) - g(\eta)| < L|z - \eta|$$

$$x_k \rightarrow z$$

$$|z - x_k| = |g(z) - g(x_{k-1})| < L|z - x_{k-1}|$$

$\forall k \geq 1$ Generally,

$$0 \leq |z - x_k| \leq L^k |x_0 - z|$$

Squeeze & done. \square

If $|g'(x)| < L$ in (a, b) then

g is contraction:

$$\frac{g(x) - g(y)}{x - y} = g'(c)$$

& shown.

11/14/2017

Thm: (Local Contraction Thm)

$g: [a,b] \rightarrow [a,b]$ cont.
 $\bar{z} \in [a,b]; \bar{z} = g(\bar{z})$

(\exists by Brauer). Assume g' cont. in some neigh of \bar{z} and $|g'(\bar{z})| < 1$.
 The seq $\{x_k\}; g(x_k) = x_{k+1}$ conv. to \bar{z} provided x_0 'close' to \bar{z}

Pf: g' cont. $[\bar{z}-h, \bar{z}+h]$

$|g'(\bar{z})| < 1$. Let $I_\delta = [\bar{z}-\delta, \bar{z}+\delta]$ such that $|g'| \leq L < 1$ in I_δ .

Take $L = \frac{1}{2}(1 + |g'(\bar{z})|) < 1$

Choose $\delta \leq h \Rightarrow$

$$|g'(y) - g'(\bar{z})| \leq \frac{1}{2}(1 - |g'(\bar{z})|)$$

For $x \in I_\delta$. g' cont. \bar{z} . For all

$$\begin{aligned} x \in I_\delta, |g'(x)| &\leq |g'(x) - g'(\bar{z})| + |g'(\bar{z})| \\ &\leq \frac{1}{2}(1 - |g'(\bar{z})|) + |g'(\bar{z})| \\ &= \frac{1}{2}(1 + |g'(\bar{z})|) < 1 \end{aligned}$$

Suppose $x_k \in I_\delta$.

$$x_{k+1} - \bar{z} = g(x_k) - g(\bar{z}) = (x_k - \bar{z}) \underbrace{g'(\eta_k)}_{\text{between } x_k, \bar{z}}$$

$$|x_{k+1} - \bar{z}| \leq L |x_k - \bar{z}| \leq L\delta < \delta$$

$x_{k+1} \in I_\delta$, if $x_0 \in I_\delta \Rightarrow x_k \in I_\delta$

$$|x_k - \bar{z}| \leq L^k |x_0 - \bar{z}| \quad \square$$

~~4~~

$f: \mathbb{R} \rightarrow \mathbb{R}$. f cont. diff

$f(\alpha) = 0$. $f'(x) = 0$? Suppose $f'(\alpha) \neq 0$.

Construct $H(x)$, Hermite int. poly of f . If $x_n \approx \alpha$.

$$H(x_n) = f(x_n)$$

$$H'(x_n) = f'(x_n)$$

$$H(x) = \underbrace{f(x_n) + f'(x_n)(x - x_n)}_{\text{lin. of } f}$$

$$H(x) = 0$$

$$x_{n+1} = x_n - \frac{f(x_n)}{f'(x_n)} \quad \text{Newton's}$$

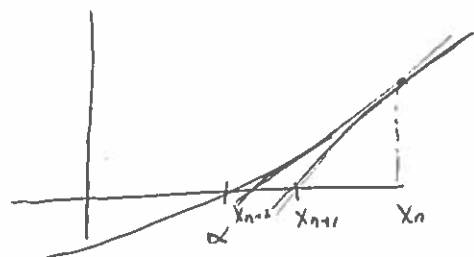
$$g(x) := x - \frac{f(x)}{f'(x)}$$

$x_{n+1} = g(x_n)$ find fixed point of g . $g(\alpha) = \alpha$.

$$g'(x) = 1 - \frac{f''(x)f(x) - f'(x)^2}{f'(x)^2}$$

$$g'(\alpha) = 1 - \frac{f''(\alpha)f(\alpha) - f'(\alpha)^2}{f'(\alpha)^2} = 0$$

In neigh α , $|g'(x)| \leq L < 1$
 Newton's convergence.



$$f: \mathbb{R}^n \rightarrow \mathbb{R}^n$$

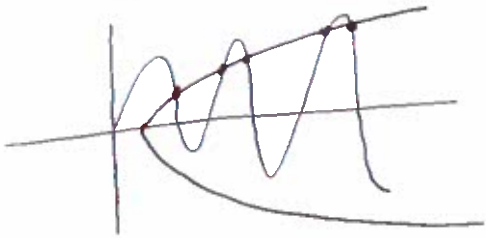
$$f(x) = \begin{pmatrix} f_1(x) \\ \vdots \\ f_n(x) \end{pmatrix}; \quad x = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix}$$

$$f(x) = 0$$

$$\begin{pmatrix} f_1(x) \\ \vdots \\ f_n(x) \end{pmatrix} = \vec{0}$$

$$f_1(x_1, x_2) = \frac{1}{2}x_1 \sin\left(\frac{1}{2}\pi x_1\right) - x_2 = 0$$

$$f_2(x_1, x_2) = x_2^2 - x_1 + 1$$



$$0 = f(x) \approx \underbrace{f(x_n) + f'(x_n)(x - x_n)}_{H(x)}$$

$$f(x_n) + f'(x_n)(x - x_n) = 0$$

$$f'(x_n)(x - x_n) = -f(x_n)$$

$$f'(x) = \begin{pmatrix} \frac{\partial f_1}{\partial x_1} & \dots & \frac{\partial f_1}{\partial x_n} \\ \vdots & \dots & \vdots \\ \frac{\partial f_n}{\partial x_1} & \dots & \frac{\partial f_n}{\partial x_n} \end{pmatrix} \text{ Jacobian}$$

$$x_{n+1} = x_n - f^{-1} f$$

Secant Method

$$f(x) = 0 \quad f(x) = L(x)$$

$$L(x) = f(x_n) + \frac{f(x_n) - f(x_{n-1})}{x_n - x_{n-1}} (x - x_{n-1}) = 0$$

$$L(x_{n-1}) = f(x_{n-1})$$

$$L(x_n) = f(x_n)$$

$$(x - x_n) \frac{f(x_n) - f(x_{n-1})}{x_n - x_{n-1}} = -f(x_n)$$

$$x_{n+1} = x_n - \frac{f(x_n)}{\frac{f(x_n) - f(x_{n-1})}{x_n - x_{n-1}}} = f'(x_n)$$

$$x_{n+1} = x_n - \frac{(x - x_{n-1}) f(x_n)}{f(x_n) - f(x_{n-1})}$$

$$f = x^2 - 6 \quad (\sqrt{6})^2 - 6 = 0$$

$$x_{n+1} = x_n - \frac{x_n^2 - 6}{2x_n}$$

$$= x_n - \frac{1}{2}x_n + \frac{6}{2x_n}$$

$$x_{n+1} = \frac{1}{2} \left(x_n + \frac{6}{x_n} \right)$$

Let 'a' be a first guess at $a > \sqrt{6}$

$$\frac{b}{a} < \frac{b}{\sqrt{b}} = \sqrt{b}$$

$$\bar{x} = \frac{1}{2} \left(a + \frac{b}{a} \right)$$

$$\begin{array}{ccc} & \uparrow & \uparrow \\ & >\sqrt{b} & <\sqrt{b} \\ \hline & \text{avg.} & \end{array}$$

$b < x_1^2 < x_0^2$
Mnemonic (Inv. Thm)

11/16/2017

Kantorovich Thm, Global Conv. of Newton's Method:

$$x_{n+1} = x_n + \underbrace{[DF(x_n)]^{-1}}_{\text{Jac. matrix}} F(x_n)$$

where $f: \mathbb{R}^n \rightarrow \mathbb{R}^n$; $n=0, 1, 2, \dots$

Thm: $f: U \subset \mathbb{R}^n \rightarrow \mathbb{R}^n$ be diff mapping. $x_0 \in U$ & $DF(x_0)$ inv. Define $h_0 = -[DF(x_0)]^{-1} f(x_0)$.

Define

$$(1) \begin{cases} h_0 = -[DF(x_0)]^{-1} f(x_0) \\ x_1 = x_0 + h_0 \\ U_1 = B_{1/2}(x_1) \end{cases}$$

If $\bar{U}_1 \subset U$ & $DF(x)$ satisfy the Lipschitz cond.

(2) $|DF(x_1) - DF(x_2)| \leq M |x_1 - x_2|$

and if the inequality

(3) $|f(x_0)| | [DF(x_0)]^{-1} |^2 M \leq 1/2$

is satisfied, the equation $f(x)=0$ has a unique solution \bar{u}_1 & Newton's Method x_n converges to this solution.

$$u = \begin{pmatrix} u_1 \\ \vdots \\ u_n \end{pmatrix}; |u| = \sqrt{u_1^2 + \dots + u_n^2}$$

$$M = [m_{ij}]; |M| = \sqrt{m_{ij}^2}$$

Units:

u unit u

r unit $f(u)$

r unit $|f(x_0)|$

r/u unit $DF(x_0)$

$\frac{r^2}{u^2}$ unit $([DF(x_0)]^{-1})^2$

$\frac{r^2}{u^2}$

Then M unit r/u^2

Finding Lipschitz constant M

Ex: $f: \mathbb{R}^2 \rightarrow \mathbb{R}^2$

$$f\left(\begin{pmatrix} x_1 \\ x_2 \end{pmatrix}\right) = \begin{pmatrix} x_1 - x_2^2 \\ x_1^2 + x_2 \end{pmatrix}$$

$$DF = \begin{pmatrix} 1 & -2x_2 \\ 2x_1 & 1 \end{pmatrix}$$

$$DF(x) - DF(y) = \begin{pmatrix} 0 & -2(x_2 - y_2) \\ 2(x_1 - y_1) & 0 \end{pmatrix}$$

$$|DF(x) - DF(y)| = \frac{\sqrt{4(x_1 - y_1)^2 + 4(x_2 - y_2)^2}}{2\sqrt{(x_1 - y_1)^2 + (x_2 - y_1)^2}} = 2|x - y|$$

$$|DF(x) - DF(y)| \leq \frac{2}{3}|x - y|$$

Ex:

$$f\left(\begin{pmatrix} x \\ y \end{pmatrix}\right) = \begin{pmatrix} x^5 - y^2 + xy - a \\ y^4 + x^2y - b \end{pmatrix}$$

$$u_1 = x^2 \quad u_2 = y^2 \quad u_3 = xy = u_1 u_2$$

$$\tilde{u} = \begin{pmatrix} x \\ y \\ u_1 \\ u_2 \\ u_3 \end{pmatrix} = \begin{pmatrix} x u_3 - u_2 + xy - a \\ u_2^2 + u_1 y - b \\ u_1 - x^2 \\ u_2 - y^2 \\ u_3 - u_1^2 \end{pmatrix}$$

$$\tilde{f} = \begin{pmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{pmatrix}$$

$$D\bar{F} = \begin{pmatrix} u_3 + y & x & 0 & -1 & x \\ 0 & u_1 & y & 2y_2 & 0 \\ -2x & 0 & 1 & 0 & 0 \\ 0 & -2y & 0 & 1 & 0 \\ 0 & 0 & -2y_1 & 0 & 1 \end{pmatrix}$$

$$D\bar{F}_p - D\bar{F}_{p'} =$$

$$\begin{pmatrix} u_3 - u_3' + y - y' & x - x' & 0 & 0 & x - x' \\ 0 & u_1 - u_1' & y - y' & 2(u_2 - u_2') & 0 \\ -2(x - x') & 0 & 0 & 0 & 0 \\ 0 & 0 & -2(u_1 - u_1') & 0 & 0 \\ 0 & -2(y - y') & 0 & 0 & 0 \end{pmatrix}$$

$$(a+b)^2 \leq 2(a^2 + b^2)$$

$$\text{as } (a-b)^2 \geq 0$$

$$|D\bar{F}_p - D\bar{F}_{p'}| \leq$$

$$\left(6|x-x'|^2 + 7|y-y'|^2 + 5|u_1-u_1'|^2 + 4|u_2-u_2'|^2 + 2|u_3-u_3'|^2 \right)^{1/2}$$

$$\leq \sqrt{7} \left\| \begin{pmatrix} x-x' \\ y-y' \\ \vdots \end{pmatrix} \right\| = \sqrt{7} |p-p'|$$

Want to prove 4 statements

$$1) [D\bar{F}] \text{ at } x_i \text{ inv. so } h_i = -[D\bar{F} x_i]^{-1} f(x_i) \text{ well def}$$

$$2) \|h_i\| \leq \|h_0\|/2$$

$$3) \|f(x_i)\| \| [D\bar{F} x_i]^{-1} \|^2 \leq \|f(x_0)\| \| [D\bar{F} x_0]^{-1} \|^2$$

$$4) \|f(x_i)\| \leq \frac{M}{2} \|h_0\|^2$$

If 1, 2, 3 hold \rightarrow we can define h_i, x_i, u_i given by

$$x_{i+1} = x_i + h_i$$

$$u_{i+1} = \{x : |x - x_{i+1}| \leq \|h_i\|\}$$

so that each step i , hypotheses of thm hold.

From (2), $\sum_i \|h_i\|$ conv. (Compare with geo.)

Then $\{x_i\}$ conv. say to α .

From (4)

$$\begin{aligned} |f(x_i)| &\leq \frac{M}{2} \|h_{i+1}\|^2 \\ &\leq \frac{M}{2^i} \|h_0\|^2 \\ &\rightarrow 0 \text{ as } i \rightarrow \infty \end{aligned}$$

By cont., $f(\alpha) = 0$

11/28/2017

Kantorovich Thm on Global Conv.
of Newton's Method

$$f: \mathbb{R}^n \rightarrow \mathbb{R}^n$$

$$x_{n+1} = x_n - [DF(x_n)]^{-1} f(x_n)$$

Thm: $f: U \subset \mathbb{R}^n \rightarrow \mathbb{R}^n$ diff. mapping. $x_0 \in U$ & $[DF(x_0)]$ inv.

Define $h_0 = -[DF(x_0)]^{-1} f(x_0)$

$$x_1 = x_0 + h_0$$

$$U_1 = B_{\text{inv}}(x_1)$$

If $\bar{U}_1 \subset U$ if $[DF(x)]$ satisfies

Lipschitz & $|[DF(u_1)] - [DF(u_2)]| \leq M |u_1 - u_2|$

$u_1, u_2 \in \bar{U}_1$ & if

$$|f(x_0)| \cdot |[DF(x_0)]^{-1}|^2 M \leq 1/2$$

if satisfied then $f(x) = 0$ has unique solution in \bar{U}_1 & Newton's method with starting point x_0 converges to this solution.

Ex:

$$f\left(\begin{bmatrix} x \\ y \end{bmatrix}\right) = \begin{pmatrix} x^2 - y - 2 \\ y^2 - x - 6 \end{pmatrix}$$

Find Lipschitz constant

$$DF\left(\begin{bmatrix} x \\ y \end{bmatrix}\right) = \begin{pmatrix} 2x & -1 \\ -1 & 2y \end{pmatrix}$$

$$[DF\left(\begin{bmatrix} x_1 \\ y_1 \end{bmatrix}\right)] - [DF\left(\begin{bmatrix} x_2 \\ y_2 \end{bmatrix}\right)]$$

$$\begin{pmatrix} 2(x_1 - x_2) & 0 \\ 0 & 2(y_1 - y_2) \end{pmatrix}$$

\int_0 in norm....

$$\sqrt{4(x_1 - x_2)^2 + 4(y_1 - y_2)^2}$$

$$= 2 \left| \begin{pmatrix} x_1 \\ y_1 \end{pmatrix} - \begin{pmatrix} x_2 \\ y_2 \end{pmatrix} \right|$$

\uparrow
M

$$x_0 = \begin{pmatrix} 2 \\ 3 \end{pmatrix}, DF(x_0) = \begin{pmatrix} 4 & -1 \\ -1 & 6 \end{pmatrix}$$

$$[DF(x_0)]^{-1} = \frac{1}{23} \begin{pmatrix} 6 & 1 \\ 1 & 4 \end{pmatrix}$$

$$f(x_0) = \begin{pmatrix} -1 \\ 1 \end{pmatrix}$$

$$h_0 = \frac{1}{23} \begin{pmatrix} 6 & 1 \\ 1 & 4 \end{pmatrix} \begin{pmatrix} -1 \\ 1 \end{pmatrix} = \frac{1}{23} \begin{pmatrix} -5 \\ 3 \end{pmatrix}$$

$$U_1 = B_{\text{inv}}(x_1) = ; x_1 = x_0 + h_0$$

Kant. assumption?

$$\sqrt{2} \frac{1}{23} \sqrt{36 + 2 + 1} \cdot 2 = \frac{\sqrt{108}}{23} \cdot 2 = 0.288 < \frac{1}{2}$$

Want to prove the \Leftarrow

statements from before.

Lemma: $f: \mathbb{R}^n \rightarrow \mathbb{R}^n$ diff w/ deriv. satisfying

$$|DF(x) - DF(y)| \leq M |x - y| ; x, y \in \bar{U}$$

Then for $x, y \in \bar{U}$

$$|f(y) - f(x) - DF(x)(y-x)| \leq \frac{M}{2} |y-x|^2$$

PF: $x, y \in \bar{U}, h = y - x, g(t) = f(x + th)$

$$g(1) - g(0) = \int_0^1 g'(t) dt ; g'(t) = DF(x + th)h$$

$$f(y) - f(x) = \int_0^1 DF(x + th)h dt$$

$$g'(t) = DF(x)h + (DF(x + th)h - DF(x)h)$$

$$f(x+h) - f(x) = \int_0^1 Df(x) h dt + \int_0^1 Df(x+th) h - Df(x) dt$$

$$|f(x+h) - f(x) - Df(x)h| \leq \left| \int_0^1 Df(x+th) - Df(x) dt \right|$$

$$\leq \int_0^1 |Df(x+th) - Df(x)| dt$$

$$\leq \int_0^1 M |th|^2 dt = \frac{M}{2} |h|^2 \quad \square$$

Lem 2: $Df(x)$ inv.

$$|[Df(x_1)]^{-1}| \leq 2 |[Df(x_0)]^{-1}|$$

$$[Df(x_0)]^{-1} Df(x_1) \approx I$$

$$A = I - [Df(x_0)]^{-1} [Df(x_1)]$$

$$= [Df(x_0)]^{-1} [Df(x_0)] - [Df(x_0)]^{-1} [Df(x_1)]$$

$$= [Df(x_0)]^{-1} ([Df(x_0) - Df(x_1)])$$

$$|[Df(x_0)] - [Df(x_1)]| \leq M |x_0 - x_1|$$

$$\leq M |h_0|$$

$$|A| \leq |[Df(x_0)]^{-1}| |h_0| M$$

$$h_0 = -[Df(x_0)]^{-1} f(x_0)$$

$$|h_0| \leq |[Df(x_0)]^{-1}| \cdot |f(x_0)|$$

$$\text{Then } |A| \leq |[Df(x_0)]^{-1}|^2 |f(x_0)| M \leq \frac{1}{2}$$

$$I - A \text{ inv.} \rightarrow [Df] \text{ inv.}$$

$$\uparrow$$

$$[Df(x)] \text{ inv.}$$

$$B = (I - A)^{-1} = [Df(x_1)]^{-1} [Df(x_0)]$$

$$[Df(x_1)]^{-1} = B [Df(x_0)]^{-1}$$

$$= (I + A + A^2 + \dots) [Df(x_0)]^{-1}$$

$$|[Df(x_1)]^{-1}| \leq |I + A + A^2 + \dots| |[Df(x_0)]^{-1}|$$

$$\leq (1 + \frac{1}{2} + \frac{1}{4} + \dots) |[Df(x_0)]^{-1}|$$

$$= 2 |[Df(x_0)]^{-1}| \quad \square$$

Lem 3 $|f(x_1)| \leq \frac{M}{2} |h_0|^2$

Lem 1 gives

$$|f(x_1) - f(x_0) - [Df(x_0)]h_0| \leq \frac{M}{2} |h_0|^2$$

$$f(x_0) + [Df(x_0)](y_1 - x_0) = 0$$

$$h_0 = x_1 - x_0 = -[Df(x_0)]^{-1} f(x_0)$$

$$|h_1| \leq \frac{|h_0|}{2}$$

$$|h_1| \leq |[Df(x_1)]^{-1}| |f(x_1)|$$

$$\text{Lem 2} \rightarrow \leq |f(x_1)| \cdot 2 |Df(x_1)|$$

$$\leq \frac{M}{2} |h_0|^2 \cdot |Df(x_1)| |h_0|$$

$$\leq M |h_0| |[Df(x_0)]^{-1}|^2 |f(x_0)|$$

Wed Dec 6 1pm - 2pm pick up 'final exam'

11/30/2017

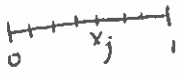
- Why Newton's method is important in cliff eq.
- How to use Newton to factor real quad (find quad factors).

Nonlinear alg system resulting from a nonlin. system b.v.p.

$$u''(x) = \underbrace{f(x, u)}_{\sin(u(x))}; \quad 0 \leq x \leq 1$$

$$u(0) = u(1) = 0$$

$$x_j = jh; \quad h = \frac{1}{n+1} \quad j=0, \dots, n+1$$



$$u''(x_j) \approx \frac{u(x_{j+1}) - 2u(x_j) + u(x_{j-1}))}{h^2}$$

$$\frac{1}{h^2} (u(x_{j+2}) - 2u(x_j) + u(x_{j-1}))) =$$

$$F(x_j, u(x_j)) + r(x_j, h)$$

goes to 0
as h → 0

$$u_j \equiv u(x_j)$$

$$u_{j+1} - 2u_j + u_{j-1} = h^2 f(x_j, u_j); \quad 1 \leq j \leq n$$

$$0 = \underbrace{\begin{pmatrix} 2 & -1 & & & \\ -1 & 2 & -1 & & \\ & & \ddots & \ddots & \\ & & & -1 & 2 \\ & & & & -1 & 2 \end{pmatrix}}_{F(u)} + h^2 \begin{pmatrix} f(x_1, u_1) \\ \vdots \\ f(x_n, u_n) \end{pmatrix}$$

$F: \mathbb{R}^n \rightarrow \mathbb{R}^n$

$$u = \begin{pmatrix} u_1 \\ \vdots \\ u_n \end{pmatrix}$$

$$0 = Au + \bar{F}(u) = F(u)$$

Suppose want $f(x)=0$,
Poly with real coefficients.

Want to find quad factors (will always have if deg > 1 but not nec. linear and can then use quad. form).

$$F(x) = Q^*(x) m^*(x)$$

$$m^*(x) = m^*(x, p^*, r^*) = x^2 - p^*x - r^*$$

$$m(x) = m(x, p, r) = x^2 - px - r$$

* $F(x) = m(x, p, r)Q(x, p, r) + q_1(p, r)x + q_2(p, r)$
 m^* quad. factor of f if $q_1(p^*, r^*) = 0$
 and $q_2(p^*, r^*) = 0$ to determine we need p^*, r^* such that

$$\begin{cases} q_0(p, r) = 0 \\ q_1(p, r) = 0 \end{cases}$$

Start at $\begin{pmatrix} p_0 \\ r_0 \end{pmatrix}$

$$\begin{pmatrix} p_{k+1} \\ r_{k+1} \end{pmatrix} = \begin{pmatrix} p_k \\ r_k \end{pmatrix} - \begin{pmatrix} \frac{\partial q_0}{\partial p}(p_k, r_k) & \frac{\partial q_0}{\partial r}(p_k, r_k) \\ \frac{\partial q_1}{\partial p}(p_k, r_k) & \frac{\partial q_1}{\partial r}(p_k, r_k) \end{pmatrix}^{-1} \begin{pmatrix} q_0(p_k, r_k) \\ q_1(p_k, r_k) \end{pmatrix}$$

How to compute Jacobian? Diff * wrt p or r

$$\begin{cases} 0 = \frac{\partial f}{\partial p} = \frac{\partial m}{\partial p} Q + m \frac{\partial Q}{\partial p} + \frac{\partial q_1}{\partial p} x + \frac{\partial q_0}{\partial p} \\ = -xQ + m \frac{\partial Q}{\partial p} + \frac{\partial q_1}{\partial p} x + \frac{\partial q_0}{\partial p} \\ 0 = \frac{\partial f}{\partial r} = \frac{\partial m}{\partial r} Q + m \frac{\partial Q}{\partial r} + \frac{\partial q_1}{\partial r} x + \frac{\partial q_0}{\partial r} \end{cases}$$

$$x Q(x, p, r) = m(x, p, r) \frac{\partial Q(x, p, r)}{\partial p} + \frac{\partial q_1(p, r)}{\partial p} x + \frac{\partial q_0(p, r)}{\partial p}$$

$$Q(x, p, r) = m(x, p, r) \frac{\partial Q(x, p, r)}{\partial r} + \frac{\partial q_1(p, r)}{\partial r} x + \frac{\partial q_0(p, r)}{\partial r}$$

We obtain $\frac{\partial q_1}{\partial p}, \frac{\partial q_1}{\partial r}, \frac{\partial q_0}{\partial p}, \frac{\partial q_0}{\partial r}$ by dividing xQ and Q by m .

Div. by quad. factor

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

$$Q(x) = \sum_{i=0}^{n-2} q_{i+2} x^i$$

$$m = x^2 - px - r$$

$$q_n = a_n$$

$$q_{n-1} = pq_n + a_{n-1}$$

$$q_i = pq_{i+1} + rq_{i+2} + a_i; \quad i = n-2, n-3, \dots, 0$$

Thm (superconv. of Newton's method)

$$\text{Let } K = |f(x_0)| | [DF(x_0)]^{-1} |^2 M < 1/2$$

$$\text{Let } c = \frac{1-K}{1-2K} | [DF(x_0)]^{-1} | \frac{M}{2}$$

$$\text{If } |h_n| \leq \frac{1}{2c} \rightarrow |h_{n+1}| \leq \frac{1}{c} \left(\frac{1}{2}\right)^{2^n}$$

Lem: If $K < 1/2$, if $|h_n| \leq 1/2c$ then $|h_{n+1}| \leq c |h_n|^2$

Pf (Thm using Lem): Denote $y_i = c |h_i|$

$$y_{i+1} = c |h_{i+1}| \leq c^2 |h_i|^2 = y_i^2$$

$$\text{If } |h_n| \leq 1/2c, \quad y_n = c |h_n| \leq c \cdot 1/2c = 1/2$$

$$y_{n+1} \leq y_n^2 \leq \left(\frac{1}{2}\right)^2$$

$$y_{n+2} \leq y_{n+1}^2 \leq \left(\frac{1}{4}\right)^2 = \left(\frac{1}{2}\right)^{2^2}$$

$$y_{n+m} \leq y_n^{2^m} \leq \left(\frac{1}{2}\right)^{2^m} \quad \square$$

Pf (Lem): From prev lem,

$$|f(x_i)| \leq \frac{M}{2} |h_{i-1}|^2$$

$$h_i = - [DF(x_i)]^{-1} f(x_i)$$

$$|h_i| \leq | [DF(x_i)]^{-1} | |f(x_i)|$$

$$\leq \frac{M}{2} | [DF(x_i)]^{-1} | |h_{i-1}|^2$$

Need bound for $| [DF(x_i)]^{-1} |$

Lem 5: If $K < 1/2$ then

$[DF(x_i)]^{-1}$ exists and satisfies

$$| [DF(x_i)]^{-1} | \leq | [DF(x_0)]^{-1} | \frac{1-K}{1-2K}$$

12/05/2017

Uniqueness in Newton's Method

$$f: \mathbb{R}^n \rightarrow \mathbb{R}^n$$

$$U \subset \mathbb{R}^n$$

$$x_0 \in U$$

$DF(x_0)$ invertible

$$h_0 = - [DF(x_0)]^{-1} f(x_0)$$

$$x_1 = x_0 + h_0$$

$$U_1 = B_{\text{int}}(x_1)$$

Assume DF Lipschitz (with M)

$$|f(x_i)| | [DF(x_0)]^{-1} |^2 M \leq 1/2$$

Then $f(x) = 0$ has unique solution in U_1 .

If $y \in U_1$ & $f(y) = 0 \rightarrow$

$$* |y - x_{i+1}| \leq \frac{1}{2} |y - x_i| \quad \text{Know } x_i \rightarrow \alpha$$

with $f(\alpha) = 0, \quad x_i \rightarrow \alpha$

$$|y - \alpha| \leq \frac{1}{2} |y - \alpha|$$

$$\frac{1}{2} |y - \alpha| \leq 0$$

$$y = \alpha.$$

$$*DF: f(y) = f(x_i) + [DF(x_i)](y - x_i) + r_i$$

$$-f(x_i) = [DF(x_i)](y - x_i) + r_i$$

$$y - x_i = -[DF(x_i)]^{-1} f(x_i) - [DF(x_i)]^{-1} r_i$$

$$y - x_{i+1} = -[DF(x_i)]^{-1} r_i$$

$$\text{lem 1: } |r_i| = |f(y) - f(x_i) - DF(x_i)(y - x_i)|$$

$$\leq \frac{M}{2} |y - x_i|^2$$

$$|y - x_i| \leq |DF(x_i)^{-1}| |r_i|$$

$$\leq |DF(x_i)^{-1}| \frac{M}{2} |y - x_i|^2$$

Now...

$$|y - x_i| \leq |DF(x_0)^{-1}| \frac{M}{2} |y - x_0|^2$$

$$= |DF(x_0)^{-1}| \frac{M}{2} |y - x_0| |y - x_0|$$

$$\leq |DF(x_0)^{-1}| \frac{M}{2} \cdot 2|h_0| \cdot |y - x_0|$$

(y in ball radius that centered at x_i so next 2|h_0| from x_0).

$$\leq |DF(x_0)^{-1}| M |h_0| |y - x_0|$$

$$\leq |DF(x_0)^{-1}|^2 M |f(x_0)| |y - x_0|$$

$\leq 1/2$
 \uparrow induction
 win

Assume induction up to $|y - x_j| \leq \frac{1}{2} |y - x_{j-1}|$

$$\frac{|y - x_{i+1}|}{|y - x_i|} \leq |DF(x_i)^{-1}| \frac{M}{2} |y - x_i|$$

$$\leq 2 |DF(x_{i-1})^{-1}| \frac{M}{2} |y - x_{i-1}|$$

$$\leq \frac{1}{2} M |y - x_0| |DF(x_0)^{-1}|$$

$$\leq \frac{1}{2} M 2|h_0| |DF(x_0)^{-1}|$$

$$\leq M |DF(x_0)^{-1}|^2 |f(x_0)| \leq 1/2$$

$$\text{If } K < \frac{1}{2} \text{ and } C = \frac{1-K}{1-2K} |DF(x_0)^{-1}| \frac{M}{2}$$

$$\text{If } |h_0| \leq \frac{1}{2C} \rightarrow |h_{n+1}| \leq \frac{1}{C} \left(\frac{1}{2}\right)^{n+1}$$

$$\text{Holdy if } |h_{i+1}| \leq C |h_i|^2$$

$$\text{lem 3 } |f(x_i)| \leq \frac{M}{2} |h_0|^2$$

From they

$$|f(x_i)| \leq \frac{M}{2} |h_{i-1}|^2$$

$$h_i = -[DF(x_i)^{-1}] f(x_i)$$

$$|h_i| \leq |DF(x_i)^{-1}| |f(x_i)|$$

$$\leq |DF(x_i)^{-1}| \frac{M}{2} |h_{i-1}|^2$$

\uparrow
Show $\leq C$

$$\text{WTS } |DF(x_i)^{-1}| \leq |DF(x_0)^{-1}| \frac{1-K}{1-2K}$$

$$|h_i| \leq |f(x_i)| |DF(x_i)^{-1}| \leq \frac{M}{2} |h_0|^2 2 |DF(x_0)^{-1}|$$

$$= \frac{M}{2} |h_0| |h_0| |DF(x_0)^{-1}| = |h_0| M |DF(x_0)^{-1}|^2 |f(x_0)|$$

$$\leq |f(x_0)| |DF(x_0)^{-1}| = K |h_0|$$

$$|h_1| \leq K |h_0|$$

$$K = |f(x_0)| |DF(x_0)^{-1}|^2 M$$

Assume $|h_i| \leq K |h_{i-1}|$

$$|x_n - x_0| = \left| \sum_{i=0}^{n-1} h_i \right| \leq \sum_{i=0}^{n-1} |h_i|$$

$$= |h_0| + |h_1| + \dots + |h_{n-1}|$$

$$\leq K |h_0| + K |h_0| + \dots + K^{n-1} |h_0|$$

$$= |h_0| (1 + K + \dots + K^{n-1})$$

$$|h_0| \frac{1 - K^n}{1 - K} \rightarrow |h_0| \frac{1}{1 - K}$$

hence
 \leq

$$A_n = I - DF(x_0)^{-1} DF(x_n) = DF(x_0)^{-1} (DF(x_0) - DF(x_n))$$

$$|A_n| \leq |DF(x_0)^{-1}| M |x_0 - x_n|$$

$$\leq |DF(x_0)^{-1}| M \frac{|h_0|}{1 - K}$$

$$\leq |DF(x_0)^{-1}| M \frac{|DF(x_0)^{-1}| |f(x_0)|}{1 - K}$$

$$= \frac{K}{1 - K} < 1 \text{ if } K < 1/2$$

$|A_n| < 1$. Hence, $I - A_n$ invertible.

$$(I - A_n)^{-1} = I + A_n + A_n^2 + \dots$$

$$(I - A_n)^{-1} = [DF(x_n)]^{-1} [DF(x_0)]$$

$$[DF(x_n)]^{-1} = (I + A_n + A_n^2 + \dots) [DF(x_0)]^{-1}$$

$$= \dots$$

$$|DF(x_n)^{-1}| \leq |DF(x_0)^{-1}| (1 + |A_n| + |A_n|^2 + \dots)$$

$$= |DF(x_0)^{-1}| \frac{1}{1 - |A_n|} \leq |DF(x_0)^{-1}| \frac{1}{1 - \frac{K}{1 - K}} = |DF(x_0)^{-1}| \frac{1 - K}{1 - 2K}$$