

## Today

- Other PDEs
- Quadrature

## Announcements

- Extra Office Hours
- Presentations
- Projection methods section rewritten

## 'Off-the-shelf' ways to compute integrals

How do I compute an integral of a nasty singular kernel?

Symbolic integration

brittle

Why not Gaussian?

$$\| \text{phys} | \int_{\Omega} \omega p(\cdot) | \leq C h^2 \| f^{(2)} \|_{\infty}$$




$$\int_{\Omega} \log(|x-y|) \sigma(y) dy$$

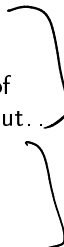
$$= \int_0^1 \log(x - \gamma(t)) \sigma(\gamma(t)) \gamma'(t) dt$$



# Singular and Near-Singular Quadrature

$$\int_0^1 \log(x-f(t)) \cdot g(t)/g'(t) dt$$


Numerically distinct scenarios:

- ▶ Near-Singular quadrature
    - ▶ Integrand nonsingular
    - ▶ But may locally require lots of
    - ▶ Adaptive quadrature works, but...
  - ▶ Singular quadrature
    - ▶ Integrand singular
    - ▶ Conventional quadrature fails
- 

## Kussmaul-Martensen quadrature

$$\int_0^1 \log(x - \gamma(t)) \sigma(\gamma(t)/\gamma'(t))$$

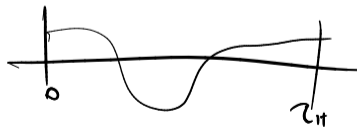
Theorem (A special integral [Kress LIE Lemma 8.21])

$$\frac{1}{2\pi} \int_0^{2\pi} \log\left(4 \sin^2 \frac{t}{2}\right) e^{imt} dt = \begin{cases} 0 & m = 0, \\ -\frac{1}{|m|} & m = \pm 1, \pm 2, \dots \end{cases}$$

Why is that exciting?

**Demo:** Kussmaul-Martensen quadrature

$$\sigma(x) = \sum \alpha_m e^{imx}$$

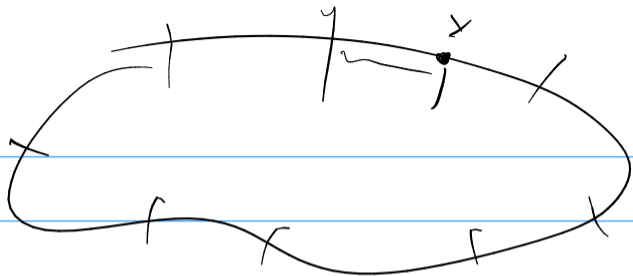


## Singularity Subtraction

$$\begin{aligned} & \int \langle \text{Thing } X \text{ you would like to integrate} \rangle \\ &= \int \langle \text{Thing } Y \text{ you can integrate} \rangle \\ &+ \int \langle \text{Difference } X - Y \text{ which is } \underbrace{\text{easy to integrate (numerically)}}_{\text{smooth}} \rangle \end{aligned}$$

Give a typical application.

Drawbacks?



## High-Order Corrected Trapezoidal Quadrature

- ▶ Conditions for new nodes, weights  
( $\rightarrow$  linear algebraic system, dep. on  $n$ )  
to integrate

$$\langle \text{smooth} \rangle \cdot \langle \text{singular} \rangle + \langle \text{smooth} \rangle$$

- ▶ Allowed singularities:  $|x|^\lambda$  (for  $|\lambda| < 1$ ),  $\log |x|$
- ▶ Generic nodes and weights for log singularity
- ▶ Nodes and weights copy-and-pasteable from paper

[Kapur, Rokhlin '97]  $\leftarrow$

Alpert '99 conceptually similar:

# Generalized Gaussian

- ▶ “Gaussian”:
  - ▶ Integrates  $2n$  functions exactly with  $n$  nodes
  - ▶ Positive weights
- ▶ Clarify assumptions on system of functions (“Chebyshev system”) for which Gaussian quadratures exist
- ▶ When do (left/right) singular vectors of integral operators give rise to Chebyshev systems?
  - ▶ In many practical cases!
- ▶ Find nodes/weights by Newton’s method
  - ▶ With special starting point
- ▶ Very accurate
- ▶ Nodes and weights for download

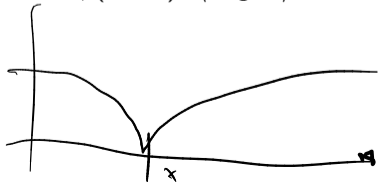
[Yarvin/Rokhlin '98]



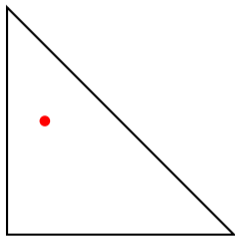
## Singularity cancellation: Polar coordinate transform

$$\begin{aligned} & \int \int_{\partial\Omega} K(\mathbf{x}, \mathbf{y}) \phi(\mathbf{y}) ds_{\mathbf{y}} \\ &= \\ & \int_0^R \int_{\mathbf{x}+\mathbf{r} \in \partial\Omega \cap \partial B(\mathbf{x}, r)} K(\mathbf{x}, \mathbf{x} + \mathbf{r}) \phi(\mathbf{x} + \mathbf{r}) d\langle \text{angles} \rangle r dr \\ &= \\ & \int_0^R \int_{\mathbf{x}+\mathbf{r} \in \partial\Omega \cap \partial B(\mathbf{x}, r)} \frac{K_{\text{less singular}}(\mathbf{x}, \mathbf{x} + \mathbf{r})}{r} \phi(\mathbf{x} + \mathbf{r}) d\langle \text{angles} \rangle r dr \end{aligned}$$

where  $K_{\text{less singular}} = K \cdot r$ .

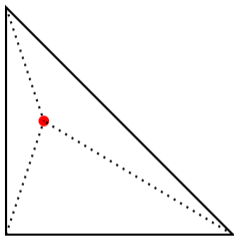


## Quadrature on Triangles



**Problem:** Singularity can sit *anywhere* in triangle  
→ need *lots* of quadrature rules (one per target)

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## Kernel regularization

Singularity makes integration troublesome: *Get rid of it!*

$$\frac{\dots}{\sqrt{(x-y)^2}} \rightarrow \frac{\dots}{\sqrt{(x-y)^2 + \epsilon^2}}$$

Use Richardson extrapolation to recover limit as  $\epsilon \rightarrow 0$ .

(May also use geometric motivation: limit along line towards singular point.)

Primary drawbacks:

- ▶ Low-order accurate
- ▶ Need to make  $\epsilon$  smaller (i.e. kernel more singular) to get better accuracy

Can take many forms—for example:

- ▶ Convolve integrand to smooth it  
( $\rightarrow$  remove/weaken singularity)
- ▶ Extrapolate towards no smoothing

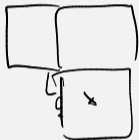
Related: [Beale/Lai '01]

## Acceleration and Quadrature

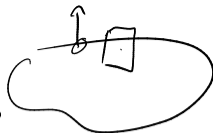


$$\oint_{\Gamma} \left( \frac{1}{2} \sigma \right) \sigma(x) = \int_{\Gamma} \partial_n G(x-y) \delta(y) dy$$

How can singular quadrature and FMM acceleration be made compatible?



## FMMs and other Layer Potentials



How does an FMM evaluate a double layer?

compute multipoles of dip kernel

How does an FMM evaluate  $S'$ ?

eval  $\nabla$  of local expansions at the end

What effect does this have on accuracy?

$D$  does not lose an order  
 $S'$  loses an order

# Outline

Introduction

Dense Matrices and Computation

Tools for Low-Rank Linear Algebra

Rank and Smoothness

Near and Far: Separating out High-Rank Interactions

Outlook: Building a Fast PDE Solver

Going Infinite: Integral Operators and Functional Analysis

Singular Integrals and Potential Theory

Boundary Value Problems

Back from Infinity: Discretization

**Computing Integrals: Approaches to Quadrature**

A Bag of Quadrature Tricks

**Quadrature by expansion ('QBX')**

QBX Acceleration

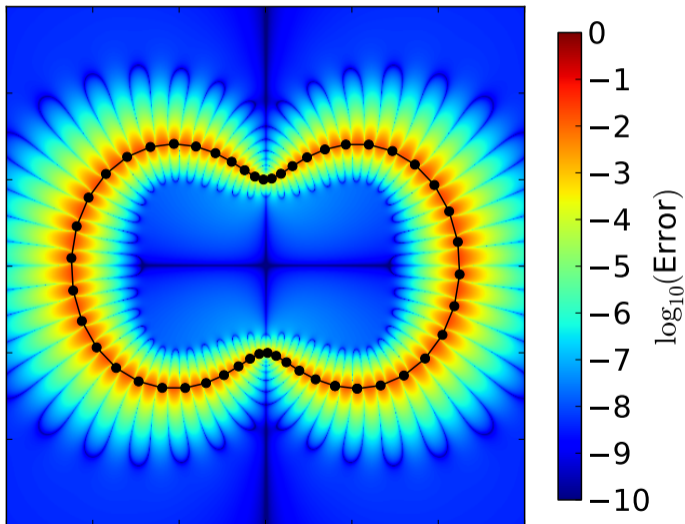
Reducing Complexity through better Expansions

Results: Layer Potentials

Results: Poisson

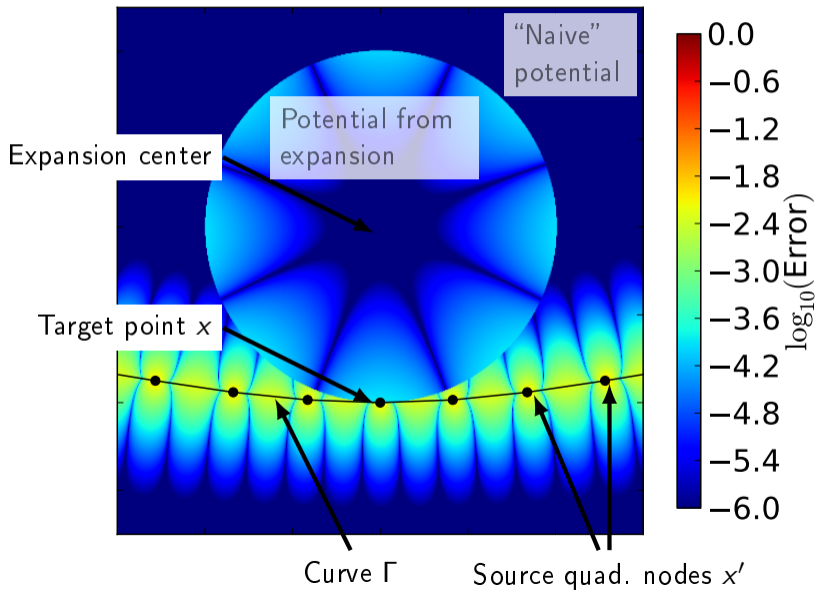
Going General: More PDEs

## Layer Potential Evaluation: Some Intuition



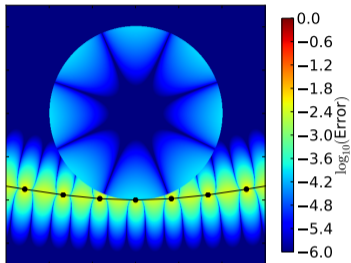


# QBX: Idea



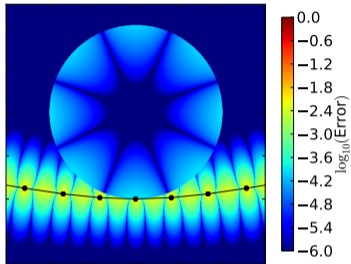
# QBX: An Experiment

$p = 3, N = 80$

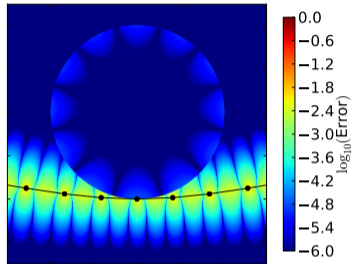


# QBX: An Experiment

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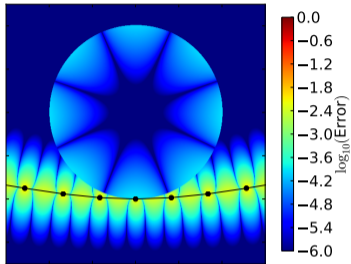


$p = 6, N = 80$

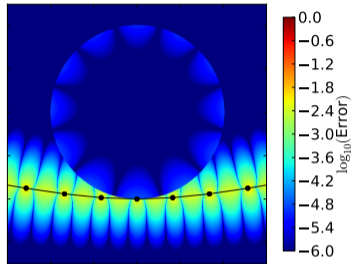


# QBX: An Experiment

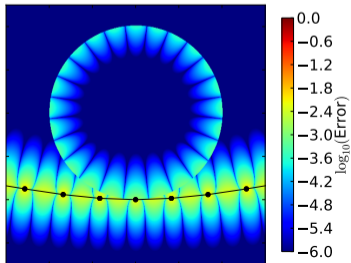
$p = 3, N = 80$



$p = 6, N = 80$

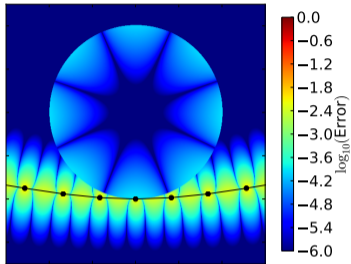


$p = 12, N = 80$

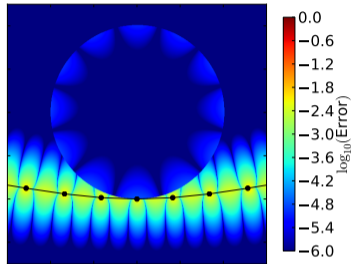


# QBX: An Experiment

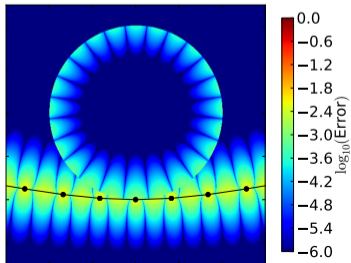
$p = 3, N = 80$



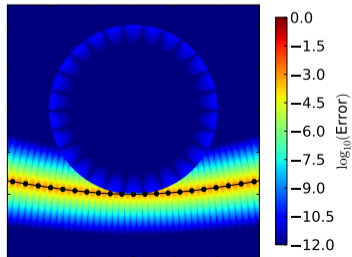
$p = 6, N = 80$



$p = 12, N = 80$

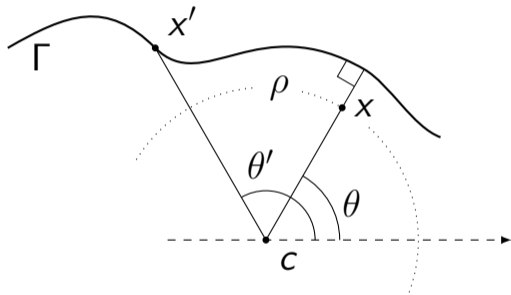


$p = 12, N = 240$

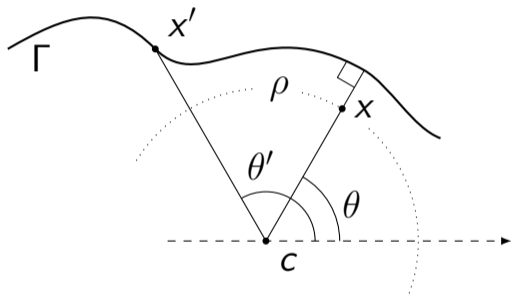


## QBX: Notation, Basics

Graf's addition theorem



Requires:  $|x - c| < |x' - c|$  ("local expansion")



$$H_0^{(1)}(k|x - x'|) = \sum_{l=-\infty}^{\infty} H_l^{(1)}(k|x' - c|) e^{il\theta'} J_l(k|x - c|) e^{-il\theta}$$

## QBX: Formulation, Discretization

Compute layer potential on the disk as

$$S_k \sigma(x) = \sum_{l=-\infty}^{\infty} \alpha_l J_l(k\rho) e^{-il\theta}$$

with

$$\alpha_l = \left( \frac{i}{4} \int_{\Gamma} H_l^{(1)}(k|x' - c|) e^{il\theta'} \sigma(x') dx' \quad (l = -\infty, \dots, \infty) \right)$$

$S\sigma$  is a smooth function up to  $\Gamma$ .



## QBX: Formulation, Discretization

$$S_{\text{QBX}} = \sum \int H_e(\dots) \sigma(y) dy$$

Compute layer potential on the disk as

$$S_k \sigma(x) = \sum_{l=-p}^p \alpha_l J_l(k\rho) e^{-il\theta} = \int \underbrace{\sum H_e(\dots)}_{\text{continuous}} \sigma(y) dy$$

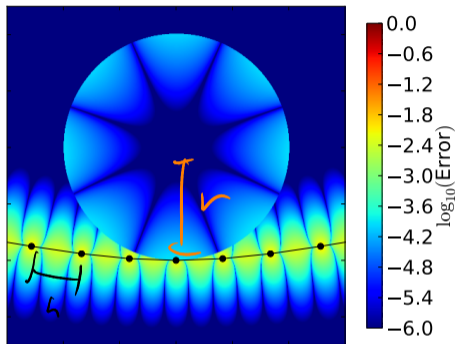
with

$$\alpha_l = \frac{i}{4} T_N \left( \int_{\Gamma} H_l^{(1)}(k|x' - c|) e^{il\theta'} \sigma(x') dx' \right) \quad (l = -\infty, \dots, \infty)$$

$S\sigma$  is a smooth function up to  $\Gamma$ .

# Quadrature by Expansion (QBX)

$r \sim h$   
 $r = 2h$



$$\text{Error} \leq \left( C \underbrace{r^{p+1}}_{\substack{\text{Truncation error} \\ h^{p+1}}} + C \underbrace{\left(\frac{h}{r}\right)^q}_{\substack{\text{Quadrature error} \\ + \epsilon}} \right) \|\sigma\|$$





$\left(\frac{1}{2}\right)^9$

[K, Barnett, Greengard, O'Neil JCP '13]

## Achieving high order

$$\text{Error} \leq \left( C \underbrace{r^{p+1}}_{\text{Truncation error}} + C \underbrace{\left(\frac{h}{r}\right)^q}_{\text{Quadrature error}} \right) \|\sigma\|$$

Two approaches:

- ▶ *Asymptotically convergent*:  $r = \sqrt{h}$ 
  - ▶  Error  $\rightarrow 0$  as  $h \rightarrow 0$
  - ▶  Low order:  $h^{(p+1)/2}$
- ▶ *Convergent with controlled precision*:  $r = 5h$ 
  - ▶  Error  $\not\rightarrow 0$  as  $h \rightarrow 0$
  - ▶  High order:  $h^{p+1}$  to controlled precision  $\epsilon := (1/5)^q$

## Other layer potentials

Can't just do single-layer potentials:

$$\alpha_l^D = \frac{i}{4} \int_{\Gamma} \frac{\partial}{\partial \hat{n}_{x'}} H_l^{(1)}(k|x' - c|) e^{i\theta'} \mu(x') dx'.$$

Even easier for target derivatives ( $S'$  et al.): **Take derivative of local expansion.**

**Analysis says:** Will lose an order.

**Slight issue:** QBX computes one-sided limits.

Fortunately: Jump relations are known—e.g.

$$(PV)D^* \mu(x)|_{\Gamma} = \lim_{x^{\pm} \rightarrow x} D\mu(x^{\pm}) \mp \frac{1}{2} \mu(x).$$

*Alternative:* Two-sided average → Preferred because of conditioning