

Overview of Nyström (and not-so-Nyström) high-order surface quadratures for fast solvers

Alex Barnett¹

SIAM CSE, Spokane, minisymposium MS355, 3/1/19

¹Center for Computational Mathematics, Flatiron Institute, Simons Foundation

Setting: solving linear BVPs

$Lu = 0$ in Ω , $\Omega \subset \mathbb{R}^d$, interior or exterior domain in $d = 2, 3$ dims

$u = f$, or $\partial u / \partial n = f$, or mix on boundary $\Gamma := \partial\Omega$ (& decay conds)

$L = 2^{nd}$ -order elliptic diff. op. whose fundamental soln G known

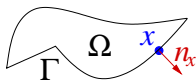
L usually constant-coeff. but need not be! (B–Nelson–Mahoney '15)

Apps: electrostatics, waves (EM/acoustic), fluids & vesicles, t-step heat

u scalar: Laplace, Helmholtz (& mod.), biharmonic (& mod.)

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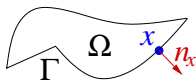
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- Convert to IE: e.g. "indirect", interior Dirichlet Lap. 2D, unknown "density" τ on Γ

$$u(x) = (\mathcal{D}\tau)(x) := \int_{\Gamma} \frac{\partial G(x,y)}{\partial n_y} \tau(y) ds_y \quad G(x,y) = \log(1/r)/2\pi, \quad r = \|x - y\|$$

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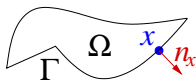
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jump relation: $u^-(x) := \lim_{h \rightarrow 0^+} u(x - hn_x) = ((D - \frac{1}{2})\tau)(x) = f(x) \quad x \in \Gamma$

(Also \exists "direct" formulation: adjoint BIE, physical unknown, RHS more painful)

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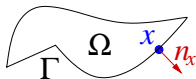
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Fredholm 2^{nd} -kind BIE on Γ : $(I + K)\tau = f$ \leftarrow task: approx. this by lin. sys.

$K(x,y)$ is as smooth as Γ , plus weak (i.e. integrable) singularity at $x = y$:

If Γ smooth: $\log r$ or $r^2 \log r$ in $d = 2$; r^{-1} in $d = 3$

Two routes to represent density τ

(G) global spectral accuracy, convergence via degree $p \rightarrow \infty$, $\text{err} \sim c^{-p}$, $N \sim p^{d-1}$

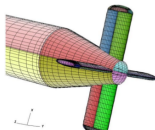


peri. trap. rule



sph. harms.

(Rahimian et al)



macro Cheby. patches

(Bruno, Turc et al.)

obstacles simple, smooth

some adaptivity poss. (Kress corners)

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underlying basis: Fourier / sph. harm.

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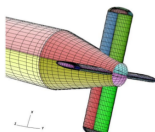
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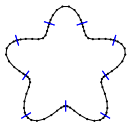
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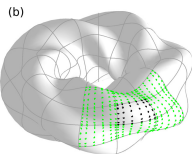
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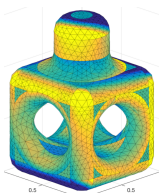
(L) local fixed panel order p , conv. via $h \rightarrow 0$, $\text{err} \mathcal{O}(h^p)$. $N \sim h^{1-d}$



G-L panels



G-L quads
(B et al. '19)



4th-ord tri's
(O'Neil '17)

adaptivity and/or CAD geoms.

can split any panel indep. of others

$h \rightarrow 0$ with a CAD mesh? we wish

recent tri nodes:

Vioreanu-Rokhlin



Well-cond $p^{d-1} \times p^{d-1}$ matrices map values at nodes \leftrightarrow basis coeffs

Two religions of discretizing the IE $(I + K)\tau = f$

- Nyström, impose IE at nodes $\{x_i\}_{i=1}^N$: $\tau_i + (K\tau)(x_i) = f_i \quad i = 1, \dots, N$
 N $(d - 1)$ -dim singular ints, using τ interp from $\{\tau_j\} := \{\tau(x_j)\}$ ↗

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(G) global 3D: smoothly deformed sphere, cost $\mathcal{O}(p^5)$ (Graham–Sloan '02)

Stokes: vesicles, red blood cells

(Rahimian, Veerapaneni, Biros, ...)

smooth bodies, Helmholtz, Maxwell

(Ganesh, Hawkins, ...)

(L) tri/quad panels: many 4D quadrature rules

(Sauter–Schwab, Ch. 5)

software, $p = 0, 1$ & Maxwell RWG: BEM++

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Which use? At same order, accuracy basically same

(Kress '99, etc)

Galerkin: nastier numerical integrals, slower set-up

(Canino '98)

Galerkin more mature convergence theory, industrial codes

Thus, the rest of this review is about Nyström...

Quadrature task & categories

Recall Nyström: $\tau_i + (K\tau)(x_i) = f_i$, surface nodes $i = 1, \dots, N$

Task: given vector $\tau := \{\tau_j\}_{j=1}^N$, eval. $(K\tau)(x_i)$ at all N targets x_i

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$$\int_{\Gamma} g(x) ds_x \approx \sum_{j=1}^N w_j g(x_j) \quad \text{holds to order } \geq p, \text{ for } g \in C^\infty(\Gamma)$$

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Apart from (G) vs (L) for density τ rep, other axes to categorize. . .

- FMM/FDS-compatible? only $\mathcal{O}(N)$ els. differ from native $A_{ij} = K(x_i, x_j)w_j$
- Precomputation (store A ; good for rigid body) vs on-the-fly? (moving geoms.)
- Solely Nyström (on-surf.) task, vs also bonus off-surf. target evals.?
- Needs only on-surf. geom, vs also needs off-surf. pts?
- Ease of switching to new kernels? e.g. toroidal . . .

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2D: product quadratures, exact for freqs. up to $\pm N/2$ (Kress '91)

split $K(t, s) = \psi(t, s) \log(4 \sin^2 \frac{1}{2}|t - s|) + \phi(t, s)$, for some $\psi, \phi \in C^\infty([0, 2\pi)^2)$

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Related: spectral close off-surf. eval, $u = \text{Re}$ (Cauchy int. in barycentric form)

(Ioakimidis '91; DLP Helsing '08, SLP B-Veerapaneni-Wu '14)

Helm+Lap: MPSpack (B '09), Stokes+Lap (FMM'ed Kress!): pybie2d (Stein '18)

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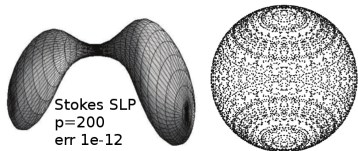
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3D: diffeo. of sphere: $\mathcal{O}(N^2 \log N)$ recall $N = p^2$ (Gimbutas-Veerapaneni '14)



grid vals \leftrightarrow sph harms fast

(azimuthal 1D FFT)

eval sph harms at p^4 rot grid pts

(elevational 1D NUFFT)

A.(G) on-surf. global [cont.]

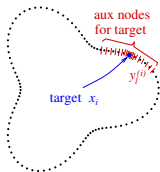
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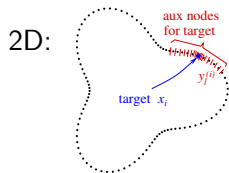


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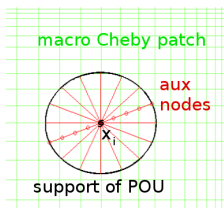
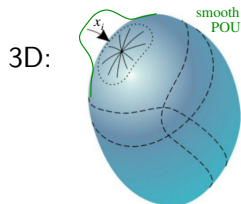
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polar idea:

$1/r$ cancelled by metric $rdrd\theta$

note $\lim_{r \rightarrow 0} rK(x_i, y(r, \theta))$ varies with θ !

partition of unity, polar aux grid

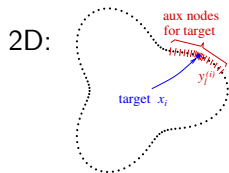
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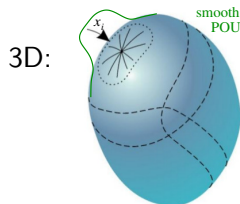
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macro Cheby patch



support of POU

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How get τ at aux nodes? local p -order Lagrange interp.

Thus: (near-diag blk of A) \approx (kernel eval at aux nodes) \times (interp matrix)

A.(L) on-surf. for local panel density



Common idea in $d = 2, 3$ for p th order scheme:

- special rules for near (self+nei) panels
- native rule $A_{ij} = K(x_i, x_j)w_j$ for far FMM-compat.

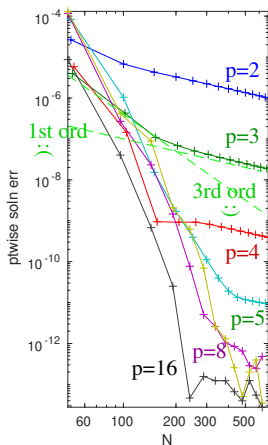
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Here test soln err, Helm CFIE

freq $k = 2$, std level-restricted splits

Why bottom out at 1st-order?

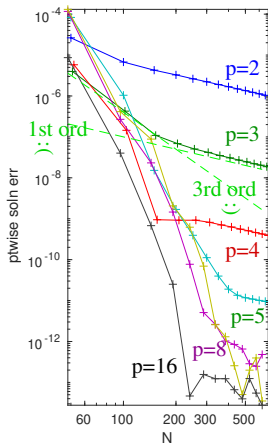
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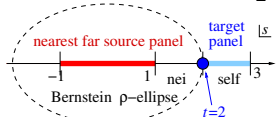
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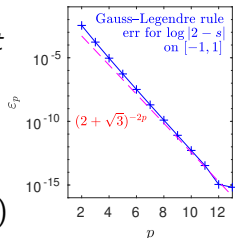
“nearest far panel” (NFP) controls *native* err:

$$G-L \ \varepsilon_p \sim \rho^{-2p}, \quad \frac{\rho + \rho^{-1}}{2} = t$$



scheme formally $\mathcal{O}(h^p + \varepsilon_p h)$

Thus my rule of thumb: pick $p \approx \#$ digits you seek



With that caveat ... on-surf. panel schemes: two types

A.(L).i: analytic

2D: complex Cauchy integral ψ, ϕ formulae avail. (Helsing-Ojala '08)

Ideas: monomial rep. $\tau(z) = \sum_{n=0}^{p-1} c_n z^n$, $z \in \mathbb{C}$ (prefer $\tau(s)$: af Klinteberg)

2-term recurrence for $\int_{-1}^1 \frac{y^n}{x-y} dy$, then e.g. DLP $(D\tau)(x) = \operatorname{Re} \frac{1}{2\pi i} \int_{\Gamma} \frac{\tau(y)}{x-y} dy$

Bonus: great for close off-surf. eval. code: `demo*.m` (Helsing)

3D: singularity subtraction? in infancy, hard (Helsing note arXiv:1301.7276)

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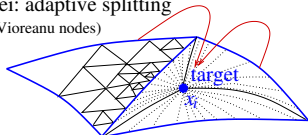
A.(L).ii: aux-node based analytic split not needed

2D: good aux nodes by “gen. Gauss. quadr.” (Rokhlin, Xiao, Gimbutas, ...)

3D: high- p tri's (Bremer–Gimbutas '13)

nei: adaptive splitting

(Vioreanu nodes)



self: radial aux nodes

custom gen Gauss quad in θ

reach 10^{-12} , even v. high aspect ratio

∋ Fortran (Gimbutas, Rachh in prep.)

With that caveat ... on-surf. panel schemes: two types

A.(L).i: analytic

2D: complex Cauchy integral ψ, ϕ formulae avail. (Helsing–Ojala '08)

Ideas: monomial rep. $\tau(z) = \sum_{n=0}^{p-1} c_n z^n$, $z \in \mathbb{C}$ (prefer $\tau(s)$: af Klinteberg)

2-term recurrence for $\int_{-1}^1 \frac{y^n}{x-y} dy$, then e.g. DLP $(D\tau)(x) = \text{Re} \frac{1}{2\pi i} \int_{\Gamma} \frac{\tau(y)}{x-y} dy$

Bonus: great for close off-surf. eval. code: demo*.m (Helsing)

3D: singularity subtraction? in infancy, hard (Helsing note arXiv:1301.7276)

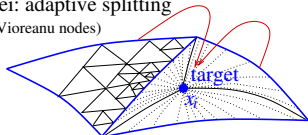
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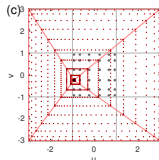


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radial aux scheme for 3×3 near patch:



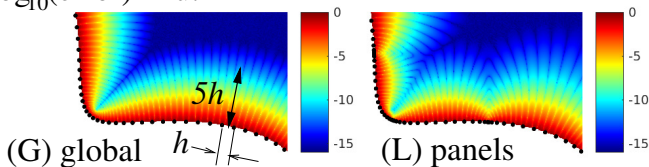
not versatile yet: torus diffeo's only

(wave eqn: B–Hagstrom–Greengard '19)

Interlude: close evaluation task

Recall native rule, off-surf. eval. $u(x) \approx \sum_{j=1}^N K(x, x_j) w_j \tau_j$ $x \in \Omega$

$\log_{10}(\text{error})$ in u :



How accurate is it? Exponential in N , but rate depends on target x :

Thm: (B '14) For global peri. trap. rule, analytic curve,
rate = Im (preimage of x under complexification of Γ param.)

• Similar estimates for panels (af Klinteberg–Tornberg '17)

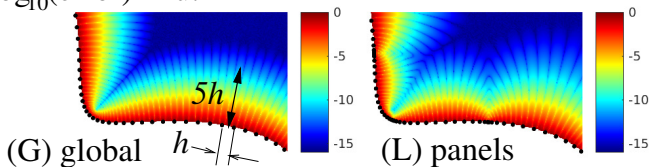
2D summary: $\text{err} \approx \mathcal{O}(e^{-2\pi d/h})$ $d = \text{dist to surf}$, $h = \text{local node spacing}$

“5h rule”: $d \geq 5h$ gets you 10^{-14} , closer and lose digits linearly

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Idea: native eval to points *near* Γ , then extrapolate *back* to target on Γ

(Yes, sounds crazy. Bonus: also does close-eval task!)

Let's call idea CATEGORY B: off-surface methods

Method B.1: “Hedgehog” quadrature

Originally scheme for eval at *close* target x : (Ying–Biros–Zorin '06; Quai...)

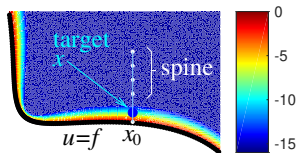
- i) pick line (“spine”) through x hitting Γ at x_0 , near
- ii) upsample τ by factor $\beta > 1$ in each dim, e.g. $\beta = 2-4$
- iii) eval at few pts dist $\geq 5h/\beta$ via upsampled native rule
- iv) interpolate to x , from these pts *plus known* $u(x_0) = f(x_0)$

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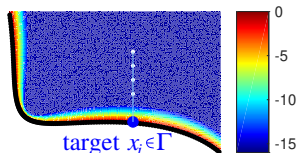
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Interpolate for close eval:



Can also extrapolate for Nyström surf quadr:

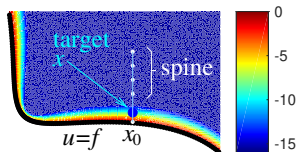


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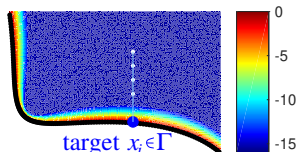
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Interpolate for close eval:



Can also extrapolate for Nyström surf quadr:



Adv: PDE-indep, dim-indep, τ -rep-indep, FMM'able... but params to adjust

movie: 10^4 vesicles + quad panels: $N = 6 \times 10^6$

(Morse–Lu–Rahimian–Zorin, in prep)

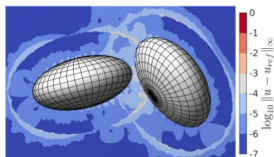
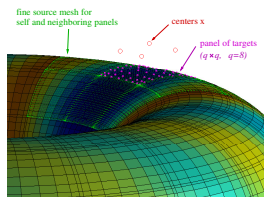
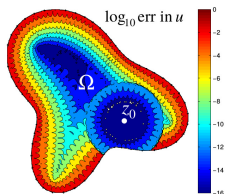
- Note: all category B methods eval $(I + A)\tau$, not $A\tau$
→ need 2-sided average, explicit I , to avoid GMRES stagnation

B.2: QBX (quadrature by expansion)

Finally (!) use the fact: u satisfies the PDE $Lu = 0$

Idea: eval. "local exp." $u(z) = \operatorname{Re} \sum_{n=0}^P a_n (z - z_0)^n$ here 2D Lap. case

- center $z_0 \in \Omega$, pick e.g. $3h$ from Γ
 - each a_n given by a surf int (addition thm): use β -upsampled native rule
- "Global" all of Γ , vs "local" just near panels (Klöckner–B–O'Neil–Greengard, '13; B '14)



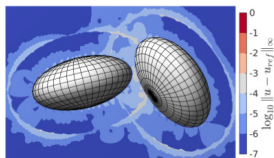
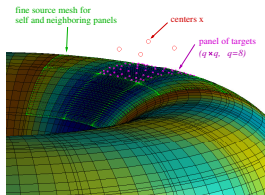
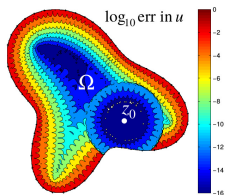
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(b) Error using QBX.

Recent variants:

- AQBX: automated target-wise choice of p, β (af Klinteberg–Tornberg '17)
- QBKIX: PDE-indep, via proxy sources (Rahimian–B–Zorin '17)
- 3D Line QBX: p not p^2 terms, closer to hedgehog (Siegel–Tornberg '18)
- GIGAQBX: integrate w/ FMM: `pyntial` (Wala–Klöckner, '17, '18)

Promise: prescribed-tolerance black-box A apply or A_{ij} fill

not yet

Omitted topics

- corners and edges either geometric refinement, or make custom quadr, or both
(Chandler–Graham) (Helsing) (Serkh–Rokhlin) (Lintner–Bruno)
- various special methods (Slevinsky–Olver '15) (Carvalho–Khatrı–Kim, '18)
- analysis
- transmission BVPs, diel. contrast, D-N junctions
- bodies of revolution curves w/ toroidal kernels
- line integrals in 3D (Tornberg–Shelley '04; af Klinteberg–B in prep.)
- fundamental solutions (MFS) as alternative to SKIE+Nyström ...

State of the art & community to do list

- 2D: use Alpert/Kress if global, Helsing if panels; speed faster than FMM
corners: RCIP (Helsing), or (Serkh, Hoskins–Rachh) if analysis avail, $N/\text{corner} \approx 40$
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Lots of fun challenges:

- 3D speed currently $10^2\text{--}10^3$ targs/sec/core err 10^{-6} : $\geq 10\times$ slower than 3D FMM
- automatic apply of high-order Nyström to CAD/industrial meshes
- 3D corners, cones, and (generic curving) edges, to high order
- related: high aspect ratio / skew panels
- fair error/speed comparisons on 3D test probs (2D basically benchmarked)
we're starting to address: needs uniform code interface (O'Neil–Rachh–B)
- 2D/3D documented code for non-experts w/ sensible/adaptive params

Review references

R. Kress, *Linear Integral Equations*, '99

Colton–Kress, *Inverse Acoustic & Electromagnetic Scattering Theory*, '89

I. Sloan, *Acta Num.* '91

J. Helsing, RCIP tutorial, arXiv:1207.6737 '18

K. E. Atkinson book, '97

Hao–B–Martinsson–Young, *Adv. Comput. Math.* '14

<http://math.dartmouth.edu/~fastdirect/notes/quadr.pdf>