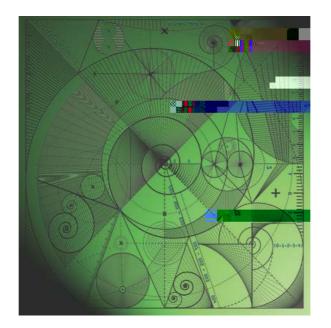


# **Department of Mathematics and Statistics**

Preprint MPS-2013-18

11 August 2014



problems. They are the simplest canonical problems that exhibit multiple di raction, yet have applications in acoustics (see, e.g., [39, 23]), electromagnetics (see, e.g., [14, 42]) and water waves (the \breakwater" problem, see e.g. [2], [26, chapter 4.7]). In this paper, we propose a numerical method (supported by a complete analysis) that we believe to be the rst method of any kind (numerical or analytical) for this problem that is provably e ective at all frequencies. Precisely, we prove that increasing the number of degrees

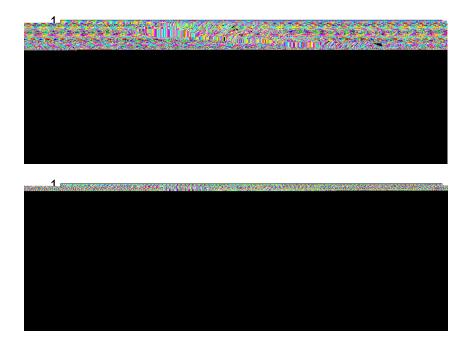


Figure 1: Total eld **u**, solving Problem **P**, for  $\mathbf{d} = (1 = \sqrt[p]{2}; 1 = \sqrt[p]{2})$  with  $\mathbf{k} = 5$  (upper) and  $\mathbf{k} = 20$  (lower).



Figure 2: Total eld  $\mathbf{u}^0$ , solving Problem  $\mathbf{P}^0$ , for  $\mathbf{d} = (1 = 2; 1 = 2)$  with  $\mathbf{k} = 5$  (upper) and  $\mathbf{k} = 20$  (lower).

An outline of the paper is as follows: we begin in  $x^2$  by reviewing details of the

speci c incident angles (see, e.g., [2, 31]), but these approaches still require a solution to those speci c problems. Thus, in general, both Problems P and P<sup>0</sup> must be solved either numerically, or else asymptotically in the high (k / 1) or low (k / 0) frequency limit.

Asymptotic and numerical approaches are usually viewed as being rather comple-

In future work, it might be of interest to use elements of our approximation space (de ned in x5) with the weighted integral operators proposed by [4, 25], to see what

Then  $C_{\text{comp}}^1$  ( ) := fUj : U  $2C_0^1$  (R<sup>n</sup>)g is a dense subset of H<sup>s</sup>( ). Second, let H<sup>s</sup>( ) denote the closure of  $C_0^1$  ( ) :=  $fU 2C_0^1$  (R<sup>n</sup>) : supp(U) g in the space H<sup>s</sup>(R<sup>n</sup>), equipped with the norm  $k k_{H_k^s()} := k k_{H_k^s(R^n)}$ . When is su ciently regular (e.g. when is C<sup>0</sup>, cf.<sub>H</sub>[27, Thm 3.29]) we have that

Theorem 3.2. Suppose that  $u^0$  is a solution of Problem P<sup>0</sup>. Then the representation formula

$$\mathbf{u}^{0}(\mathbf{x}) = \begin{pmatrix} \mathbf{u}^{i}(\mathbf{x}) + \mathbf{u}^{r}(\mathbf{x}) & S_{k} f \mathcal{D} \boldsymbol{\theta} = \boldsymbol{\mathcal{D}}_{1} g (\mathbf{x}); & \mathbf{x} \ 2 \mathbf{U}^{+}; \\ S_{k} f \mathcal{D} \boldsymbol{\theta} = \boldsymbol{\mathcal{D}}_{1} g (\mathbf{x}); & \mathbf{x} \ 2 \mathbf{U}^{+}; \end{pmatrix}$$
(14)

holds, where  $\mathcal{F}@\textcircled{l}=@h@(x) := @f(u^0) + @(u^0) 2 H^{1=2}()$ , and is an arbitrary element of  $C^1_{0;1}(\mathbb{R}^2)$ . Furthermore,  $\mathcal{F}@\textcircled{l}=@h@(x) 2 H^{1=2}()$  satisfies the integral equation (13). Conversely, suppose that  $2 H^{1=2}()$  satisfies (13). Then  $u^0$ , defined by  $u^0 := u^i + u^r - S_k$  in  $U^+$  and  $u^0 := S_k$  in U, satisfies Problem P<sup>0</sup>, and  $\mathcal{F}@\textcircled{l}=@h@ = .$ 

The following continuity and coercivity properties of the operator  $\boldsymbol{S}_k$  have been proved recently in [7, 8]:

Lemma 3.3 ([7, Theorem 5.2]). Let s 2 R. Then  $S_k : H^s() \neq H^{s+1}()$  is bounded, and for kL  $c_0 > 0$  there exists a constan $C_0 > 0$ , depending only onc<sub>0</sub> (speci cally,  $C_0 = C \log(2 + c_0^{-1})$ , where C is independent of c<sub>0</sub>), such that

$$kS_{k} k_{H_{k}^{s+1}()} = C_{0}(1 + \frac{P_{k}}{kL}) k k_{H_{k}^{s}()}; H_{k} \in exists as a 95036(s)]T_{J/F_{34}} = 0.613 IT_{J}/F_{34}$$

#### if k`

Problem P<sup>0</sup>

Proposition 4.7. The solution u of Problem P satis es the pointwise bound

$$j\mathbf{u}(\mathbf{x})\mathbf{j} = \mathbf{C} + \mathbf{p}\frac{1}{\mathbf{k}\mathbf{L}} + \mathbf{p}\frac{1}{\mathbf{k}\mathbf{d}} = \log 2 + \frac{1}{\mathbf{k}\mathbf{d}} + \log^{1-2}(2 + \mathbf{k}\mathbf{L})(1 + \mathbf{p}\frac{1}{\mathbf{k}\mathbf{L}}); \quad \mathbf{x} \ge \mathbf{D};$$
(22)

where d = dist(x; ), and C > 0 is independent of x, k and .

The second stage in the proof of Lemma 4.5 involves the derivation of a uniform bound on  $j\mathbf{u}(\mathbf{x})j$  valid on a neighbourhood of . We begin by using a separation of variables argument to bound  $j\mathbf{u}(\mathbf{x})j$  close to in terms of the L<sup>2</sup> norm of the scattered eld in a neighbourhood of .

Lemma 4.8. Let be of the form (1), and let be the corresponding solution of Problem P, with  $u^s = u - u^i$ . Let " := min  $f_{min}$  =2; = (3k

$$\begin{split} & \mathsf{K}_{\mathsf{n}} := \stackrel{\mathsf{q}}{\underset{\mathsf{k}_{\mathsf{R}}=2}{\overset{z\,dz}{\mathsf{j}_{\mathsf{J}_{\mathsf{n}=2}}(z)\mathsf{j}^2}}}{\underset{\mathsf{k}_{\mathsf{R}}=2\mathsf{k}}{\overset{z\,dz}{\mathsf{k}_{\mathsf{R}}=2\mathsf{k}}}} \text{ and } \mathsf{A}_{\mathsf{R}=2\mathsf{R}} \text{ is the annulus de ned by } \mathsf{A}_{\mathsf{R}=2\mathsf{R}} := f(\mathsf{r}; \ ) : \mathsf{R}=2 < \mathsf{r} < \mathsf{R}; \ 0 \qquad 2 \ g. \text{ To bound } j\mathsf{K}_{\mathsf{n}}j, \text{ we note that (cf., e.g., [10, (3.12)])} \end{split}$$

$$\cos z \quad \frac{J(z)(1+)}{(z=2)} \quad 1; \qquad 0 \quad z = 2; \quad > \quad 1=2:$$
(25)

where (), in (25){(27), denotes the Gamma function. Hence if 0 < kR = 3 (so that 1=2 cos z 1 for kR=2 z kR) then

$$\int K_{n} j = 2^{1 + n = 2} (1 + n = 2) \sum_{kR=2}^{s} \frac{z^{1} R}{z^{1} dz} = \frac{2^{1 + n} (1 + n = 2)}{\sqrt{n}} (kR)^{1 n = 2}$$
(26)

Thus

$$ja_{n}j = \frac{2^{4+n} (1 + n=2)(kR)^{-n=2}}{3^{p} R^{-p} \overline{n}} kuk_{L^{2}(A_{R=2;R})}; \qquad (27)$$

and, using (25) again,

$$ja_n J_{n=2}(kr)j = \frac{16(2r=R)^{n=2}}{3}kuk_{L^2(A_{R=2;R})}$$

Then, for  $\mathbf{x} \ge \mathbf{B}_{\mathsf{R}=2}(\mathbf{e}_{\mathsf{j}})$ ,

$$j\mathbf{u}(\mathbf{x})j = j\mathbf{u}(\mathbf{r}; \ )j \quad \frac{16}{3^{p}} \frac{\mathsf{X}}{\mathsf{R}} \left(2\mathbf{r} = \mathsf{R}\right)^{\mathsf{n} = 2} k \mathbf{u} k_{\mathsf{L}^{2}(\mathsf{A}_{\mathsf{R} = 2;\mathsf{R}})} = \frac{16}{3^{p}} \frac{(2\mathbf{r} = \mathsf{R})^{\mathsf{1} = 2}}{\mathsf{1} (2\mathbf{r} = \mathsf{R})^{\mathsf{1} = 2}} k \mathbf{u} k_{\mathsf{L}^{2}(\mathsf{A}_{\mathsf{R} = 2;\mathsf{R}})} = \frac{16}{3^{p}} \frac{\mathsf{R}}{\mathsf{R}} = \frac{16}{\mathsf{1} (2\mathbf{r} = \mathsf{R})^{\mathsf{1} = 2}} \frac{\mathsf{R} u k_{\mathsf{L}^{2}(\mathsf{A}_{\mathsf{R} = 2;\mathsf{R}})}}{\mathsf{R}} = \frac{16}{3^{p}} \frac{\mathsf{R}}{\mathsf{R}} = \frac{16}{\mathsf{R}} \frac{\mathsf{R} u k_{\mathsf{L}^{2}(\mathsf{R} = 2;\mathsf{R})}}{\mathsf{R}} = \frac{16}{3^{p}} \frac{\mathsf{R}}{\mathsf{R}} = \frac{16}{\mathsf{R}} \frac{\mathsf{R} u k_{\mathsf{L}^{2}(\mathsf{R} = 2;\mathsf{R})}}{\mathsf{R}} = \frac{16}{3^{p}} \frac{\mathsf{R} u k_{\mathsf{L}^{2}(\mathsf{R} = 2;\mathsf{R})}}{\mathsf{R} u k_{\mathsf{L}^{2}(\mathsf{R} = 2;\mathsf{R})}} = \frac{16}{3^{p}} \frac{\mathsf{R} u k_{\mathsf{L}^{2}(\mathsf{R} = 2;\mathsf{R})}}{\mathsf{R} u k_{\mathsf{L}^{2}(\mathsf{R} = 2;\mathsf{R})}} = \frac{16}{3^{p}} \frac{\mathsf{R} u k_{\mathsf{L}^{2}(\mathsf{R} = 2;\mathsf{R})}}{\mathsf{R} u k_{\mathsf{L}^{2}(\mathsf{R} = 2;\mathsf{R})}} = \frac{16}{3^{p}} \frac{\mathsf{R} u k_{\mathsf{L}^{2}(\mathsf{R} = 2;\mathsf{R})}}{\mathsf{R} u k_{\mathsf{L}^{2}(\mathsf{R} = 2;\mathsf{R})}} = \frac{16}{3^{p}} \frac{\mathsf{R} u k_{\mathsf{L}^{2}(\mathsf{R} = 2;\mathsf{R})}}{\mathsf{R} u k_{\mathsf{L}^{2}(\mathsf{R} = 2;\mathsf{R})}} = \frac{16}{3^{p}} \frac{\mathsf{R} u k_{\mathsf{L}^{2}(\mathsf{R} = 2;\mathsf{R})}}{\mathsf{R} u k_{\mathsf{L}^{2}(\mathsf{R} = 2;\mathsf{R})}} = \frac{16}{3^{p}} \frac{\mathsf{R} u k_{\mathsf{L}^{2}(\mathsf{R} = 2;\mathsf{R})}}{\mathsf{R} u k_{\mathsf{L}^{2}(\mathsf{R} = 2;\mathsf{R})}} = \frac{16}{3^{p}} \frac{\mathsf{R} u k_{\mathsf{L}^{2}(\mathsf{R} = 2;\mathsf{R})}}{\mathsf{R} u k_{\mathsf{L}^{2}(\mathsf{R} = 2;\mathsf{R})}} = \frac{16}{3^{p}} \frac{\mathsf{R}$$

In particular, for  $\mathbf{x} \ge \mathbf{B}_{R=4}(\mathbf{e}_j)$  we have

$$j\mathbf{u}(\mathbf{x})j = \frac{16}{3^{P-(P-1)}} k\mathbf{u}k_{L^{2}(A_{R=2;R})}$$

Recalling that  $\mathbf{u} = \mathbf{u}^{i} + \mathbf{u}^{s}$ , and noting that  $k\mathbf{u}^{i}k_{L^{2}(\mathbf{A}_{R=2;R})}$   $\mathcal{P}_{\overline{3}R=2}$ , this implies that

$$j\mathbf{u}(\mathbf{x})j = \frac{\beta}{\overline{3}(2 - 1)} + \frac{16}{3^{p} - (2 - 1)R} k\mathbf{u}^{s} k_{L^{2}(A_{R=2;R})}; \qquad \mathbf{x} \ge B_{R=4}(\mathbf{e}_{j}):$$
(28)

To satisfy both  $\mathbf{R} = (3\mathbf{k})$  and  $\mathbf{R} < \sum_{\min}$ , it success to set, e.g.,  $\mathbf{R} = \mathbf{R}_j := \min \hat{\mathcal{T}}_{\min} = 2; = (3\mathbf{k})g$ . A similar estimate to (28) can be obtained in a neighbourhood of the right endpoint  $\mathbf{e}_i^0$ .

Now let  $\mathbf{x}_j$  denote an interior point of  $\mathbf{j}$  and let (**r**; ) be polar coordinates centered at  $\mathbf{x}_j$ , so that  $\mathbf{j}$  is a subset of the lines = 0 and =. By a similar analysis to that presented above, but with the separation of variables carried out only in a halfdisk 0 or 2 and  $\mathbf{n}=2$  replaced by  $\mathbf{n}$  etc., we can show that, if  $0 < \mathbf{R} = (3\mathbf{k})$  and  $\mathbf{R} < \min f j \mathbf{x} + \mathbf{e}_j \mathbf{i} \mathbf{j} \mathbf{j} \mathbf{x}$  then

$$j\mathbf{u}(\mathbf{x})j = \frac{4^{D_{\overline{2}}}}{\sqrt{3}} + \frac{16}{3^{P_{\overline{R}}}} k\mathbf{u}^{s} k_{L^{2}(A_{R=2;R})}; \qquad \mathbf{x} \ge B_{R=4}(\mathbf{x}_{j});$$
(29)

where  $A_{R=2R} := f(\mathbf{r}; \ ) : \mathbf{R}=2 < \mathbf{r} < \mathbf{R}; \ 0$  g is a semi-annulus centered at  $\mathbf{x}_j$ .

To combine these results we note that if  $\min fj\mathbf{x} = \mathbf{e}_j j$ ;  $j\mathbf{x} = \mathbf{e}_j^0 j g > \mathbf{R}_j = 4$  then we can take  $\mathbf{R} = \mathbf{R}_j = 4$  in (29). Then the union of the balls  $\mathbf{B}_{\mathbf{R}_j = 1} (\mathbf{x}_j)$  over all such  $\mathbf{x}_j$ , together with the balls  $\mathbf{B}_{\mathbf{R}_j = 4} (\mathbf{e}_j)$  and  $\mathbf{B}_{\mathbf{R}_j = 4} (\mathbf{e}_j^0)$ , certainly covers a ( $\mathbf{R}_j = 32$ )-neighbourhood of j. Hence we can conclude that

$$j\mathbf{u}(\mathbf{x})j = \frac{8}{\overline{3}(2 - 1)} + \frac{16}{3^{p} - (2 - 1)R_{j}} k\mathbf{u}^{s} k_{L^{2}((-j)R_{j})}; \qquad \mathbf{x} \ 2(-j)_{R_{j}} = 3\dot{2}$$
(30)

from which the result follows, since  $1=R_j = 2k(1 + (k_{min})^{-1})$ .

To use Lemma 4.8 we require an estimate of  $k\mathbf{u}^{\mathbf{s}}k_{\mathbf{L}^{2}((), \cdot, \cdot)}$ , which is provided by the following result:

Lemma 4.9. Let " > 0. Then there exists a constanC > 0, independent of", k and , such that

$$kS_{\mathbf{k}} k_{\mathbf{L}^{2}((), \cdot)} \quad \mathbf{C}^{\mathcal{D}} \overline{\mathbf{k}^{"}}(1 + \mathbf{k}^{"}) \mathbf{k}^{-1} \log (2 + (\mathbf{k}\mathbf{L})^{-1}) (1 + (\mathbf{k}\mathbf{L})^{1=2}) k k_{\mathbf{H}_{\mathbf{k}}^{-1=2}()}; \qquad 2\mathbf{H}^{-1=2}():$$
(31)

**Proof.** Arguing as in the proof of [7, Lemma 5.1 and Thm 5.2], one can show that for any " > 0 (see also [8] for slightly sharper bounds)

$$kS_k$$
 (;x<sub>2</sub>) $k_{L^2(^{-*})}$  C log (2 + (kA) <sup>1</sup>)(1 + (kA)<sup>1=2</sup> + (k/x<sub>2</sub>)<sup>1=2</sup>log (2 + kA)) k  $k_{H_k^{-1}()}$ ;

where A = L + ",  $\sim_{"} := fx 2 R$ : dist (x;  $\sim$ ) < " g,  $x_2 2 R$ , and C > 0 is independent of k, and ". From this one can show that

$$kS_{k}$$
 (;  $\mathbf{x}_{2}$ ) $k_{L^{2}((), \cdot)}$  C(1 + k") log (2 + (kL)<sup>1</sup>)(1 + (kL)<sup>1=2</sup>) k  $k_{H_{k}^{1}()}$ ;  $j\mathbf{x}_{2}j$  ";

where again **C** is independent of **k**, and ". The estimate (31) then follows from integrating over  $\mathbf{x}_2 2$  (";") and noting that  $k k_{\mathbf{H}_k^{-1}()} \mathbf{k}^{-1=2} k k_{\mathbf{H}_k^{-1=2}()} \cdots \square$ 

Combining Lemmas 4.8 and 4.9 gives:

Proposition 4.10. Under the assumptions of Lemma 4.8, we have

$$j\mathbf{u}(\mathbf{x})\mathbf{j} = \mathbf{C} + (\mathbf{k}_{\min})^{-1} \log (2 + (\mathbf{k}\mathbf{L})^{-1})(1 + \mathbf{k}\mathbf{L}); \qquad \mathbf{x} \ge (1 + (\mathbf{k}_{\min})^{-1} \log (2 + (\mathbf{k}\mathbf{L})^{-1})(1 + \mathbf{k}\mathbf{L}); \qquad \mathbf{x} \ge (1 + (\mathbf{k}_{\min})^{-1} \log (2 + (\mathbf{k}\mathbf{L})^{-1})(1 + \mathbf{k}\mathbf{L}); \qquad \mathbf{x} \ge (1 + (\mathbf{k}_{\min})^{-1} \log (2 + (\mathbf{k}\mathbf{L})^{-1})(1 + \mathbf{k}\mathbf{L}); \qquad \mathbf{x} \ge (1 + (\mathbf{k}_{\min})^{-1} \log (2 + (\mathbf{k}\mathbf{L})^{-1})(1 + \mathbf{k}\mathbf{L}); \qquad \mathbf{x} \ge (1 + (\mathbf{k}_{\min})^{-1} \log (2 + (\mathbf{k}\mathbf{L})^{-1})(1 + \mathbf{k}\mathbf{L}); \qquad \mathbf{x} \ge (1 + (\mathbf{k}_{\min})^{-1} \log (2 + (\mathbf{k}\mathbf{L})^{-1})(1 + \mathbf{k}\mathbf{L}); \qquad \mathbf{x} \ge (1 + (\mathbf{k}_{\min})^{-1} \log (2 + (\mathbf{k}\mathbf{L})^{-1})(1 + \mathbf{k}\mathbf{L}); \qquad \mathbf{x} \ge (1 + (\mathbf{k}_{\min})^{-1} \log (2 + (\mathbf{k}\mathbf{L})^{-1})(1 + \mathbf{k}\mathbf{L}); \qquad \mathbf{x} \ge (1 + (\mathbf{k}_{\min})^{-1} \log (2 + (\mathbf{k}\mathbf{L})^{-1})(1 + \mathbf{k}\mathbf{L}); \qquad \mathbf{x} \ge (1 + (\mathbf{k}_{\min})^{-1} \log (2 + (\mathbf{k}\mathbf{L})^{-1})(1 + \mathbf{k}\mathbf{L}); \qquad \mathbf{x} \ge (1 + (\mathbf{k}_{\min})^{-1} \log (2 + (\mathbf{k}\mathbf{L})^{-1})(1 + \mathbf{k}\mathbf{L}); \qquad \mathbf{x} \ge (1 + (\mathbf{k}_{\min})^{-1} \log (2 + (\mathbf{k}\mathbf{L})^{-1})(1 + \mathbf{k}\mathbf{L}); \qquad \mathbf{x} \ge (1 + (\mathbf{k}_{\min})^{-1} \log (2 + (\mathbf{k}\mathbf{L})^{-1})(1 + \mathbf{k}\mathbf{L}); \qquad \mathbf{x} \ge (1 + (\mathbf{k}_{\min})^{-1} \log (2 + (\mathbf{k}\mathbf{L})^{-1})(1 + \mathbf{k}\mathbf{L}); \qquad \mathbf{x} \ge (1 + (\mathbf{k}_{\min})^{-1} \log (2 + (\mathbf{k}\mathbf{L})^{-1})(1 + \mathbf{k}\mathbf{L}); \qquad \mathbf{x} \ge (1 + (\mathbf{k}_{\min})^{-1} \log (2 + (\mathbf{k}\mathbf{L})^{-1})(1 + \mathbf{k}\mathbf{L}); \qquad \mathbf{x} \ge (1 + (\mathbf{k}_{\min})^{-1} \log (2 + (\mathbf{k}\mathbf{L})^{-1})(1 + \mathbf{k}\mathbf{L}); \qquad \mathbf{x} \ge (1 + (\mathbf{k}_{\min})^{-1} \log (2 + (\mathbf{k}\mathbf{L})^{-1})(1 + \mathbf{k}\mathbf{L}); \qquad \mathbf{x} \ge (1 + (\mathbf{k}_{\min})^{-1} \log (2 + (\mathbf{k}\mathbf{L})^{-1})(1 + \mathbf{k}\mathbf{L}); \qquad \mathbf{x} \ge (1 + (\mathbf{k}_{\min})^{-1} \log (2 + (\mathbf{k}\mathbf{L})^{-1})(1 + \mathbf{k}\mathbf{L}); \qquad \mathbf{x} \ge (1 + (\mathbf{k}\mathbf{L})^{-1})(1 + (\mathbf{$$

where C > 0 is independent of x, k and .

**Proof.** Noting that  $\mathbf{u}^{s} = S_{\mathbf{k}} [@\mathbf{u} = \mathbf{Q}]$ , and that  $[@\mathbf{u} = \mathbf{Q}] = \mathbf{S}_{\mathbf{k}}^{1} \mathbf{u}^{i} j$ , the result follows from Lemmas 4.6(i), 4.8 and 4.9, the stability estimate (17), and the fact that  $\mathbf{k}^{"} = 3$ .

The third and nal stage in the proof of Lemma 4.5 involves combining Propositions 4.7 and 4.10 to obtain a bound which holds uniformly throughout **D**. Speci cally, we combine (32), which holds in the region d < = 32, with (22), applied in the region d = 32. Noting that in the latter case we have  $(kd)^{-1} = 32 = (k^{-1}) - C(1 + (k_{min})^{-1})$ , we can obtain the following estimate in which the constant **C** is independent of both **k** and :

$$ju(\mathbf{x})j = \mathbf{C} + \frac{1}{\mathbf{k}_{\min}} \log 1 + \frac{1}{\mathbf{k}_{\min}} (1 + \mathbf{k}\mathbf{L}); \quad \mathbf{x} \ge \mathbf{D}:$$

The statement of Lemma 4.5 then follows immediately.

# 5 hp approximation space and best approximation results

Our numerical method for solving the integral equation (13) uses a hybrid numericalasymptotic approximation space based on Theorem 4.1. Rather than approximating itself using piecewise polynomials (as in conventional methods), we use the decomposition (20), with the factors  $v_j^+$  and  $v_j^-$  replaced by piecewise polynomials. The advantage of our approach is that, as is quanti ed by Theorem 4.1, the functions  $v_j^-$  are nonoscillatory (cf. Remark 4.2), and can therefore be approximated much more eciently than the full (oscillatory) solution  $\cdot$ . Explicitly, the function we seek to approximate is

$$' (s) := \frac{1}{k} ( (x(s)) (x(s))); \quad s 2 \sim (0; L);$$
 (33)

which represents the di erence between and its GO approximation (recall Remark 4.4), scaled by 1=k so that '

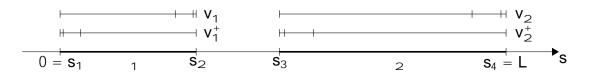


Figure 3: Illustration of the overlapping geometrically graded meshes used to approximate the amplitudes  $v_j$  in (34), in the case where comprises two components, 1 and 2

for some 2 [0; 1] and some integer **p** 0 (the highest polynomial degree on the mesh). The choice = 0 corresponds to a constant degree across the mesh (this was the only choice considered in [21]), while for 2 (0; 1] the degree decreases linearly in the direction of re nement.

For each  $j = 1; :::; n_i$  let  $n_j$  1 and  $p_j 2 (N_0)^{n_j}$  denote respectively the number of layers and the degree vector associated with the approximation of the factor  $v_j$  in (34). The total number of degrees of freedom in  $V_{N;k}$  is then

$$N := \dim(V_{N;k}) = \begin{pmatrix} 0 & 1 \\ x^{i} & 0 \\ y^{j} & 1 \end{pmatrix}_{m = 1} (p_{j}^{+})_{m} + 1 + \begin{pmatrix} x^{j} \\ p_{j} \end{pmatrix}_{m} + 1 A :$$
(37)

The regularity results provided by Theorem 4.1 allow us to prove that, under certain assumptions, the best approximation error in approximating ' by an element of  $V_{N;k}$  decays exponentially as p, the maximum degree of the approximating polynomials, increases. Our best approximation results in the space H <sup>1=2</sup>(~) are stated in the following theorem, which is the main result of this section. For simplicity of presentation we assume that the mesh parameters are the same in each of the meshes used to approximate the di erent components  $v_j$  (similar estimates hold in the more general case).

Theorem 5.1. Let  $k_{min}$   $c_0 > 0$ . Suppose that  $n_j = n$  and  $p_j = p$  for each  $j = 1; \ldots; n_i$ , where n and p are de ned by (36) with n cp for some constant c > 0. Then, for any 0 < 1=2, there exists a constant  $C_3 > 0$ , depending only on , ,  $n_1^{30}$ 

The following result is essentially stated in [6, equation (A.7)], but we need to restate it here as we are working with a  $\mathbf{k}$ -dependent norm and want  $\mathbf{k}$ -explicit estimates.

Lemma 5.3. For 1 q 2 and s < 1=2 1=q,  $L^q(R)$  can be continuously embedded in  $H^s(R)$ , with

$$k \quad k_{\mathsf{H}^{\mathsf{s}}_{\mathsf{k}}(\mathsf{R})} \quad \mathsf{c}(\mathsf{s};\mathsf{k}; \ ) \quad k \quad k_{\mathsf{L}^{\mathsf{q}}(\mathsf{R})}; \qquad 2 \,\mathsf{L}^{\mathsf{q}}(\mathsf{R}); \tag{39}$$

where is as de ned in Lemma 5.2 and

$$c(s;k;) = \frac{1}{2} \sum_{1}^{Z} (k^{2} + 2)^{s} d^{s}$$

**Proof.** By the density of  $C_0^1$  (R) in  $L^q(R)$  it su ces to prove (39) for  $2 C_0^1$  (R). Let  $2 C_0^1$  (R), let 1 < q 2 (the case q = 1 requires an obvious trivial modi cation of the proof), let r be such that 1=q+1=r=1, and let = 2=q-1 as in Lemma 5.2. Provided that s < 1=2 1=q we have s= < 1=2, so that the function  $(k^2 + 2)$  is in  $L^{1=}(R)$ , and Holder's inequality gives

$$k \ k_{H_{k}^{s}(R)}^{2} = \frac{Z}{R} (k^{2} + \frac{2}{3})^{s} j^{n} (j)^{2} d \qquad \begin{array}{c} Z & Z & Z & 2 \\ (k^{2} + \frac{2}{3})^{s} = d & (j^{n} (j)^{2})^{r=2} d \\ R & R & R \end{array}$$

$$= c(s; k; )^{2} (2) \ k^{n} k_{L^{r}(R)}^{2} \\ c(s; k; )^{2} k \ k_{L^{q}(R)}^{2};$$

the nal inequality following from an application of Lemma 5.2.

Corollary 5.4. For 1 < q = 2,  $L^{q}(R)$  can be continuously embedded in 1=2(R) with

2 *k k*<sub>H4</sub>**P**; **k** 14.98.746 cm5 Theorem 5.5. Suppose that a function g(z) is analytic in  $\operatorname{Re}[z] > 0$  and satisfies the bound

$$jg(z)j$$
  $\hat{C}jzj^{1=2}$ , Re[z] > 0;

for some  $\hat{C} > 0$ . Given I > 0, 2 [0; 1], and integers 1 and p 0, let the degree vector p be defined by (36), and suppose that cp for some constant > 0. Then for any 0 < < 1=2 there exists a constan $\hat{C} > 0$ , depending only on and (with C / 7 as / 0 or / 1=2), and a constant > 0, depending only on , , and c (with / 0 as / 0), such that

$$\inf_{v2P_{p;n}(0;l)} kg \quad vk_{H_{k}^{1=2}(0;l)} \quad C\hat{C}k^{1=2}(kl) e^{p}:$$
(42)

**Proof.** Our aim is to use Corollary 5.4 to derive a best approximation error estimate in the  $H_k^{1=2}$  norm in terms of estimates in L<sup>q</sup> norms, 1 < q < 2. For the sharpest results (in terms of k-dependence) one might want to take q = 2 in Corollary 5.4. However, this is not possible because g cannot be assumed to be square integrable at s = 0; this is why we assume that 1 < q < 2.

We begin by de ning a candidate approximant V  $2 P_{p;n}(0;I)$ , which we take to be zero on  $(0; x_1)$ , and on  $(x_{i-1}; x_i)$ , i = 2; ...; n, to be equal to the L<sup>1</sup> best approximation to  $g_{i}(x_{i-1};x_i)$  in

Now, since

$$(p)_i$$
 1 +  $\frac{(i \ 1)}{n}$  p;  $i = 2$ 

#### 6 Galerkin method

Having designed an approximation space  $V_{N;k}$  which can e-ciently approximate ', we now select an element of  $V_{N;k}$  using the Galerkin method. That is, we seek ' $_N$  2  $V_{N;k}$  H  $^{1=2}($ ) such that (recall (13) and (34))

$$h\mathbf{S}_{\mathbf{k}' \mathbf{N}}; \mathbf{v}_{i} = \frac{1}{\mathbf{k}} h\mathbf{f} \quad \mathbf{S}_{\mathbf{k}} ; \mathbf{v}_{i} ; \text{ for all } \mathbf{v} \ge V_{\mathbf{N};\mathbf{k}}:$$
(47)

We note that since '\_N; v 2 V<sub>N;k</sub> L<sup>2</sup>( ) the duality pairings in (47) can be evaluated simply as inner products in L<sup>2</sup>( ) (see the discussion after (8) and the implementation details in x7). Existence and uniqueness of the Galerkin solution '\_N is guaranteed by the Lax-Milgram Lemma and Lemmas 3.3 and 3.4. Furthermore, by Cea's lemma we have the quasi-optimality estimate

$$k' \quad '_{N} k_{H_{k}^{1=2}()} \quad \frac{C_{0}(1 + \frac{\rho_{\overline{kL}}}{2})}{2^{\rho_{\overline{2}}}} \inf_{v^{2} V_{N;k}} k' \quad v k_{H_{k}^{1=2}()}; \quad (48)$$

where  $C_0$  is the constant from Lemma 3.3. Combined with Theorem 5.1, this gives:

Theorem 6.1. Under the assumptions of Theorem 5.1, we have

$$k' \quad '_{N} k_{H_{k}^{1=2}()} \quad 1$$

An object of interest in applications is the **far eld pattern** of the scattered eld. An asymptotic expansion of the representation (12) reveals that (cf. [12])

$$\mathbf{u}^{\mathbf{s}}(\mathbf{x}) = \frac{e^{i=4}}{2} \stackrel{e^{i\mathbf{k}\mathbf{r}}}{\not\sim} \mathbf{F}(\mathbf{\hat{x}}); \text{ as } \mathbf{r} := j\mathbf{x}j \neq 1;$$

where  $\hat{\mathbf{x}} := \mathbf{x} = j \mathbf{x} j 2 \mathbf{S}^1$ , the unit circle, and

$$F(\mathbf{\hat{x}}) := \begin{array}{c} Z \\ e^{-i\mathbf{k}\cdot\mathbf{\hat{x}}\cdot\mathbf{y}} & \underline{@}\mathbf{u} \\ \hline @\mathbf{n} \end{array} (\mathbf{y}) d\mathbf{s}(\mathbf{y}); \qquad \mathbf{\hat{x}} \ 2 \ \mathbf{S}^{1}: \tag{52}$$

An approximation

Hence  $L_1 = 2$ ,  $L_2 = 2 = 5$ ,  $L_3 = 7 = 10$ ,  $L_4 = 2$  and  $L_5 = 39 = 10$ , so that the smallest component has length 2 = 5 the longest has length 39 = 10, and the sum of the length of all of the components is  $\prod_{i=1}^{n_i} L_i = 9 = 9k = 2$  (where = 2 = k is the wavelength). We present results below for values of k ranging from k = 10 (in which case the smallest segment is two wavelengths long) up to k = 10240 (in which case the longest segment is nearly 20000 wavelengths long). The plots in Figures 1 and 2 show the total elds for the \non-grazing'' incident direction d = (1 = 2; 1 = 2); in our examples below we also consider the \grazing'' incident direction d = (1; 0).

To describe our implementation of the HNA approximation space of x5, we write '  $_{\rm N}$  2 V<sub>N;k</sub> as

$$'_{N}() = \frac{X^{N}}{\sum_{i=1}^{N} v_{i}} ();$$
 (55)

where **N** is given by (37),  $\mathbf{v}$ ,  $\mathbf{v} = 1; \ldots; \mathbf{N}$ , are the unknown coeccients to be determined, and  $\mathbf{v}$ ,  $\mathbf{v} = 1; \ldots; \mathbf{N}$ , are the HNA basis functions, which we now de ne. Each basis function  $\mathbf{v}$  is supported on an interval (**a**; **b**) (**s**<sub>2</sub> 1; **s**<sub>2</sub>) for some j 2 f1;  $\ldots; \mathbf{n}_i g$ , and takes the form

$$(\mathbf{s}) = \frac{\overline{2\mathbf{q}+1}}{\mathbf{b} \mathbf{a}} \mathbf{P}_{\mathbf{q}} \quad 2 \quad \frac{\mathbf{s} \quad \mathbf{a}}{\mathbf{b} \quad \mathbf{a}} \qquad 1 \quad e^{i\mathbf{k}\mathbf{s}}; \quad \mathbf{s} \neq (\mathbf{a}; \mathbf{b});$$

where  $P_q$ , q p, denotes the Legendre polynomial of order q, and either = 1 and a  $rac{1}{2}$  1

instead, we compute  $k \ k$ , de ned by

$$k \ k := \frac{p}{jhS_k}; \ i \ j; \qquad 2 \ H^{1=2}();$$

which de nes an equivalent norm on H  $^{1=2}($ ) and is easier to compute (see, e.g., the discussion in [33, pp. A:29{A:30]}. Speci cally, it follows from (15) and (16) that

$$\frac{1}{p - \frac{1}{2^{2} - \frac{1}{2}}} k k_{H_{k}^{1=2}()} k k \qquad q \frac{1}{C_{0}(1 + \frac{p}{kL})} k k_{H_{k}^{1=2}()}; \qquad 2H^{1=2}();$$

and hence combining the right inequality with Theorem 6.1 we expect

$$k' '_{N}k C_{5}^{q} \overline{C_{0}(1 + \kappa L)} k^{-1}(1 + (\kappa L)^{3=2+}) e^{p}$$
: (57)

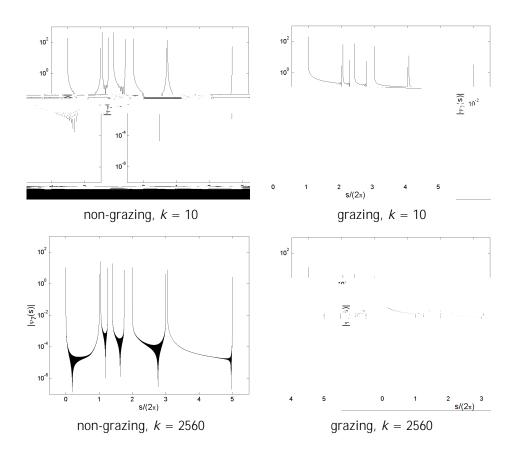
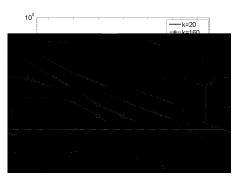


Figure 4: Boundary solution for grazing and non-grazing incidence, with  $\mathbf{k} = 10$  and  $\mathbf{k} = 2560$ .

In Table 1 we also show the condition number (COND) of the N -dimensional linear system (56), and we investigate the dependence of the condition number on both k and p further in Figure 6. For xed k, the condition number grows exponentially with respect to p (note the logarithmic scale on the vertical axis). This rapid growth in the condition number as p increases is not surprising: for weakly singular BIEs of the rst kind, the condition number for standard hp Galerkin BEM, with a geometrically graded mesh (as used here), is known to grow exponentially with respect to the number of unknowns (see, e.g., [18]). For xed p, the condition number decreases slowly as k increases (and hence as the average number of degrees of freedom per wavelength decreases), and we note that the condition numbers we encountered in our experiments were not so large as to cause problems for our direct solver. Furthermore, as remarked in *x*1, our best approximation results (though not our full analysis) hold regardless of the BIE used, so using our approximation space within a better conditioned BIE such as the second kind formulations proposed in [4, 25] might lead to reduced condition numbers.

Finally, in the last column of Table 1 we show the relative computing time (rel cpt) required for setting up and solving the linear system (we solve the system directly), measured with respect to the time required for  $\mathbf{k} = 10$ . We emphasize the fact that the computing time is independent of  $\mathbf{k}$ , re ecting that all of the integrals are evaluated using Filon quadrature in a k-independent way.



е<sub>р</sub>

Figure 6: Condition number of theN-dimensional linear system (56).

non-grazing, max\_{2 [0;28 ]} ju\_7(t) grazing, max\_{2 [0;28 ]} ju\_7(t) u\_p(t) j u\_p(t) j

non-grazing, max

d	k	$\frac{P N}{\sum_{j=1}^{n_i} L_j = 1}$	ep		r <sub>p</sub>	COND	rel cpt
$p^{1}; p^{1}$	10	10.00	9.25 10 <sup>4</sup>	-1.03	2.18 10 <sup>3</sup>	1.50 10 <sup>9</sup>	1.00
	20	5.00	4.51 10 <sup>4</sup>	-0.77	2.01 10 <sup>3</sup>	1.03 10 <sup>9</sup>	0.98
	40	2.50	2.64 10 <sup>4</sup>	-0.87	2.49 10 <sup>3</sup>	7.12 10 <sup>8</sup>	0.98
	80	1.25	1.45 10 <sup>4</sup>	-0.69	2.88 10 <sup>3</sup>	4.97 10 <sup>8</sup>	0.99
	160	0.63	8.99 10 <sup>5</sup>	-0.80	3.72 10 <sup>3</sup>	3.50 10 <sup>8</sup>	1.00
	320	0.31	5.16 10 <sup>5</sup>	-0.74	4.32 10 <sup>3</sup>	2.47 10 <sup>8</sup>	0.99
	640	0.16	3.08 10 <sup>5</sup>	-0.74	5.08 10 <sup>3</sup>	1.75 10 <sup>8</sup>	1.00
	1280	0.08	1.85 10 <sup>5</sup>	-0.67	6.48 10 <sup>3</sup>	1.23 10 <sup>8</sup>	1.00
	2560	0.04	1.16 10 <sup>5</sup>	-0.91	7.64 10 <sup>3</sup>	8.72 10 <sup>7</sup>	1.00
	5120	0.02	6.18 10 <sup>6</sup>	-0.83	9.14 10 <sup>3</sup>	6.17 10 <sup>7</sup>	1.01
	10240	0.01	3.47 10 <sup>6</sup>		1.01 10 <sup>2</sup>	4.36 10 <sup>7</sup>	1.01
(1;0)	10	10.00	3.39 10 <sup>4</sup>	-0.38	4.52 10 <sup>4</sup>	1.50 10 <sup>9</sup>	1.00
	20	5.00	2.60 10 <sup>4</sup>	-0.61	5.84 10 <sup>4</sup>	1.03 10 <sup>9</sup>	1.01
	40	2.50	1.70 10 <sup>4</sup>	-0.60	6.43 10 <sup>4</sup>	7.12 10 <sup>8</sup>	0.99
	80	1.25	1.12 10 <sup>4</sup>	-0.71	7.13 10 <sup>4</sup>	4.97 10 <sup>8</sup>	0.98
	160	0.63	6.84 10 <sup>5</sup>	-0.69	7.31 10 <sup>4</sup>	3.50 10 <sup>8</sup>	0.99
	320	0.31	4.23 10 <sup>5</sup>	-0.68	7.59 10 <sup>4</sup>	2.47 10 <sup>8</sup>	0.99
	640	0.16	2.64 10 <sup>5</sup>	-0.72	7.97 10 <sup>4</sup>	1.75 10 <sup>8</sup>	1.00
	1280	0.08	1.60 10 <sup>5</sup>	-0.62	8.13 10 <sup>4</sup>	1.23 10 <sup>8</sup>	1.00
	2560	0.04	1.04 10 <sup>5</sup>	-0.73	8.92 10 <sup>4</sup>	8.72 10 <sup>7</sup>	1.00
	5120	0.02	6.27 10 <sup>6</sup>	-0.73	9.02 10 <sup>4</sup>	6.17 10 <sup>7</sup>	1.01
	10240	0.01	3.78 10 <sup>6</sup>		9.14 10 <sup>4</sup>	4.36 10 <sup>7</sup>	1.00

Table 1: Errors  $\mathbf{e}_{\mathbf{p}}$  and relative errors  $\mathbf{r}_{\mathbf{p}}$ , for non-grazing ( $\mathbf{d} = (1 = \sqrt[p]{2}; 1 = \sqrt[p]{2})$ ) and grazing ( $\mathbf{d} = (1; 0)$ ) incidence, with  $\mathbf{p} = 5$  (and hence  $\mathbf{N} = 450$ ).

far eld pattern computed with our nest discretization) for each of the two incident directions, for  $\mathbf{k} = 1280$ , are shown in Figure 8. For non-grazing incidence, the peaks corresponding to the geometric shadow (i.e. the forward-scattering direction) and the specular re ection are indicated (compare Figure 8 with Figure 1). We also show the points at which  $\hat{\mathbf{x}}(t) \ge 1$ . For grazing incidence, the shadow peak is much lower for than for non-grazing incidence; in the grazing case, there is no re ected peak.

In Figure 9 we plot approximations to  $kF_7 = F_p k_{L^1} (s^1)$  and  $kF_7 = F_p k_{L^1} (s^1) = kF_7 k_{L^1} (s^1)$ for  $\mathbf{k} = 20$ , 80, 320 and 1280, for each of the two incident directions. To approximate the  $L^1$  norm, we compute  $F_7$  and  $F_p$  at 50,000 evenly spaced points on the unit circle. The exponential decay as  $\mathbf{p}$  increases, as predicted by Theorem 6.3, can be clearly seen (again, note the logarithmic scale on the vertical axes).

For xed **p**, the errors  $k\mathbf{F}_7 = \mathbf{F}_p k_{L^1} (s^1)$  increase slowly as **k** increases. To investigate this behaviour more carefully, in Table 2 we show results for the two angles of incidence for  $\mathbf{p} = 5$  (and hence **N** 

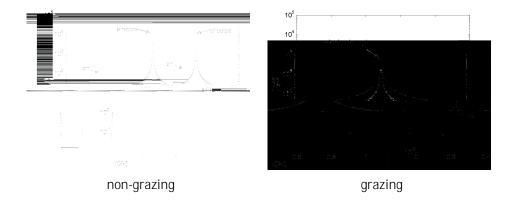
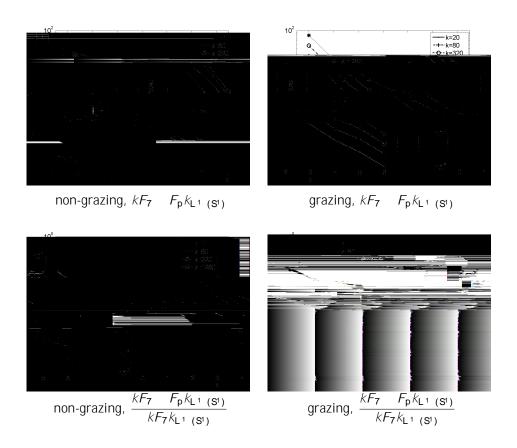
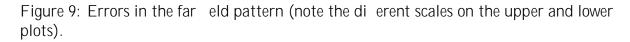


Figure 8: Far eld patterns,  $j\mathbf{F}_{7}(\mathbf{t})j = j\mathbf{F}(\mathbf{t})j$ ,  $\mathbf{k} = 1280$ .





that  $f_p(k) = k$  as  $k \neq 7$ . The values of 2 (0:02; 0:75) for non-grazing incidence, and 0:5 for grazing incidence, are considerably lower than might be anticipated from the estimate (54), suggesting that our estimates are not sharp in terms of their k-dependence. In particular, the results are again consistent with the conjecture that M(u) = O(1) (as discussed just before Lemma 4.1). In the last column of Table 2, we show how  $kF_7k_{L^1}$  ( $s^1$ ) grows with k. For non-grazing incidence,  $kF_7k_{L^1}$  ( $s^1$ ) grows approximately linearly with k, and so the relative error  $k_{T_1} = kF_7k_{L^1}$  ( $s^1$ ) decreases as k increases. For grazing incidence,  $kF_7k_{L^1}$  ( $s^1$ )

optimality estimate (48).

## Acknowledgements

We gratefully acknowledge the support of EPSRC grant EP/F067798/1, for all authors, and of EPSRC grant EP/K000012/1, for SL. We thank the anonymous reviewers for their comments, and we thank Nick Biggs, Alexey Chernov, Peter Svensson and Ashley Twigger for helpful discussions.

### References

- [1] A. G. Abul-Azm and A. N. Williams, Oblique wave di raction by segmented o shore breakwaters Ocean Eng., 24 (1997), pp. 63{82.
- [2] N. R. T. Biggs, D. Porter, and D. S. G. Stirling, Wave di raction through a perforated breakwaterQ. Jl. Mech. Appl. Math., 53 (2000), pp. 375{391.

[3]

[11]

- [26] C. Linton and P. McIver, Handbook of Mathematical Techniques for Wave/Structure Interactions, CRC Press, 2001.
- [27] W. McLean, Strongly Elliptic Systems and Boundary Integral EquationsCUP, 2000.
- [28] J. F. Nye, Numerical solution for di raction of an electromagnetic wave by slits in a perfectly conducting screenProc. R. Soc. Lond. A, 458 (2002), pp. 401{427.
- [29] S. A. Sauter and C. Schwab, **Boundary element methods**vol. 39 of Springer Series in Computational Mathematics, Springer-Verlag, Berlin, 2011. Translated and expanded from the 2004 German original.
- [30] V. M. Serdyardschargedisolutions for electromagnetic wave di raction by a slot and strip

- [42] S. N. Vorobyov and L. M. Lytvynenko, Electromagnetic wave di raction by semi-in nite strip grating, IEEE Trans. Ant. Prop., 59 (2011), pp. 2169{2177.
- [43] P. Wolfe, The diraction of waves by slits and strips SIAM J. Appl. Math., 19 (1970), pp. 20{32.