

# High Frequency Scattering by Convex Polygons

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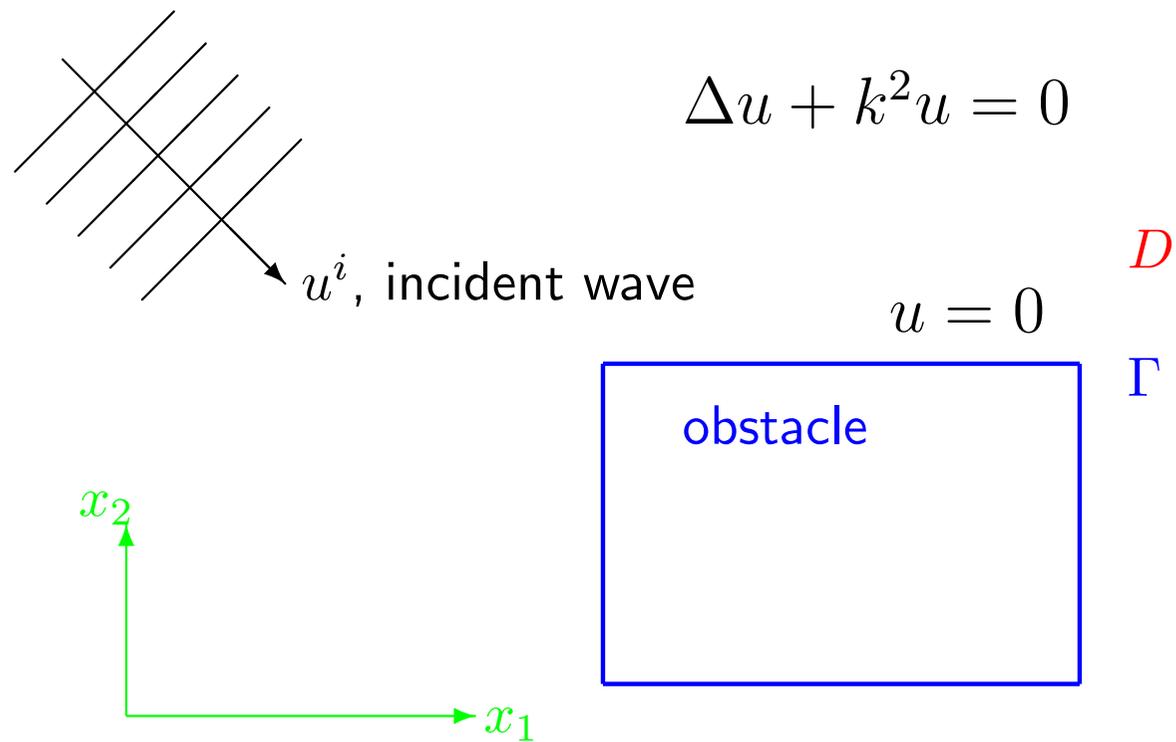
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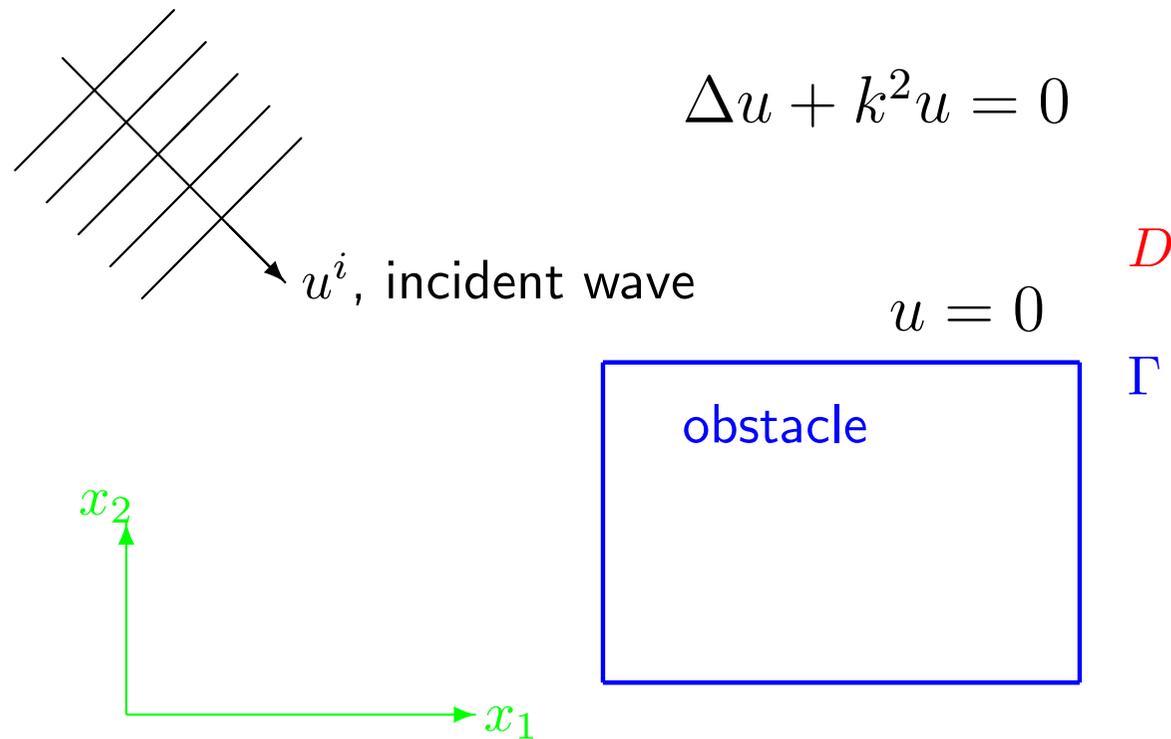
**University of Maryland, September 2005**

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## The Scattering Problem

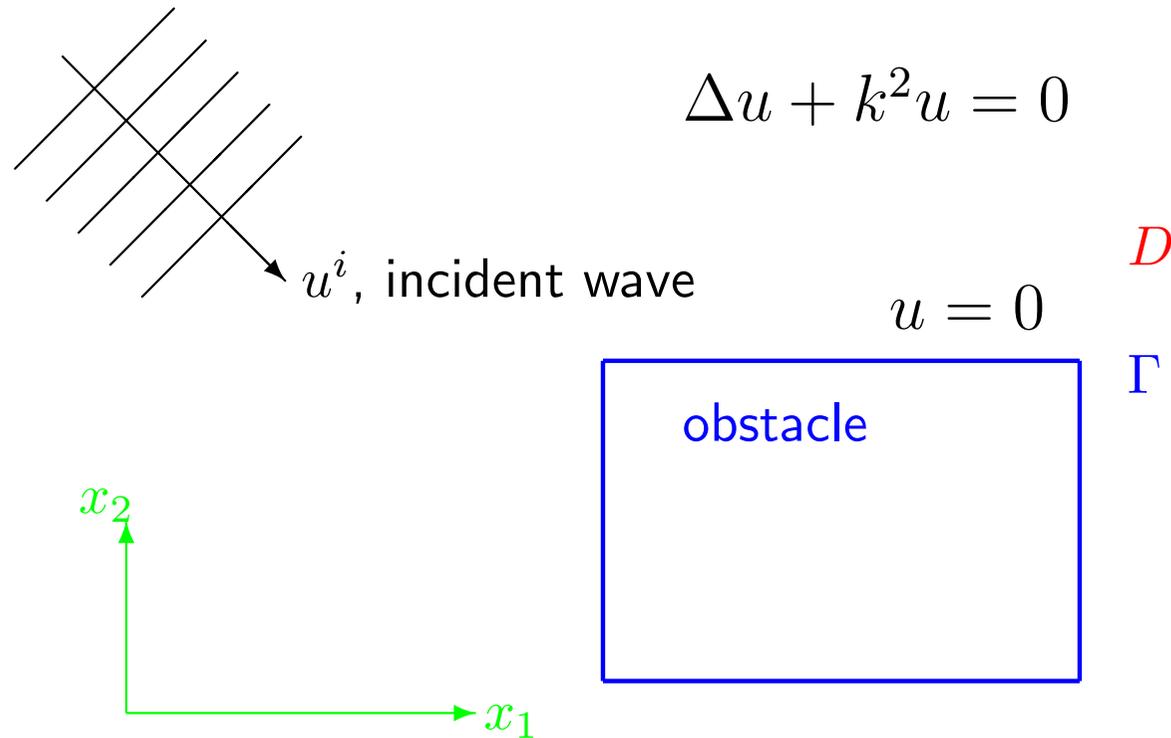




Green's representation theorem:

$$u(x) = u^i(x) - \int_{\Gamma} \Phi(x, y) \frac{\partial u}{\partial n}(y) ds(y), \quad x \in D,$$

where  $\Phi(x, y) := \frac{i}{4} H_0^{(1)}(k|x - y|)$ .

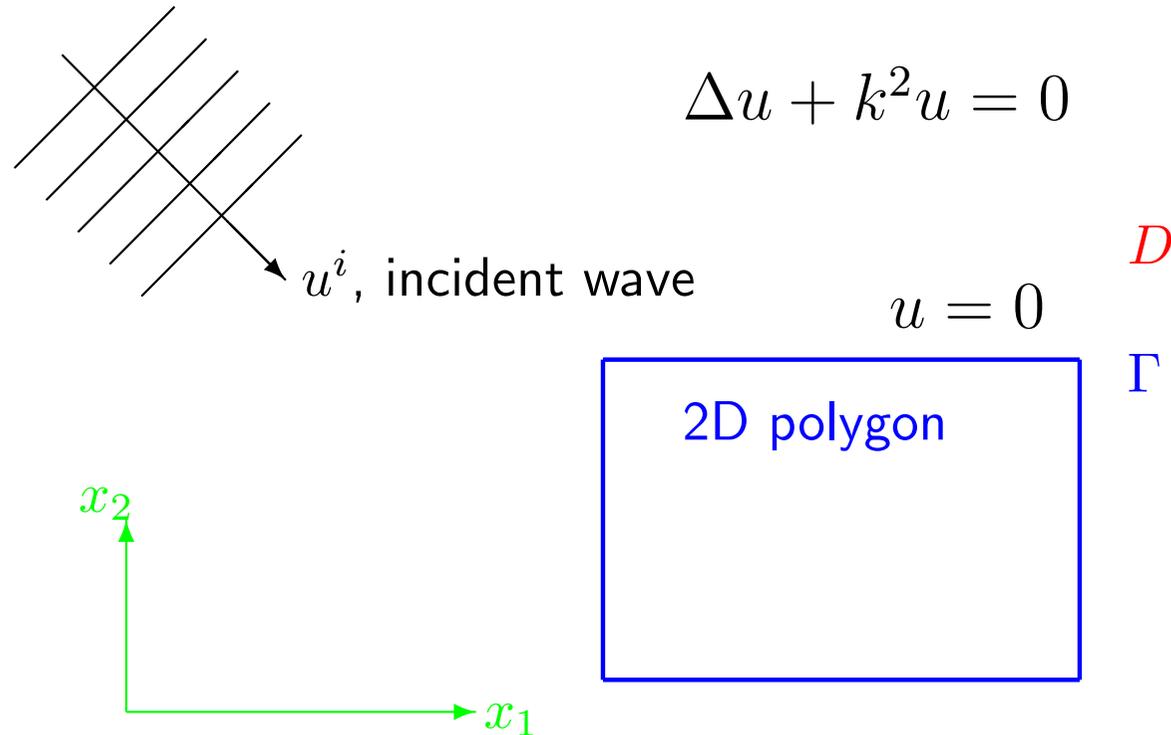


From Green's representation theorem (Burton & Miller 1971):

$$\frac{1}{2} \frac{\partial u}{\partial n}(x) + \int_{\Gamma} \left( \frac{\partial \Phi(x, y)}{\partial n(x)} + i\eta \Phi(x, y) \right) \frac{\partial u}{\partial n}(y) ds(y) = f(x), \quad x \in \Gamma,$$

where

$$f(x) := \frac{\partial u^i}{\partial n}(x) + i\eta u^i(x).$$



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**Theorem** (follows from Burton & Miller 1971, Selepov 1969) If  $\eta \in \mathbb{R}$ ,  $\eta \neq 0$ , then this integral equation is uniquely solvable in  $L^2(\Gamma)$ .

$$\frac{1}{2} \frac{\partial u}{\partial n}(x) + \int_{\Gamma} \left( \frac{\partial \Phi(x, y)}{\partial n(x)} + i\eta \Phi(x, y) \right) \frac{\partial u}{\partial n}(y) ds(y) = f(x), \quad x \in \Gamma.$$

**Conventional BEM:** Apply a Galerkin method, approximating  $\partial u/\partial n$  by a piecewise polynomial of degree  $P$ , leading to a linear system to solve with  $N$  degrees of freedom. **Problem:**  $N$  of order of  $kL$ , where  $L$  is linear dimension, so cost is  $O(N^2)$  to compute full matrix and apply iterative solver ... or close to  $O(N)$  if a fast multipole method (e.g. Amini & Profit 2003, Darve 2004) is used.

This is **fantastic** but still infeasible as  $kL \rightarrow \infty$ .

$$\frac{1}{2} \frac{\partial u}{\partial n}(x) + \int_{\Gamma} \left( \frac{\partial \Phi(x, y)}{\partial n(x)} + i\eta \Phi(x, y) \right) \frac{\partial u}{\partial n}(y) ds(y) = f(x), \quad x \in \Gamma.$$

**Alternative:** Reduce  $N$  by using new basis functions, e.g.

(i) approximate  $\partial u / \partial n$  by taking a large number of plane waves and multiplying these by conventional piecewise polynomial basis functions (Perrey-Debain et al. 2003, 2004). **This is very successful (in 2D, 3D, for acoustic/elastic waves and Neumann/impedance b.c.s), reducing number of degrees of freedom per wavelength from e.g. 6-10 to close to 2.** However  $N$  still increases proportional to  $kL$ .

$$\frac{1}{2} \frac{\partial u}{\partial n}(x) + \int_{\Gamma} \left( \frac{\partial \Phi(x, y)}{\partial n(x)} + i\eta \Phi(x, y) \right) \frac{\partial u}{\partial n}(y) ds(y) = f(x), \quad x \in \Gamma.$$

**Alternative:** Reduce  $N$  by using new basis functions, e.g.

(ii) for convex scatterers, remove some of the oscillation by factoring out the oscillation of the incident wave, i.e. writing

$$\frac{\partial u}{\partial n}(y) = \frac{\partial u^i}{\partial n}(y) \times F(y)$$

and approximating  $F$  by a conventional BEM (e.g. Abboud et al. 1994, Darrigrand 2002, Bruno et al 2004).

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$$\frac{\partial u}{\partial n}(y) = \frac{\partial u^i}{\partial n}(y) \times F(y) \quad (*)$$

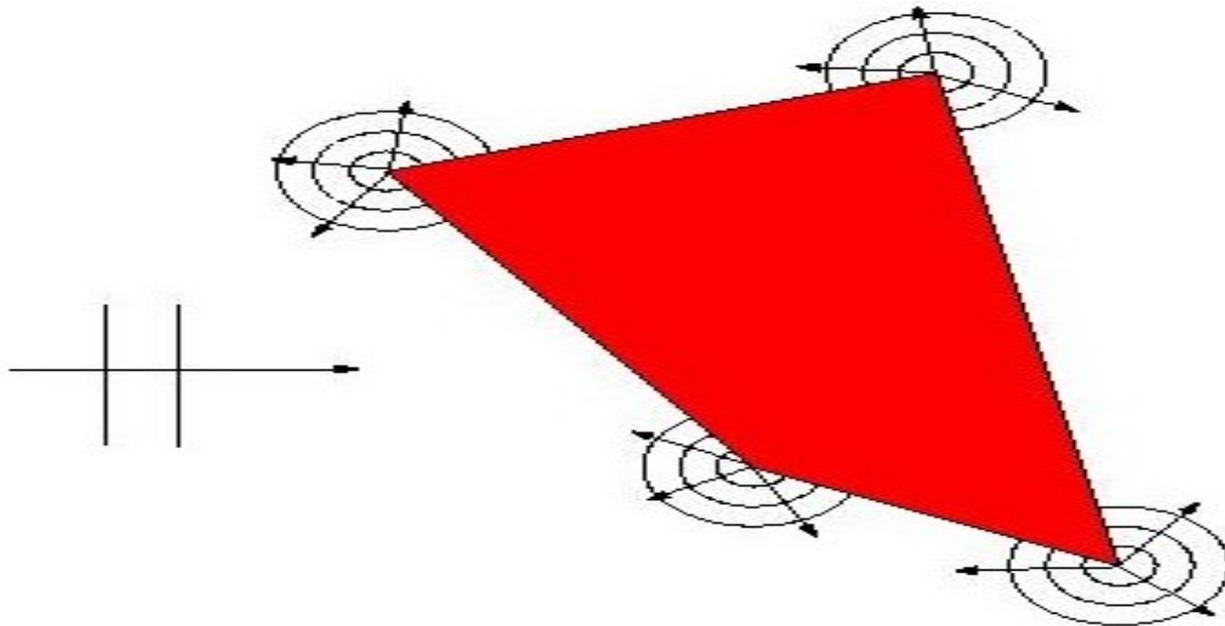
and approximating  $F$  by a conventional BEM.

For **smooth obstacles** this works well: equation  $(*)$  holds with  $F(y) \approx 2$  on the illuminated side (physical optics) and  $F(y) \approx 0$  in the shadow zone.

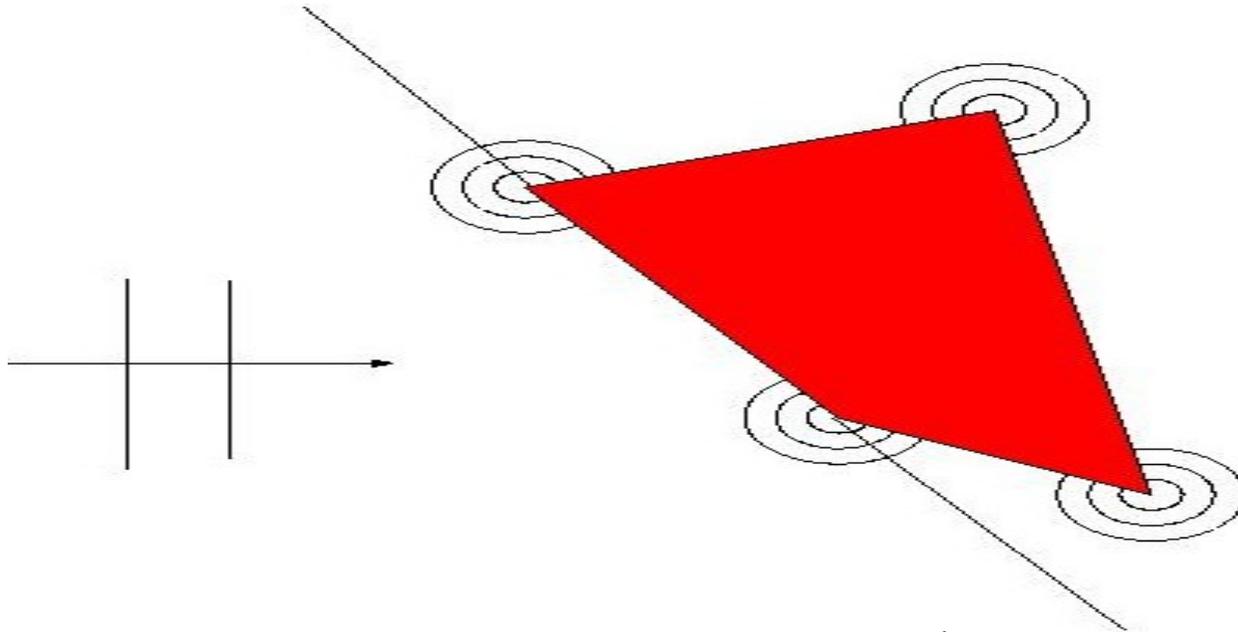
(ii) for convex scatterers, remove some of the oscillation by factoring out the oscillation of the incident wave, i.e. writing

$$\frac{\partial u}{\partial n}(y) = \frac{\partial u^i}{\partial n}(y) \times F(y) \quad (*)$$

and approximating  $F$  by a conventional BEM. **Not very effective for non-smooth scatterers.**



## Understanding solution behaviour



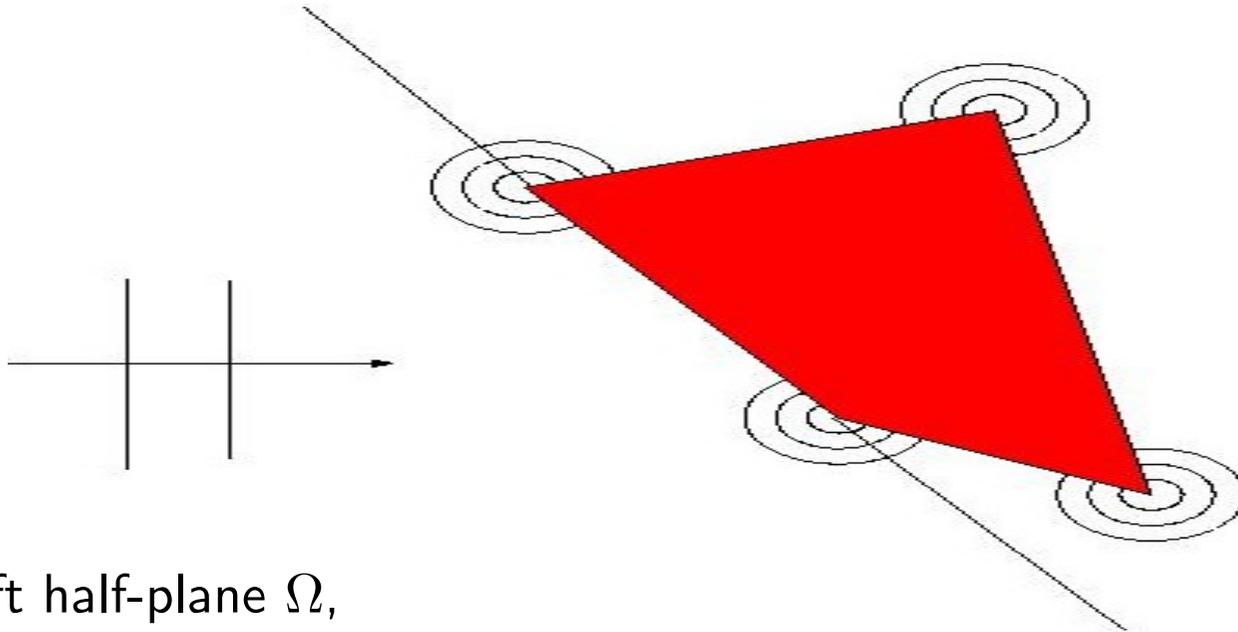
Let

$$G(x, y) := \Phi(x, y) - \Phi(x, y')$$

be the Dirichlet Green function for the left half-plane  $\Omega$ . By Green's representation theorem,

$$u(x) = u^i(x) + u^r(x) + \int_{\partial\Omega \setminus \Gamma} \frac{\partial G(x, y)}{\partial n(y)} u(y) ds(y), \quad x \in \Omega.$$

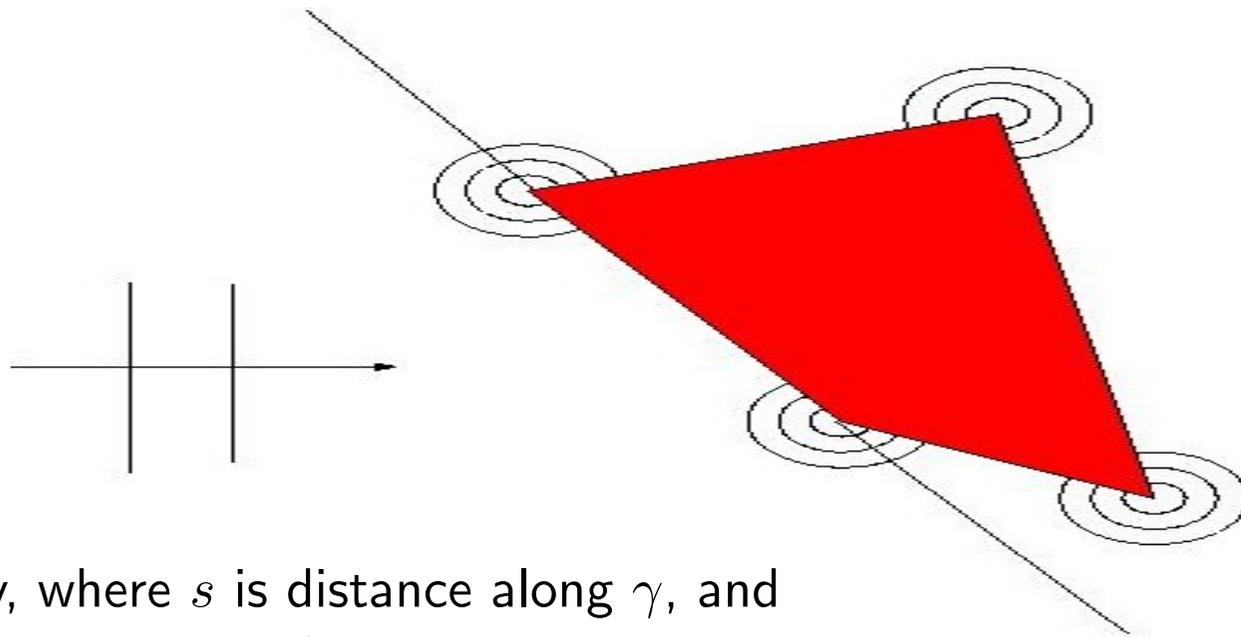
## Understanding solution behaviour



In the left half-plane  $\Omega$ ,

$$u(x) = u^i(x) + u^r(x) + \int_{\partial\Omega \setminus \Gamma} \frac{\partial G(x, y)}{\partial n(y)} u(y) ds(y)$$

$$\Rightarrow \frac{\partial u}{\partial n}(x) = 2 \frac{\partial u^i}{\partial n}(x) + 2 \int_{\partial\Omega \setminus \Gamma} \frac{\partial^2 \Phi(x, y)}{\partial n(x) \partial n(y)} u(y) ds(y), \quad x \in \gamma = \partial\Omega \cap \Gamma.$$



Explicitly, where  $s$  is distance along  $\gamma$ , and  $\phi(s)$  and  $\psi(s)$  are  $k^{-1}\partial u/\partial n$  and  $u$ , at distance  $s$  along  $\gamma$ ,

$$\phi(s) = P.O. + \frac{i}{2} [e^{iks}v_+(s) + e^{-iks}v_-(s)]$$

where

$$v_+(s) := k \int_{-\infty}^0 F(k(s-s_0))e^{-iks_0}\psi(s_0)ds_0.$$

and  $F(z) := e^{-iz}H_1^{(1)}(z)/z$

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Now  $F(z) := e^{-iz} H_1^{(1)}(z)/z$  which is non-oscillatory, in that

$$F^{(n)}(z) = O(z^{-3/2-n}) \text{ as } z \rightarrow \infty.$$

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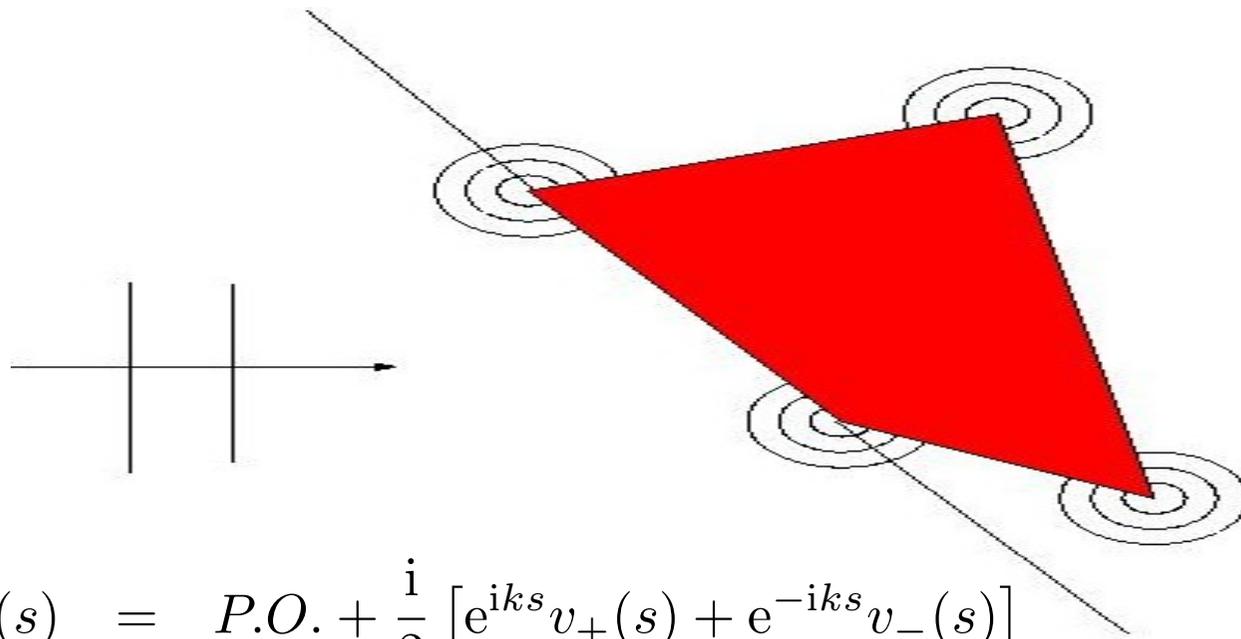
where

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Now  $F(z) := e^{-iz} H_1^{(1)}(z)/z$  which is non-oscillatory, in that

$$F^{(n)}(z) = O(z^{-3/2-n}) \text{ as } z \rightarrow \infty.$$

$$\Rightarrow v_+^{(n)}(s) = O(k^n (ks)^{-1/2-n}) \text{ as } ks \rightarrow \infty.$$



$$\phi(s) = P.O. + \frac{i}{2} [e^{iks} v_+(s) + e^{-iks} v_-(s)]$$

where  $k^{-n} |v_+^{(n)}(s)| = O((ks)^{-1/2-n})$  as  $ks \rightarrow \infty$

and (by separation of variables local to the corner),

$$k^{-n} |v_+^{(n)}(s)| = O((ks)^{-\alpha-n}) \text{ as } ks \rightarrow 0,$$

where  $\alpha < 1/2$  depends on the corner angle.

$$\phi(s) = P.O. + \frac{i}{2} [e^{iks} v_+(s) + e^{-iks} v_-(s)]$$

where

$$k^{-n} |v_+^{(n)}(s)| = \begin{cases} O((ks)^{-1/2-n}) & \text{as } ks \rightarrow \infty \\ O((ks)^{-\alpha-n}) & \text{as } ks \rightarrow 0, \end{cases}$$

where  $\alpha < 1/2$  depends on the corner angle.

Thus approximate

$$\phi(s) \approx P.O. + \frac{i}{2} [e^{iks} V_+(s) + e^{-iks} V_-(s)],$$

where  $V_+$  and  $V_-$  are piecewise polynomials on graded meshes.

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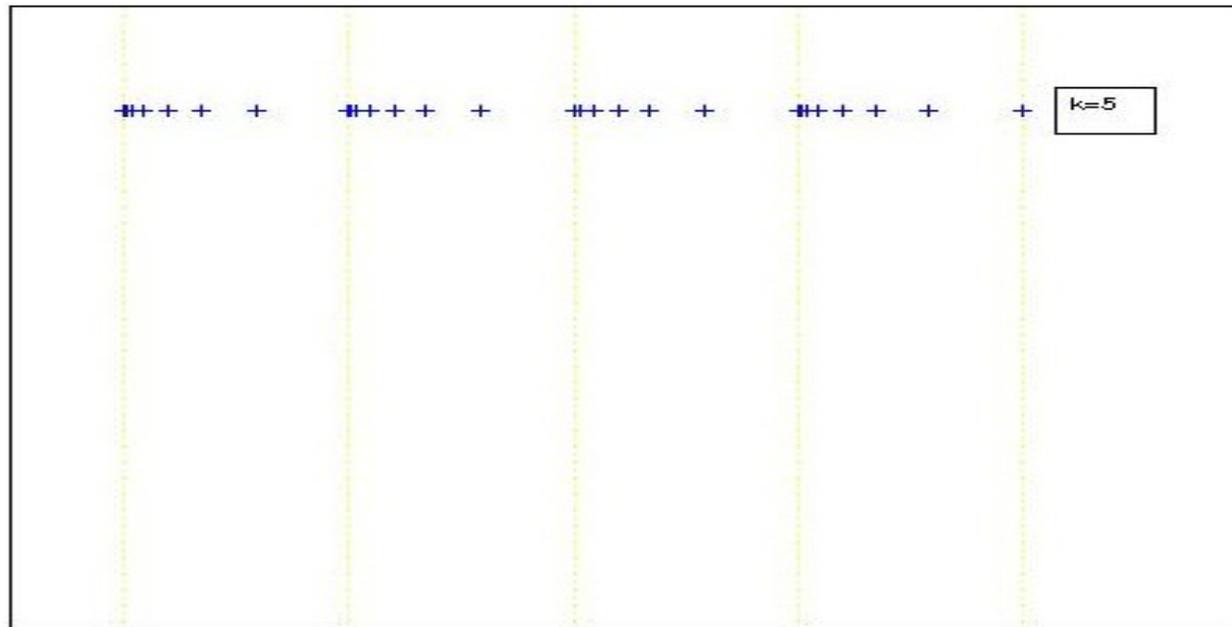


Figure 1: Scattering by a square

Thus approximate

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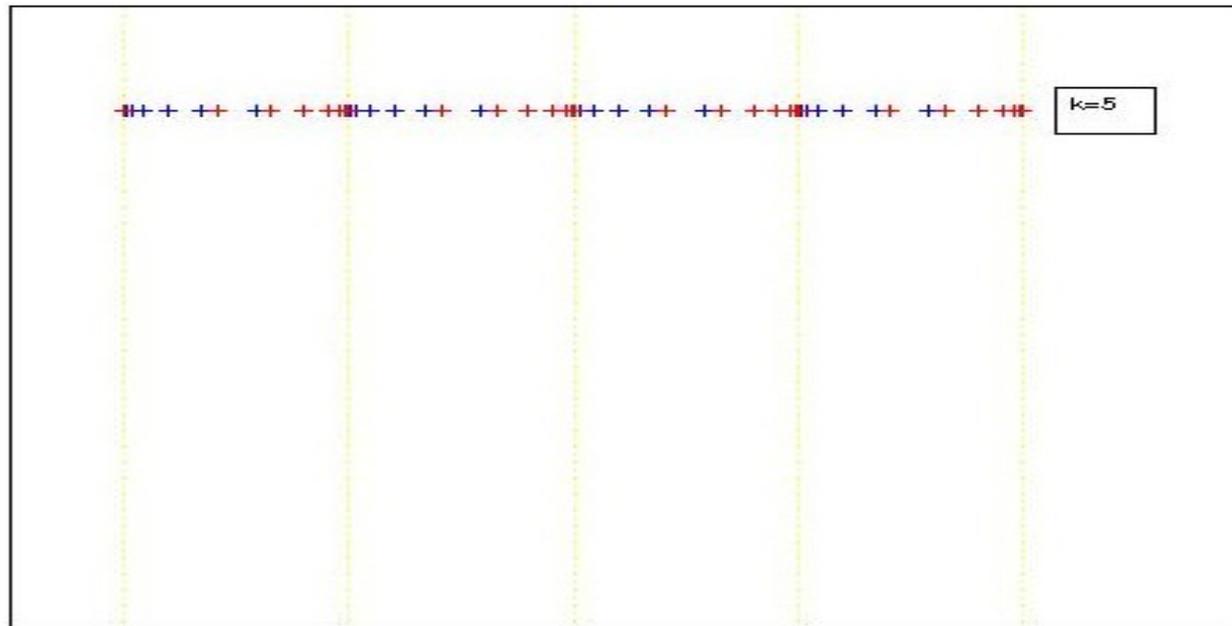


Figure 2: Scattering by a square

Thus approximate

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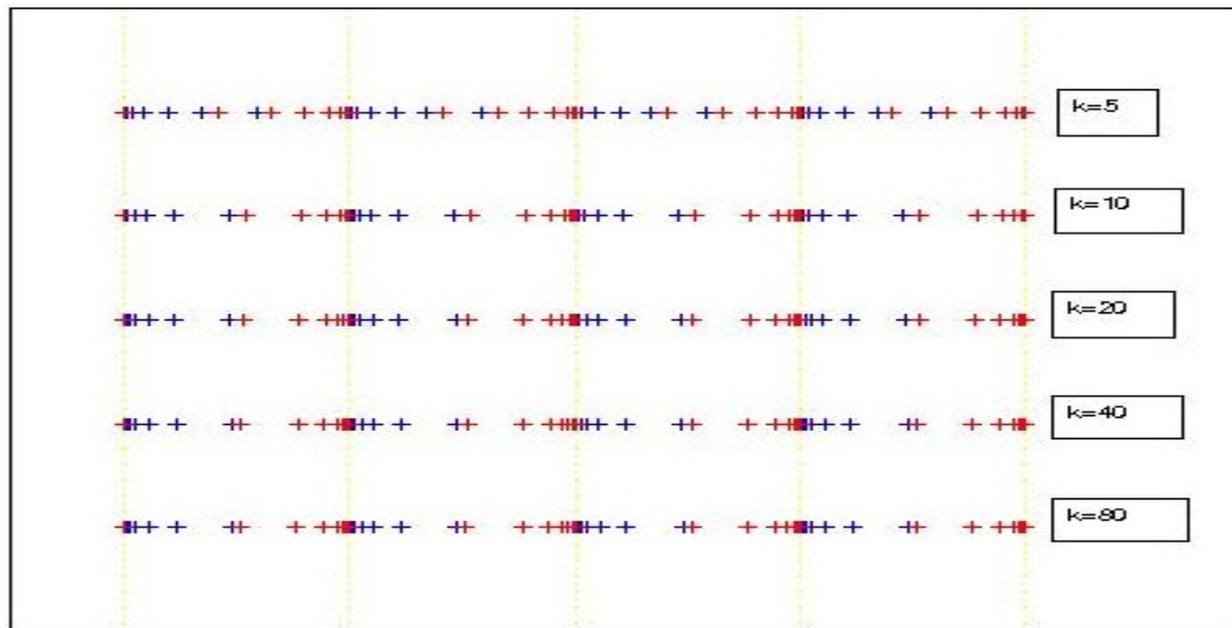


Figure 3: Scattering by a square

## Approximation error

**Theorem:** If  $V_+$  is the best  $L_2$  approximation from the approximation space, then

$$k^{1/2} \|v_+ - V_+\|_2 \leq C_p \frac{n^{1/2} (1 + \log^{1/2}(kL))}{N^{p+1}},$$

where

- $N \propto$  degrees of freedom
- $p =$  polynomial degree
- $L =$  max side length
- $n =$  number of sides of polygon

## Boundary integral equation method

Integral equation in parametric form

$$\varphi(s) + \mathcal{K}\varphi(s) = F(s),$$

where

$$\varphi(s) := \frac{1}{k} \frac{\partial u}{\partial n}(x(s)) - P.O..$$

**Theorem.** The operator  $(I + \mathcal{K}) : L_2(\Gamma) \mapsto L_2(\Gamma)$  is bijective with bounded inverse

$$\|(I + \mathcal{K})^{-1}\|_2 \leq C,$$

so that the integral equation has exactly one solution.

## Boundary integral equation method

Integral equation in parametric form

$$\varphi(s) + \mathcal{K}\varphi(s) = F(s),$$

where

$$\varphi(s) := \frac{1}{k} \frac{\partial u}{\partial n}(x(s)) - P.O..$$

**Difficulty 1** The operator  $(I + \mathcal{K}) : L_2(\Gamma) \mapsto L_2(\Gamma)$  is bijective with bounded inverse

$$\|(I + \mathcal{K})^{-1}\|_2 \leq C(k),$$

where the dependence of  $C(k)$  on  $k$  is not clear.

**Approximation space:** seek

$$\varphi_N(s) = \sum_{j=1}^M v_j \rho_j(s) \in V_N,$$

where

$\rho_j(s) := e^{\pm iks} \times$  piecewise polynomial supported on graded mesh.

**Question:** how do we compute  $v_j$ ?

## Galerkin method

To solve

$$\varphi(s) + \mathcal{K}\varphi(s) = F(s),$$

seek  $\varphi_{N_G} \in V_N$  such that

$$(I + P_{N_G}\mathcal{K})\varphi_{N_G} = P_{N_G}F,$$

where  $P_{N_G}$  is the orthogonal projection onto the approximation space.

Equivalently

$$(\varphi_{N_G}, \rho) + (\mathcal{K}\varphi_{N_G}, \rho) = (F, \rho), \quad \forall \rho \in V_N,$$

$$\Rightarrow \sum_{j=1}^M v_j [(\rho_j, \rho_m) + (\mathcal{K}\rho_j, \rho_m)] = (F, \rho_m).$$

If  $\rho_j, \rho_m$  supported on same side of polygon, integrals not oscillatory.

## Galerkin method

**Theorem.** For  $N \geq N^*$ , the operator  $(I + P_{N_G} \mathcal{K}) : L_2(\Gamma) \mapsto V_N$  is bijective with bounded inverse

$$\|(I + P_{N_G} \mathcal{K})^{-1}\|_2 \leq C_s.$$

## Galerkin method

**Difficulty 2.** For  $N \geq N^*(k)$ , the operator  $(I + P_{N_G} \mathcal{K}) : L_2(\Gamma) \mapsto V_N$  is bijective with bounded inverse

$$\|(I + P_{N_G} \mathcal{K})^{-1}\|_2 \leq C_s(k),$$

where the dependence of  $N^*(k)$  and  $C_s(k)$  on  $k$  is not clear.

## Collocation method

To solve

$$\varphi(s) + \mathcal{K}\varphi(s) = F(s),$$

seek  $\varphi_{N_C} \in V_N$  such that

$$(I + P_{N_C}\mathcal{K})\varphi_{N_C} = P_{N_C}F,$$

where  $P_{N_C}$  is the interpolatory projection onto the approximation space.

Equivalently

$$\varphi_{N_C}(s_m) + \mathcal{K}\varphi_{N_C}(s_m) = F(s_m), \quad m = 1, \dots, M,$$

$$\Rightarrow \sum_{j=1}^M v_j [\rho_j(s_m) + \mathcal{K}\rho_j(s_m)] = F(s_m).$$

If  $\rho_j$  supported on same side of polygon as  $s_m$ , integrals not oscillatory.

## Collocation method

We have not shown that  $(I + P_{N_C} \mathcal{K}) : L_2(\Gamma) \mapsto V_N$  is bijective with bounded inverse.

## Galerkin vs. Collocation: error analysis

**Theorem** There exists a constant  $C_p > 0$ , independent of  $k$ , such that for  $N \geq N^*$

$$k^{1/2} \|\varphi - \varphi_{N_G}\|_2 \leq C_p C_s \sup_{x \in D} |u(x)| \frac{n^{1/2} (1 + \log^{1/2}(kL/n))}{N^{p+1}},$$

$$k^{1/2} |u(x) - u_{N_G}(x)| \leq C_p C_s \sup_{x \in D} |u(x)| \frac{n^{1/2} (1 + \log^{1/2}(kL/n))}{N^{p+1}}.$$

- Stability and convergence not proven for collocation scheme.

## Galerkin vs. Collocation: conditioning

**Galerkin:** mass matrix  $M_G := [(\rho_j, \rho_m)]$  has  $\text{cond}M \leq (1 + \sigma)/(1 - \sigma)$ ,  
where

$$\sigma \leq \max \left\{ \frac{\min(y_j^+, y_m^-) - \max(y_{j-1}^+, y_{m-1}^-)}{\sqrt{(y_j^+ - y_{j-1}^+)(y_m^- - y_{m-1}^-)}} \right\} < 1,$$

and if side lengths and angles are equal we can prove

$$\sigma < \left( \frac{1}{kL} \right)^{1/2N \log k}.$$

**Collocation:** difficulty with choice of collocation points,  
 $M_C := [\rho_j(s_m)]$  may be ill conditioned.

## Galerkin vs. Collocation: implementation

**Galerkin:** need to evaluate numerically many integrals of form

$$\int_{-b}^{-a} \int_c^d \left[ H_0^{(1)}(k\sqrt{s^2 + t^2}) + \frac{itH_1^{(1)}(k\sqrt{s^2 + t^2})}{\sqrt{s^2 + t^2}} \right] e^{ik(\sigma_j t - \sigma_m s)} dt ds.$$

**Collocation:** need to evaluate numerically many integrals of form

$$\int_a^b \left[ H_0^{(1)}(k\sqrt{s_m^2 + t^2}) + \frac{itH_1^{(1)}(k\sqrt{s_m^2 + t^2})}{\sqrt{s_m^2 + t^2}} \right] e^{ik\sigma_j t} dt.$$

- Collocation method easier to implement

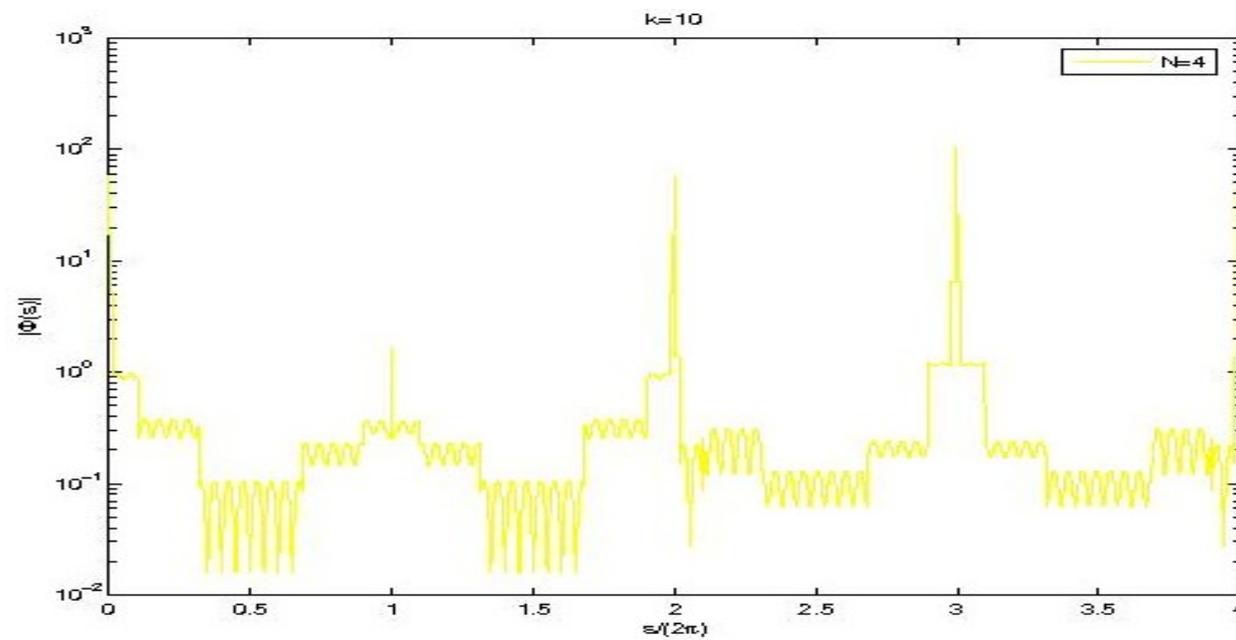
## Numerical results

scattering by a square,  $k = 5$

scattering by a square,  $k = 10$

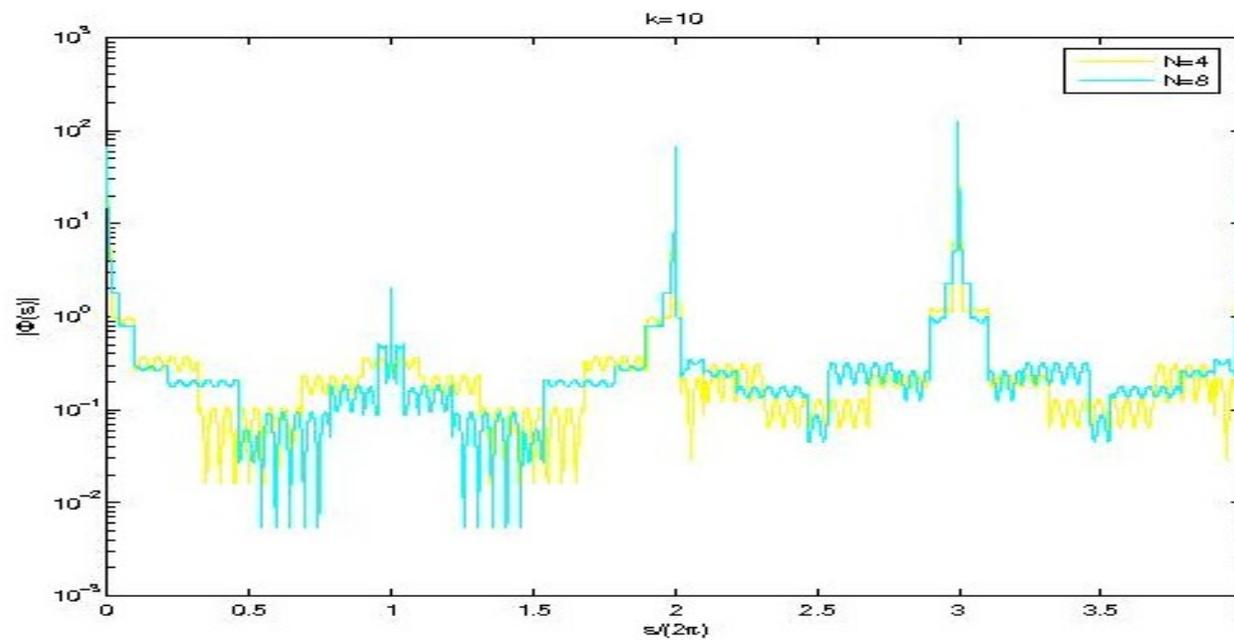
## Numerical results (scattering by a square)

Solution minus P.O. approximation;



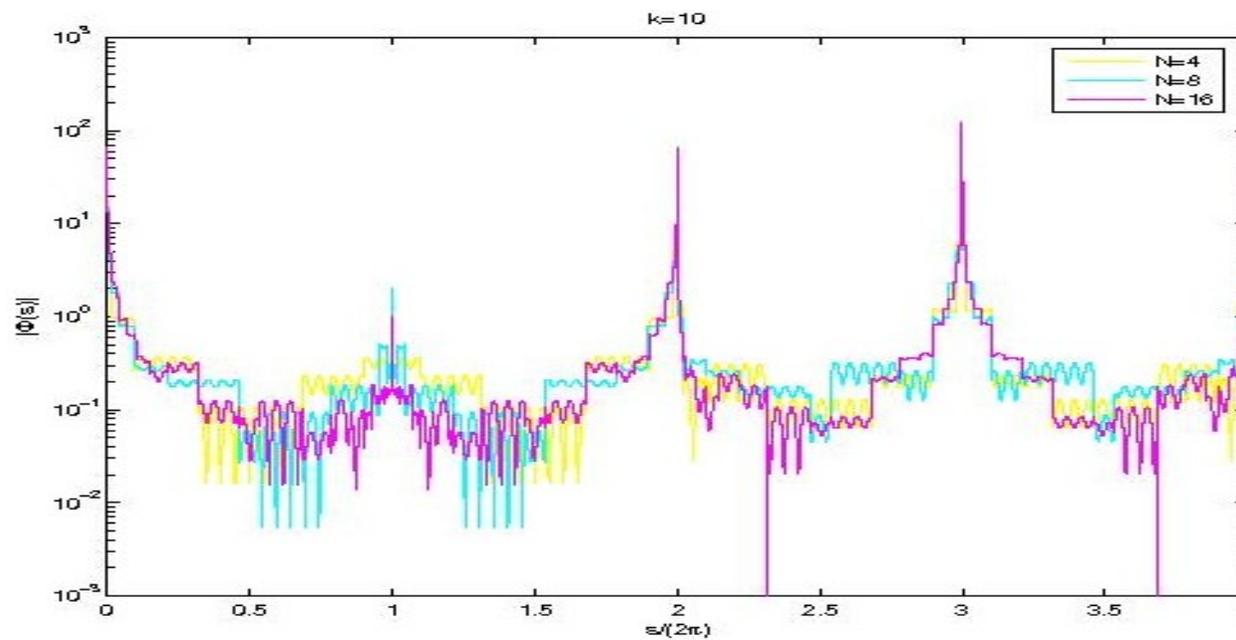
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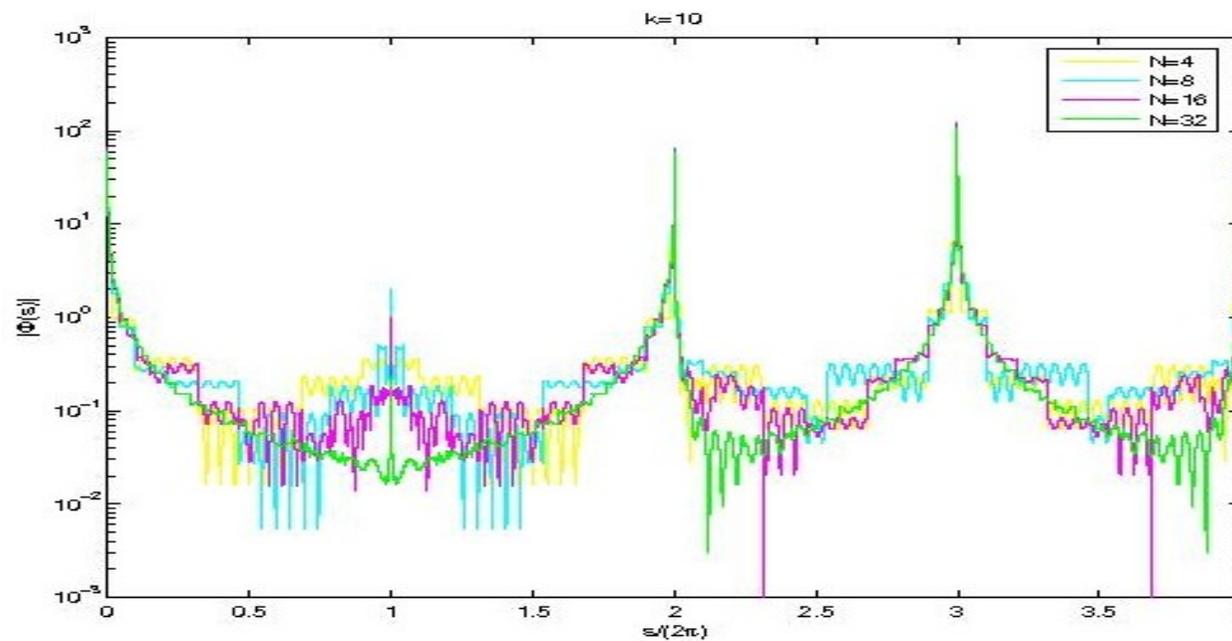
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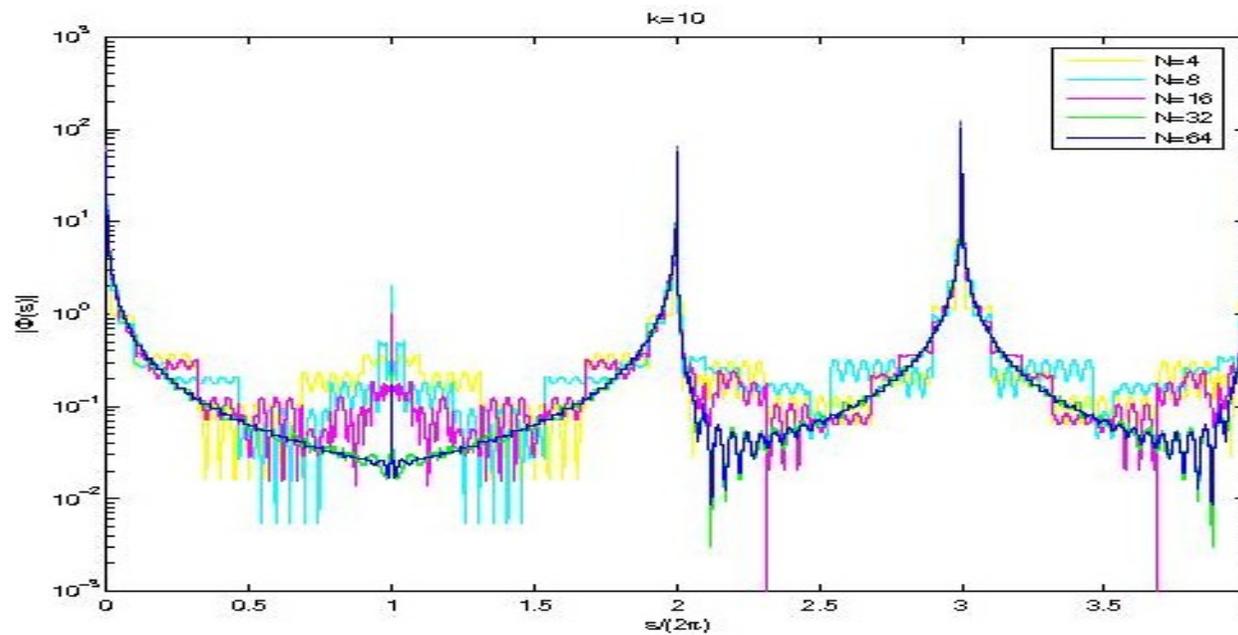
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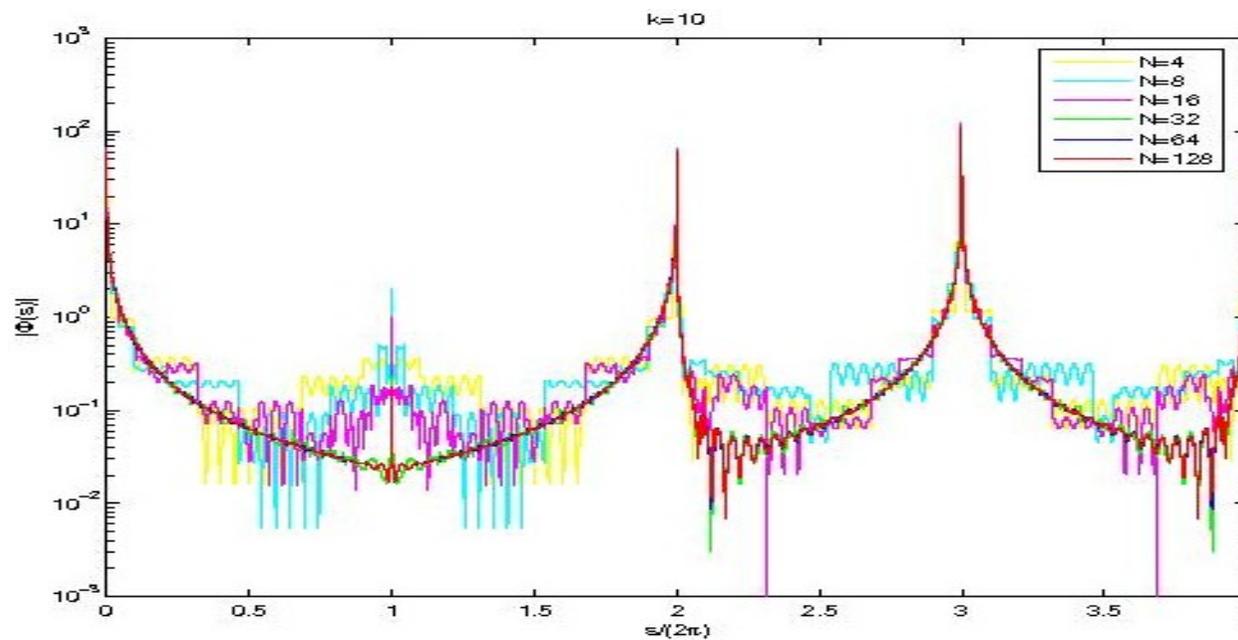
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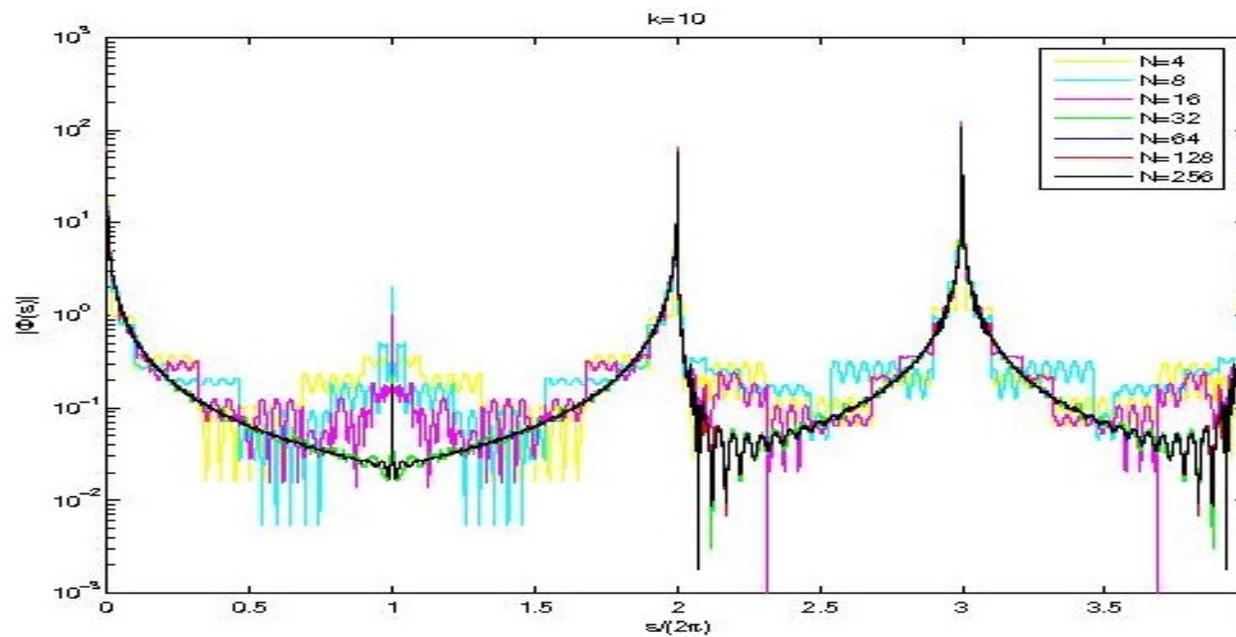
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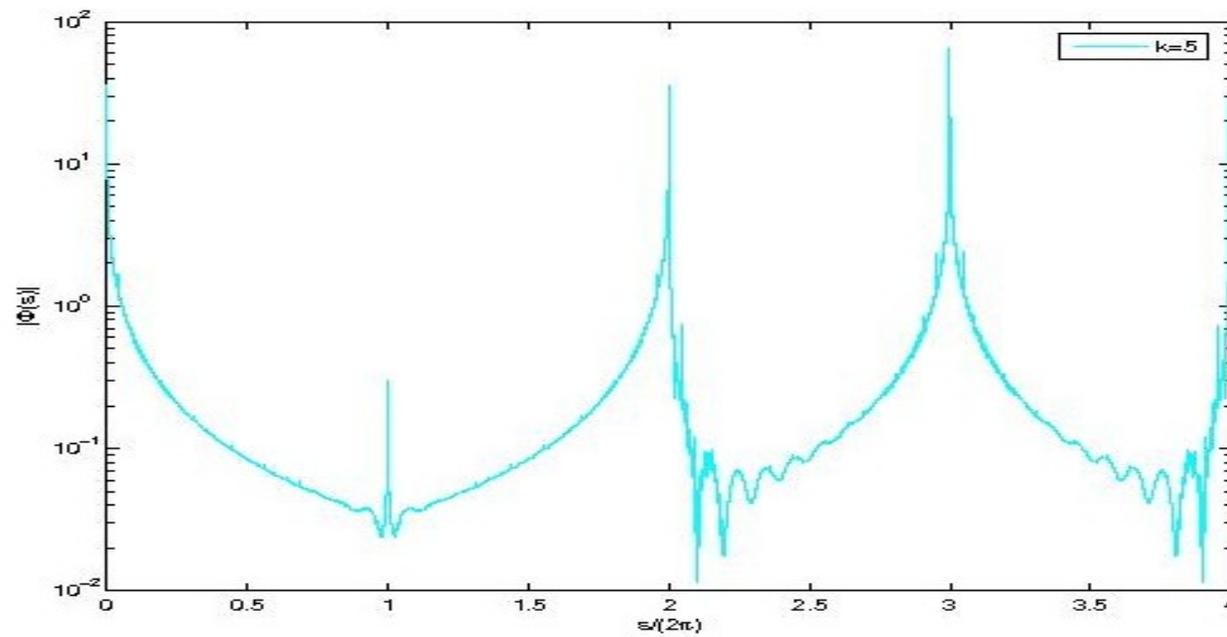
## Numerical results (scattering by a square)

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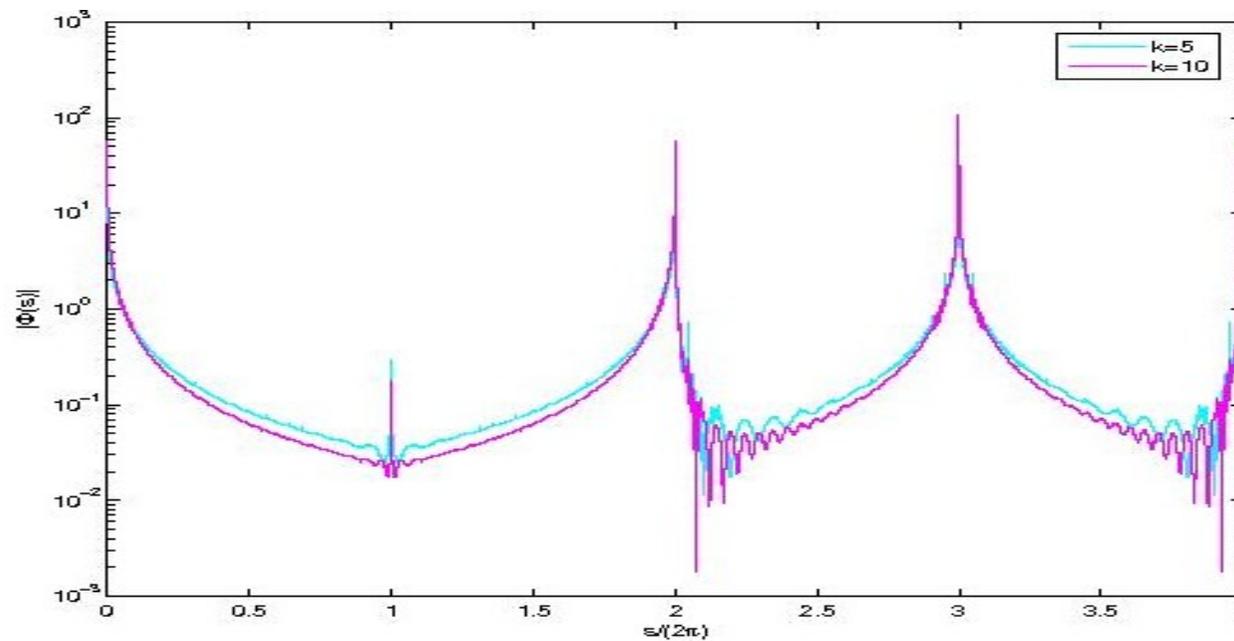
## Numerical results (scattering by a square)

"Exact" solution minus P.O. approximation,  $k = 5$ ;



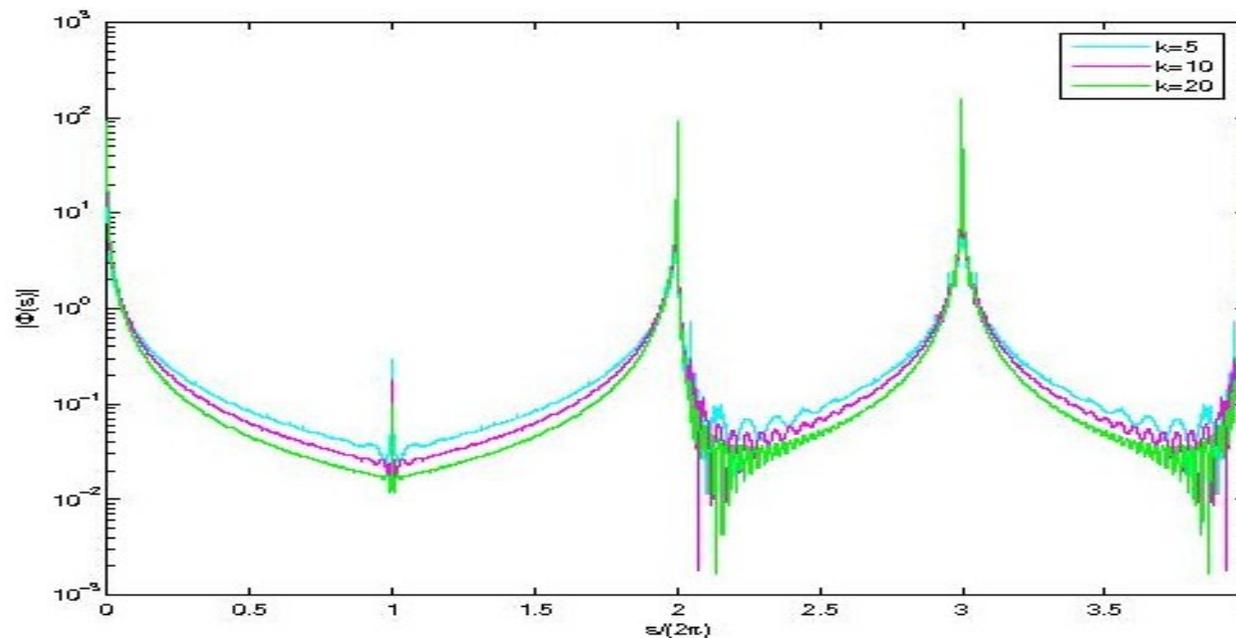
## Numerical results (scattering by a square)

"Exact" solution minus P.O. approximation,  $k = 10$ ;



## Numerical results (scattering by a square)

"Exact" solution minus P.O. approximation,  $k = 20$ ;



## Numerical results (scattering by a square)

"Exact" solution minus P.O. approximation,  $k = 40$ ;

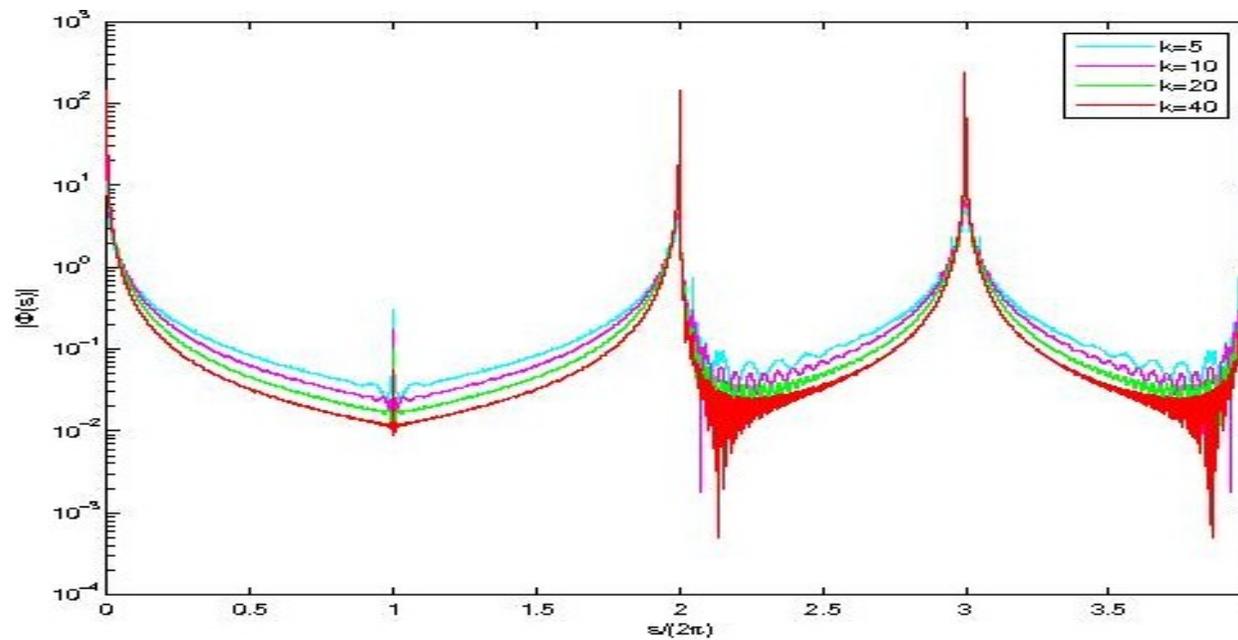


Table 1: Relative errors,  $k = 10$

$k$	$N$	dof	$\frac{\ \varphi - \varphi_{N_G}\ _2}{\ \varphi\ _2}$	$\frac{\ \varphi - \varphi_{N_C}\ _2}{\ \varphi\ _2}$
10	2	24	$1.1691 \times 10^{+0}$	$7.5453 \times 10^{-1}$
	4	48	$4.3784 \times 10^{-1}$	$4.7335 \times 10^{-1}$
	8	96	$2.2320 \times 10^{-1}$	$2.6980 \times 10^{-1}$
	16	192	$1.2106 \times 10^{-1}$	$1.2670 \times 10^{-1}$
	32	376	$1.1633 \times 10^{-1}$	$6.8440 \times 10^{-2}$
	64	752	$2.8702 \times 10^{-2}$	$3.3034 \times 10^{-2}$

Table 2: Relative errors,  $k = 160$

$k$	$N$	dof	$\frac{\ \varphi - \varphi_{N_G}\ _2}{\ \varphi\ _2}$	$\frac{\ \varphi - \varphi_{N_C}\ _2}{\ \varphi\ _2}$
160	2	32	$7.2765 \times 10^{-1}$	$6.8901 \times 10^{-1}$
	4	56	$4.2628 \times 10^{-1}$	$4.4455 \times 10^{-1}$
	8	112	$4.9060 \times 10^{-1}$	$4.6445 \times 10^{-1}$
	16	224	$1.2847 \times 10^{-1}$	$2.3456 \times 10^{-1}$
	32	456	$8.4578 \times 10^{-2}$	$9.3327 \times 10^{-2}$
	64	904	$3.4570 \times 10^{-2}$	$4.8153 \times 10^{-2}$

$k$	$M_N$	$\ \varphi - \varphi_N\ _2$	$\ \varphi - \varphi_N\ _2 / \ \varphi\ _2$	COND
5	360	$3.6171 \times 10^{-1}$	$6.8909 \times 10^{-2}$	$2.6 \times 10^1$
10	376	$8.5073 \times 10^{-1}$	$1.1633 \times 10^{-1}$	$1.8 \times 10^2$
20	392	$8.0941 \times 10^{-1}$	$7.9909 \times 10^{-2}$	$1.0 \times 10^3$
40	416	$1.1252 \times 10^0$	$8.0909 \times 10^{-2}$	$2.4 \times 10^2$
80	432	$1.6630 \times 10^0$	$8.7071 \times 10^{-2}$	$5.9 \times 10^2$
160	456	$2.1936 \times 10^0$	$8.4578 \times 10^{-2}$	$5.2 \times 10^2$
320	472	$3.5185 \times 10^0$	$1.0211 \times 10^{-1}$	$8.1 \times 10^2$

Table 3: Relative  $L_2$  errors, various  $k$ ,  $N = 32$

$k$	$N$	$ \frac{u_N - u_{256}}{u_{256}}(-\pi, 3\pi) $	$ \frac{u_N - u_{256}}{u_{256}}(3\pi, 3\pi) $	$ \frac{u_N - u_{256}}{u_{256}}(3\pi, -\pi) $
5	4	$1.9588 \times 10^{-2}$	$1.0071 \times 10^{-3}$	$1.5885 \times 10^{-2}$
	8	$4.2631 \times 10^{-3}$	$2.8032 \times 10^{-3}$	$2.3213 \times 10^{-3}$
	16	$3.6178 \times 10^{-4}$	$3.1438 \times 10^{-4}$	$1.3514 \times 10^{-3}$
	32	$6.6463 \times 10^{-5}$	$2.9271 \times 10^{-5}$	$1.7115 \times 10^{-5}$
	64	$1.1634 \times 10^{-5}$	$5.4525 \times 10^{-6}$	$3.8267 \times 10^{-6}$

Table 4: Relative errors, for  $u_N(x)$

$k$	$N$	$ \frac{u_N - u_{256}}{u_{256}}(-\pi, 3\pi) $	$ \frac{u_N - u_{256}}{u_{256}}(3\pi, 3\pi) $	$ \frac{u_N - u_{256}}{u_{256}}(3\pi, -\pi) $
320	4	$7.2339 \times 10^{-6}$	$9.1702 \times 10^{-6}$	$6.5155 \times 10^{-5}$
	8	$1.3617 \times 10^{-5}$	$4.7357 \times 10^{-6}$	$3.6329 \times 10^{-5}$
	16	$1.0694 \times 10^{-5}$	$3.0122 \times 10^{-6}$	$2.9284 \times 10^{-5}$
	32	$1.0691 \times 10^{-6}$	$5.3066 \times 10^{-7}$	$2.8225 \times 10^{-6}$
	64	$3.1606 \times 10^{-7}$	$3.0148 \times 10^{-7}$	$8.1702 \times 10^{-7}$

Table 5: Relative errors, for  $u_N(x)$

## What we actually are computing ...

The difference between the exact solution and a leading order approximation;

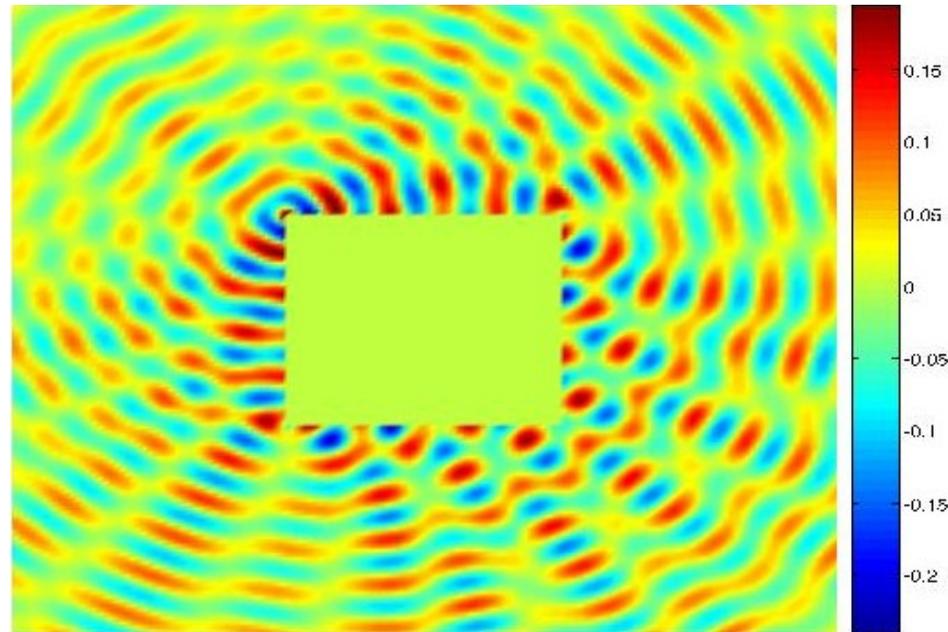


Figure 4: square,  $k = 5$

## What we actually are computing . . .

The difference between the exact solution and a leading order approximation;

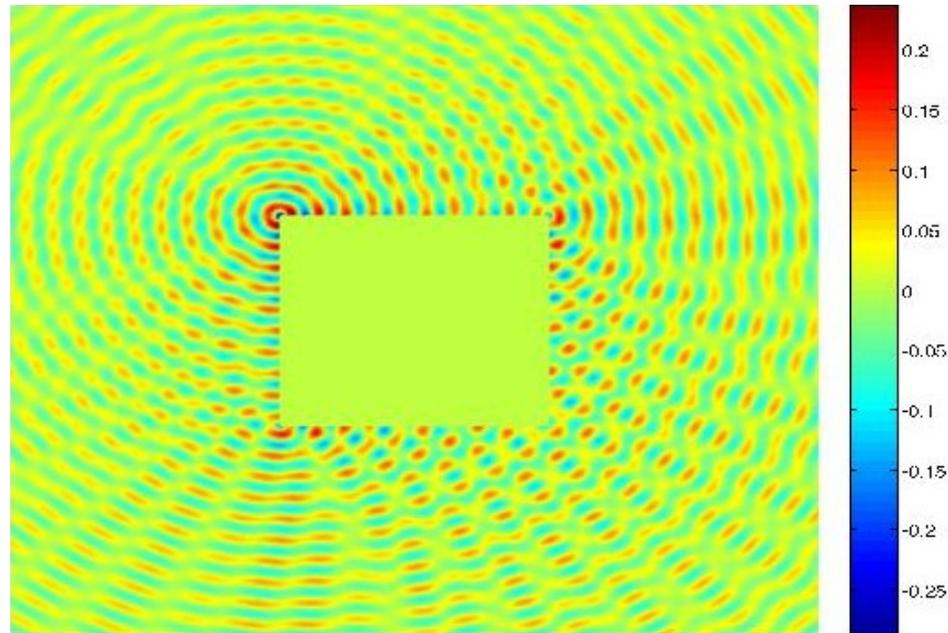


Figure 5: square,  $k = 10$

## What we actually are computing . . .

The difference between the exact solution and a leading order approximation;

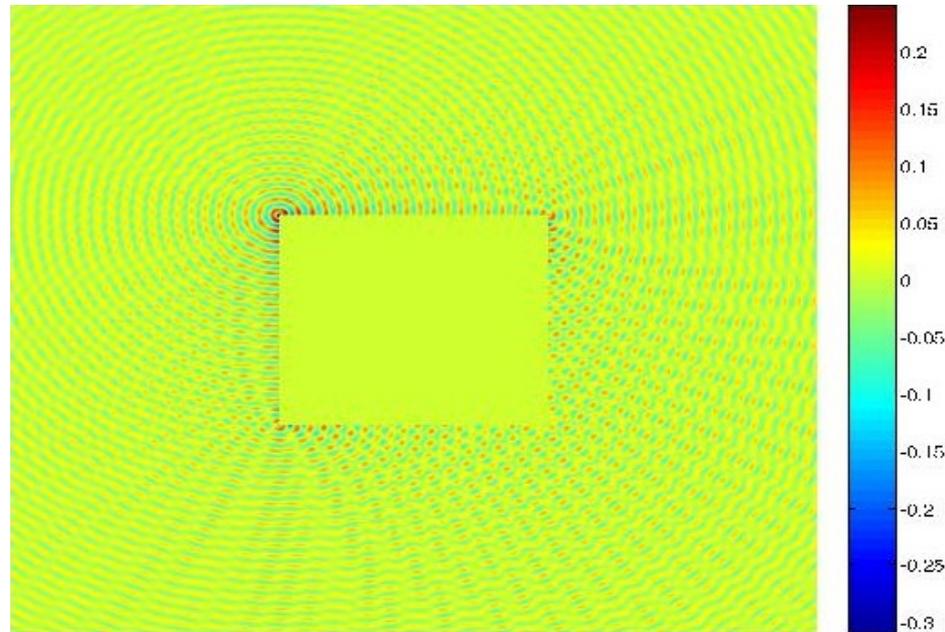


Figure 6: square,  $k = 20$

## What we actually are computing . . .

The difference between the exact solution and a leading order approximation;

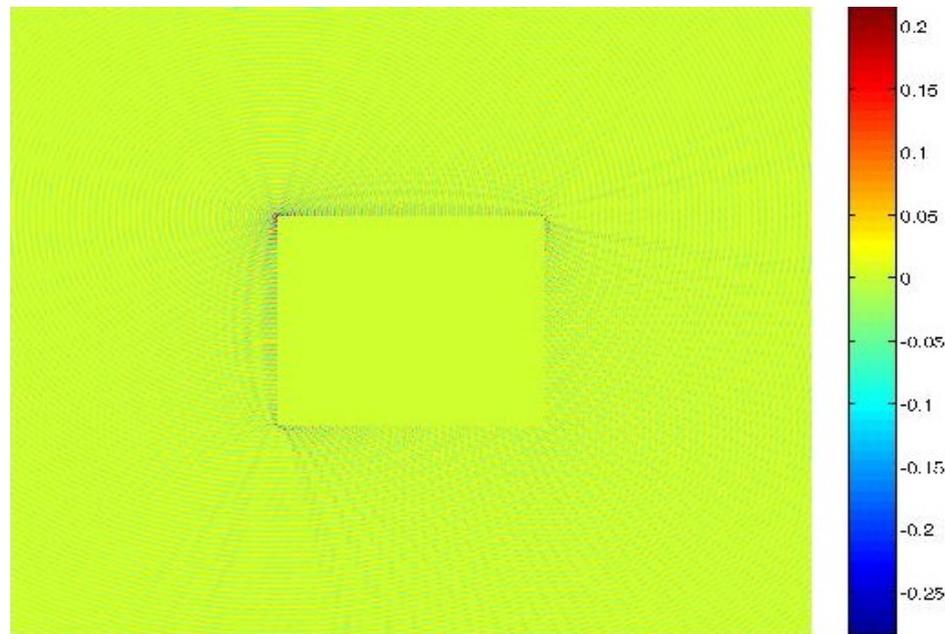


Figure 7: square,  $k = 40$

## Summary and Conclusions

- Using Green's representation theorem in a half-plane we can understand behaviour of the field on the boundary and its derivatives for scattering by a convex polygon (extends to convex polyhedron in 3D)
- For a convex polygon, design of an optimal graded mesh for piecewise polynomial approximation is then straightforward
- The number of degrees of freedom need only grow logarithmically with the wavenumber to maintain a fixed accuracy
- Ongoing considerations
  - Galerkin vs. Collocation - stability and convergence analysis
  - Better schemes for evaluating oscillatory integrals
  - *hp* ideas