Adaptive Quadrature for Nyström



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Composite Newton-Cotes



Composite Gaussian



When should we be satisfied with our Quadrature?

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Consider error: $\epsilon_n \approx |Q_n - \int_a^b f(x) dx|$









Algorithm:

- 1. Set an error tolerance τ
- 2. Compute $Q_n = Quad(f(x), [a, b], n, \tau)$
- 3. If $\varepsilon_n > \tau$
 - Increase n
 - Loop to Step 2

This seems unintelligent

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Can we do better?









Algorithm:

- 1. Set an error tolerance τ , n usually 1
- 2. Compute $Q_n^{(i)} = Quad(f(x), [a, b], n, \tau)$
- 3. If $\varepsilon_n > \tau$ - Let $m = \frac{a+b}{2}$ - Compute $Q_n^{(i+1)} = Quad(f(x), [a, m], n, \frac{\tau}{2})$ $+Quad(f(x), [m, b], n, \frac{\tau}{2})$

Amazing Quadrature

Can we make this jump directly?



Error Estimation

Main approaches:

Error ~ 1. $|Q_n^{(i_1)} - Q_n^{(i_2)}|$ 2. $|Q_{n_1} - Q_{n_2}|$ 3. $|f^n(\xi)|$ 4. $|\tilde{c_n}|$

Gonnet - A Review of Error Estimation in Adaptive Quadrature, ACM Computing Surveys (2012)

Consider the Fredholm equation of the second kind

$$\varphi - A \varphi = f$$

Where the integral operator is

$$(A\varphi)(x) = \int_G K(x,y)\varphi(y)dy, \quad x \in G$$

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If we approximate A by some quadrature rule

$$(A_n\varphi)(x) = \sum_{k=1}^n w_k^{(n)} K(x, x_k^{(n)}) \varphi(x_k^{(n)})$$

We can solve the system at the quadrature points

$$\varphi_n - A_n \varphi_n = f_n$$
$$\downarrow$$
$$(\mathbf{I} - \mathbf{KW})\varphi = \mathbf{f}$$

Once we have the density at the quadrature points

$$\varphi = (\mathbf{I} - \mathbf{K}\mathbf{W})^{-1}\mathbf{f}$$

Use Nyström interpolation to get density for all $x \in G$ $\varphi_n(x) = f(x) + \sum_{k=1}^n w_k^{(n)} K(x, x_k^{(n)}) \varphi(x_k^{(n)})$

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For example if $\mathbf{x} \in \mathbb{R}^m$, we may interpolate by: $\varphi_{\mathbf{n}} = \mathbf{f_n} + \mathbf{KW}\varphi$

where $\mathbf{K} \in \mathbb{R}^{m \times n}$, $\mathbf{W} \in \mathbb{R}^{n \times n}$

Nyström Error Estimate

$||\varphi_n - \varphi|| \le C(||(A_n - A)\varphi|| + ||f_n - f||)$

Corollary 10.11, Kress – LIE (2nd ed.)

Nyström Error Estimate

$$\begin{aligned} ||\varphi_n - \varphi|| &\leq C(||(A_n - A)\varphi|| + ||f_n - f||) \\ & ||A_n\varphi_n - A_f\varphi_f|| \end{aligned}$$

Corollary 10.11, Kress – LIE (2nd ed.)

Example 1

Consider the following second kind integral equation

$$\varphi(x) - \frac{1}{2} \int_{-1}^{1} (x+1) e^{-xy} \varphi(y) dy = g(x)$$

Where

$$g(x) = e^{-x} - \frac{1}{2}(e^{x+1} - e^{-(x+1)})$$

We can compute the solution as

$$\varphi(x) = e^{-x}$$

Ex 1 Density Error: Gauss-Legendre



Ex 1 Density Approx: Gauss-Legendre



Ex 1 Density Error: Trapezoid Rule



Ex 1 Density Approx: Trapezoid Rule



Ex 1 Density Approx: Trapezoid Rule



Ex 1 Density Approx: Trapezoid Rule



Ex 1: Trapezoid Rule Differences



Example 2

Consider the following second kind integral equation

$$\varphi(x) - \frac{1}{2} \int_{-1}^{1} |x - y| f(y) dy = g(x)$$

Where

$$g(x) = e^x$$

We can compute the solution as

$$\varphi(x) = \frac{1}{2}xe^x + c_1e^x + c_2e^{-x}$$

Ex 1 Density Error: Gauss-Legendre



Ex 1: Gauss-Legendre Differences



Ex 1 Density Error: Trapezoid Rule



Ex 1: Trapezoid Rule Differences



Not so good =(

Can we do better?

Consider the constant

$$C = \frac{1 + ||(I - A)^{-1} A_n||}{1 - ||(I - A)^{-1} (A - A_n) A_n||}$$

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Remove the norms (abuse notation)

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Remove the norms (abuse notation)

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Let

$$B_n = I + (I - A)^{-1} A_n$$

$$S_n = (I - A)^{-1} (A - A_n) A_n$$

Our equation becomes

$$C = \frac{B_n}{I - S_n}$$

It turns out $C = I - A_n$

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Shuffling around

$$B_n(I - A_n) = I - S_n$$

From collective compactness, we have

$$||S_n|| \to 0 \quad n \to \infty$$

$$B_n(I - A_n) = I - S_n$$

Stare at this real hard and consider $n \to \infty$

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Stare at this real hard and consider $n \to \infty$ 2 things happen

$$B_n \to (I - A_n)^{-1}$$

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Stare at this real hard and consider $n \to \infty$ 2 things happen

$$B_n \to (I - A_n)^{-1}$$
$$A_n \to A$$

$$B_n(I - A_n) = I - S_n$$

Stare at this real hard and consider $n \to \infty$ 2 things happen

$$B_n \to (I - A_n)^{-1}$$

$$A_n \to A$$

$$\downarrow$$

$$B_n \to (I - A)^{-1}$$

Maybe take a look at $B_n = I + (I - A)^{-1} A_n$

Let's assume we have infinite computational power and compute directly

$$B_n = I + (I - A)^{-1}A_n$$

Look at

$$(I-A_n)^{-1}-B_n$$

Ex 1: Trapezoid, n = 11



Ex 1: Recall Trapezoid Rule Error



Just to check it works



n=26

n=51

How can we estimate this?

This is expensive!

Maybe $B_n - B_m$?

Ex 1: Estimator



Ex 1: Estimator Comparison



Real



Estimator

Ex 2: Trapezoid, n = 11



Ex 1: Trapezoid Rule Differences



Ex 2: Estimator



n = 11 vs n = 21

Ex 2: Estimator comparison





Estimator

Some Thoughts

(done verbally)

=)