## JUST ANOTHER IMPLEMENTATION OF THE FAST MULTIPOLE METHOD FOR THE ELECTROSTATIC POTENTIAL IN 2D

RODRIGO ARRIETA\*

Abstract.

## 1. Introduction.

2. The multipole expansion for well separated clusters of points. Consider the electrostatic potential  $\{u_i\}_{i=1}^M \subset \mathbb{R}$  at some target points  $\{x\}_{i=1}^M \subset \mathbb{R}^2$  generated by a set of point charges  $\{q_j\}_{j=1}^N \subset R$  located at source points  $\{y_j\}_{j=1}^N$ . Mathematically, this is expressed as

(2.1) 
$$u(\mathbf{x}_i) := u_i = \sum_{j=1}^N G(\mathbf{x}_i, \mathbf{y}_j) q_j, \quad i = 1, 2, \dots, M,$$

where the kernel G is taken as a scaled version of the fundamental solution of the Laplace equation in two dimensions,

(2.2) 
$$G(\boldsymbol{x}, \boldsymbol{y}) = \begin{cases} \log |\boldsymbol{x} - \boldsymbol{y}|, & \text{if } \boldsymbol{x} \neq \boldsymbol{y}, \\ 0, & \text{if } \boldsymbol{x} = \boldsymbol{y}. \end{cases}$$

Clearly, a direct evaluation of (2.1) would take  $\mathcal{O}(NM)$  operations, which is prohibitively expensive when we have a large number of source and target points N, M >> 1. In what follows, we will describe how to compute an arbitrarily close approximation to this sum in a much faster way, subject to the constraint that source points are far away from target points.

Suppose that the source points are clustered in a box  $\Omega_{\sigma}$ , with center  $\boldsymbol{c}_{\sigma}$ , whereas the target points are clustered in a disjoint box  $\Omega_{\tau}$  of the same size, with center  $\boldsymbol{c}_{\tau}$ , as shown in Fig. 1. The separation between boxes is  $d := |\boldsymbol{c}_{\tau} - \boldsymbol{c}_{\sigma}|$ .

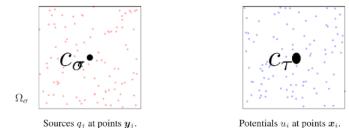


Figure 1. The problem of computing the electrostatic potential between well-separated source and target boxes.

 $\Omega_{\tau}$ 

For simplicity, let us switch to complex notation, where each point  $x, y \in \mathbb{R}^2$  is reinterpreted as a point in the complex plane  $\mathbb{C}$ , and with abuse of notation we will write the complex kernel  $G(x, y) = \log(x - y)$ , where log is the principal branch of the complex logarithm. The real kernel in (2.2) is retrieved by taking Re  $\{G(x, y)\}$ .

<sup>\*</sup>Mathematics Department, Massachusetts Institute of Technology, Cambridge, MA 02139, United States (rarrieta@mit.edu).

We can now formally expand the potential  $u_i$  in a Taylor expansion,

$$u_{i} = \sum_{j=1}^{N} \log(\boldsymbol{x}_{i} - \boldsymbol{y}_{j}) q_{j}$$

$$= \sum_{j=1}^{N} \log((\boldsymbol{x}_{i} - \boldsymbol{c}_{\sigma}) - (\boldsymbol{y}_{j} - \boldsymbol{c}_{\sigma})) q_{j}$$

$$= \sum_{j=1}^{N} \left[ \log(\boldsymbol{x}_{i} - \boldsymbol{c}_{\sigma}) + \log\left(1 - \frac{\boldsymbol{y}_{j} - \boldsymbol{c}_{\sigma}}{\boldsymbol{x}_{i} - \boldsymbol{c}_{\sigma}}\right) \right] q_{j}$$

$$= \sum_{j=1}^{N} \left[ \log(\boldsymbol{x}_{i} - \boldsymbol{c}_{\sigma}) - \sum_{p=1}^{\infty} \frac{1}{p} \frac{(\boldsymbol{y}_{j} - \boldsymbol{c}_{\sigma})^{p}}{(\boldsymbol{x}_{i} - \boldsymbol{c}_{\sigma})^{p}} \right] q_{j}$$

$$\approx \log(\boldsymbol{x}_{i} - \boldsymbol{c}_{\sigma}) \hat{q}_{0}^{\sigma} + \sum_{p=1}^{P} \frac{1}{p} \frac{1}{(\boldsymbol{x}_{i} - \boldsymbol{c}_{\sigma})^{p}} \hat{q}_{p}^{\sigma},$$

where we used the fact that  $\log(1-z)=-\sum_p z^p/p$  for |z|<1, the distance d is chosen such that  $\left|\frac{\pmb{y}_j-\pmb{c}_\sigma}{\pmb{x}_i-\pmb{c}_\sigma}\right|<1$  for every pair of target/source points, and we have approximated the infinite sum by P terms, where P is known as the *interaction rank*. The quantities  $\hat{\pmb{q}}^\sigma:=\{\hat{q}_p^\sigma\}_{p=0}^{P-1}$  given by

(2.4) 
$$\hat{q}_0^{\sigma} := \sum_{j=1}^N q_j, \\ \hat{q}_p^{\sigma} := \sum_{j=1}^N \frac{-1}{p} (\boldsymbol{y}_j - \boldsymbol{c}_{\sigma})^p q_j, \quad p = 1, 2, \dots, P - 1,$$

are the *outgoing expansions* of the source box  $\sigma$ . On the other hand, the potential u(x) is analytic, thus it can be expanded as a (truncated) Taylor series around the target box center  $\mathbf{c}_{\tau}$ ,

(2.5) 
$$u(\boldsymbol{x}_i) = u_i \approx \sum_{p=0}^{P-1} (\boldsymbol{x}_i - \boldsymbol{c}_{\tau})^p \hat{\boldsymbol{v}}_p^{\tau},$$

where  $\hat{\boldsymbol{v}}^{\tau} := \{\hat{v}_p^{\tau}\}_{p=0}^{P-1}$  are the *incoming expansions* of the target box  $\tau$ . By expanding (2.3) in Taylor, and again choosing the distance d large enough, it can be deduced that (2.6)

$$\hat{v}_0^{ au} = \log(m{c}_{ au} - m{c}_{\sigma})\hat{q}_0^{\sigma} + \sum_{p=1}^{P-1} (-1)^p rac{1}{(m{c}_{\sigma} - m{c}_{ au})^p} \hat{q}_p^{\sigma},$$

$$\hat{v}_r^{\tau} = -\frac{1}{r(\boldsymbol{c}_{\sigma} - \boldsymbol{c}_{\tau})^r} \hat{q}_0^{\sigma} + \sum_{p=1}^{P-1} (-1)^p \binom{r+p-1}{p-1} \frac{1}{(\boldsymbol{c}_{\sigma} - \boldsymbol{c}_{\tau})^{r+p}} \hat{q}_p^{\sigma}, \quad r = 1, 2, \dots, P-1.$$

Denote  $N_{\tau}$  and  $N_{\sigma}$  the numbers of target and source points, respectively,  $q^{\sigma} :=$ 

 $\{q_j\}_{j=1}^{N_\sigma}$  and  $\boldsymbol{u}^\tau := \{u_i\}_{i=1}^{N_\tau}$ . Note that we have implicitly defined the linear maps

$$\hat{\boldsymbol{q}}^{\sigma} = \mathbf{T}_{\sigma}^{\text{ofs}} \boldsymbol{q}^{\sigma},$$

(2.8) 
$$\hat{\boldsymbol{v}}^{\tau} = \mathbf{T}_{\tau,\sigma}^{\text{ifo}} \hat{\boldsymbol{q}}^{\sigma} \quad \text{and} \quad$$

(2.9) 
$$\boldsymbol{u}^{\tau} = \mathbf{T}_{\tau}^{\text{tfi}} \hat{\boldsymbol{v}}^{\tau},$$

where  $\mathbf{T}_{\sigma}^{\text{ofs}}$  is the  $P \times N_{\sigma}$  outgoing-from-sources translation operator defined by (2.4),  $\mathbf{T}_{\sigma}^{\text{ifo}}$  is the  $P \times P$  incoming-from-outgoing translation operator defined by (2.6) and  $\mathbf{T}_{\sigma}^{\text{tfi}}$  is the  $N_{\tau} \times P$  target-from-incoming translation operator defined by (2.5). With this, we have effectively computed a rank-P approximation of the potential

(2.10) 
$$\boldsymbol{u}^{\tau} \approx \mathbf{T}_{\tau}^{\text{tfi}} \left( \mathbf{T}_{\tau,\sigma}^{\text{ifo}} \left( \mathbf{T}_{\sigma}^{\text{ofs}} \boldsymbol{q}^{\sigma} \right) \right),$$

which takes  $\mathcal{O}\left(P(N_{\sigma}+N_{\tau})+P^2\right)$  operations, much less than the  $\mathcal{O}\left(N_{\sigma}N_{\tau}\right)$  operations of the naive method for (2.1).

Let us introduce the concept of well separated boxes, which gives us an estimate of the error committed in (2.10). We will say that a box  $\Omega_{\tau}$  of center  $c_{\tau}$  is well separated from a box  $\Omega_{\sigma}$  of center  $c_{\sigma}$  and side length 2a if

$$||\boldsymbol{c}_{\tau} - \boldsymbol{c}_{\tau}|| \ge 4a.$$

Now, for well separated source and target boxes of the same size,  $\Omega_{\sigma}$  and  $\Omega_{\tau}$ , the error committed in (2.10) using a rank-P approximate potential  $\boldsymbol{u}_{P}^{\tau}$  is given by

$$\left\|\boldsymbol{u}^{\tau}-\boldsymbol{u}_{P}^{\tau}\right\|_{\infty} \leq \frac{\eta^{P}}{P(1-\eta)}\left\|\boldsymbol{q}^{\sigma}\right\|_{1},$$

where  $\eta = \sqrt{2}/(4 - \sqrt{2}) \approx 0.547$ . This bound can be derived by considering the remainders of the Taylor expansions of (2.5) and (2.3), and the definition of well separated boxes. It follows that the error decays exponentially with respect to the interaction rank P, and we can pick  $P = \mathcal{O}(\log(1/\epsilon))$  to achieve an error of  $\epsilon$ .

3. The Fast Multipole Method for an uniform cluster of charges. Consider an uniform cluster of N points  $x_i$  with associated charges  $q_i$ , i = 1, ..., N, as shown in Figure 2. We wish to compute the electrostatic potential at every point  $x_i$ , this is

(3.1) 
$$u(\mathbf{x}_i) := u_i = \sum_{j=1}^{N} G(\mathbf{x}_i, \mathbf{x}_j) q_j, \quad i = 1, 2, \dots, N.$$

As before, computing the potentials using the naive formula (3.1) takes  $\mathcal{O}(N^2)$  operations. Here, we will describe the Fast Multipole Method (FMM), which allows us to compute all potentials with only  $\mathcal{O}(N)$ , using the multipole technique presented in section 2.

In the FMM we construct a quadtree of boxes, a hierarchical tree data structure, where we start with a root box which is divided into four smaller and identical children boxes. The children constitute the first level of the tree. To construct the second level, each child is again subdivided into four smaller and identical children. This process gets repeated until we reach a prescribed number of levels L+1, where level 0 corresponds

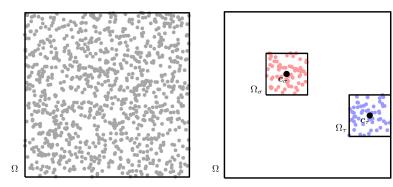


FIGURE 2. Left: uniform cluster of charges. Right: for well separated boxes inside the domain, we could resort to the fast multipole approach of section 2.

Level 0	Level 1			Level 2					Level 3								
1		3	5		11	13	19	21		43	45	51	53	75	77	83	85
										42	44	50	52	74	76	82	84
					10	12	18	20		39	41	47	49	71	73	79	81
										38	40	46	48	70	72	78	80
		2	4		7	9	15	17		27	29	35	37	59	61	67	69
										26	28	34	36	58	60	66	68
					6	8	14	16		23	25	31	33	55	57	63	65
										22	24	30	32	54	56	62	64

Figure 3. Quadtree of boxes with L+1 levels, L=3.

only to the root box, whereas level L contains the finer boxes, the *leaves*. An example of a quadtree structure with L=3 is shown in Figure 3.

The idea of this hierarchical data structure is the following. Leaf boxes compute the potentials at their points using multipole expansions for well separated boxes and direct evaluations for neighboring boxes and self-interactions. For closer well separated boxes we use the multipole expansion shown in section 2. This set of closer well separated boxes is known as the interaction list  $\mathcal{L}_{\sigma}^{\text{int}}$  of the leaf box  $\sigma$ , we will provide a precise definition below. Instead, for farther away boxes, it is much more efficient that the leaves' parents compute the multipole interactions (outgoing or incoming expansions) for the parent's interaction list, and then they pass the interaction to their children. Again, if two well separated parent boxes are very far away (meaning that they do not belong to each others interaction list) then it is more efficient that the parents' parents compute the interaction and so on. This hierarchical transfer of information is the secret behind FMM which allows to achieve a  $\mathcal{O}(N)$  complexity. In the end, boxes only care about boxes in their interaction list. If they are farther away, then their ancestry (parents, grandparents, etc.) compute the interaction. And for leaf boxes, they have to care about their interaction list, neighbors and self interactions to finally compute the potentials at their points.

Before describing the FMM algorithm, we will explain how children/parents boxes transfer their outgoing/incoming expansions to each other. First, a parent box  $\tau$  needs to compute its outgoing expansion  $\hat{q}^{\tau}$ . In fact, this can be done exactly (without any approximation) using the outgoing expansions of their children, as shown next. Fix a child box  $\sigma$  of  $\tau$ . Denote  $I_{\sigma} = \{i : x_i \in \Omega_{\sigma}\}$ . Its outgoing expansion of order p is

(3.2) 
$$\hat{q}_p^{\sigma} = \sum_{j \in I_{\sigma}} \frac{-1}{p} (\boldsymbol{x}_j - \boldsymbol{c}_{\sigma})^p q_j,$$

where p = 1, 2, ..., P-1. On the other hand, we can write the order k = 1, 2, ..., P-1 outgoing expansion of  $\tau$  given by the charges at  $\Omega_{\sigma}$  as

(3.3) 
$$\hat{q}_k^{\tau,\sigma} = \sum_{j \in I_{\sigma}} \frac{1}{k} (\boldsymbol{y}_j - \boldsymbol{c}_{\tau})^k q_j$$

$$(3.4) \qquad = \sum_{j \in I_{\sigma}} \frac{-1}{k} ((\boldsymbol{y}_{j} - \boldsymbol{c}_{\sigma}) - (\boldsymbol{c}_{\sigma} - \boldsymbol{c}_{\tau}))^{k} q_{j}$$

$$(3.5) \qquad = \sum_{j \in I_{\sigma}} \frac{-1}{k} \sum_{p=0}^{k} {k \choose p} (\boldsymbol{c}_{\sigma} - \boldsymbol{c}_{\tau})^{k-p} (\boldsymbol{y}_{j} - \boldsymbol{c}_{\sigma})^{p} q_{j}$$

(3.6) 
$$= \frac{-1}{k} (\boldsymbol{c}_{\sigma} - \boldsymbol{c}_{\tau})^{k} \hat{q}_{0}^{\sigma} + \sum_{p=1}^{k} \frac{-1}{k} {k \choose p} (\boldsymbol{c}_{\sigma} - \boldsymbol{c}_{\tau})^{k-p} \sum_{j \in I_{\sigma}} (\boldsymbol{y}_{j} - \boldsymbol{c}_{\sigma})^{p}$$

$$(3.7) \qquad = \frac{-1}{k} (\boldsymbol{c}_{\sigma} - \boldsymbol{c}_{\tau})^{k} \hat{q}_{0}^{\sigma} + \sum_{p=1}^{k} \frac{p}{k} {k \choose p} (\boldsymbol{c}_{\sigma} - \boldsymbol{c}_{\tau})^{k-p} \sum_{j \in I_{\sigma}} \frac{-1}{p} (\boldsymbol{y}_{j} - \boldsymbol{c}_{\sigma})^{p}$$

(3.8) 
$$= \frac{-1}{k} (\boldsymbol{c}_{\sigma} - \boldsymbol{c}_{\tau})^{k} \hat{q}_{0}^{\sigma} + \sum_{p=1}^{k} \frac{p}{k} {k \choose p} (\boldsymbol{c}_{\sigma} - \boldsymbol{c}_{\tau})^{k-p} \hat{q}_{p}^{\sigma}.$$

This defines a linear operator  $\mathbf{T}_{\tau,\sigma}^{\text{ofo}}$ , known as the  $P \times P$  outgoing-from-outgoing translation operator, which maps the outgoing expansion  $\hat{q}^{\sigma}$  of the child  $\sigma$  into the parent  $\tau$  expansion  $\hat{q}^{\tau}$ . This definition of  $\mathbf{T}_{\tau,\sigma}^{\text{ofo}}$  is slightly different to the one shown in [1]. Finally, the total outgoing expansion of  $\tau$  is given by the sum of the contributions of their children, this is

(3.9) 
$$\hat{\boldsymbol{q}}^{\tau} = \sum_{\sigma \in \mathcal{L}^{\text{child}}} \mathbf{T}_{\tau,\sigma}^{\text{ofo}} \hat{\boldsymbol{q}}^{\sigma},$$

where  $\mathcal{L}_{\tau}^{\text{child}}$  is the set of children of  $\tau$ .

Once a parent  $\tau$  computes its incoming expansion  $\hat{\boldsymbol{v}}^{\tau}$  using the outgoing expansions of the boxes in its interaction list, it needs to transfer this incoming expansion to each of its four children. Fix a children  $\sigma$  of  $\tau$ . Its incoming expansion is given by

(3.10) 
$$\hat{\boldsymbol{v}}^{\sigma} = \mathbf{T}_{\tau,\sigma}^{\text{iff}} \hat{\boldsymbol{v}}^{\tau} + \sum_{\nu \in \mathcal{L}^{\text{int}}} \mathbf{T}_{\sigma,\nu}^{\text{ifo}} \hat{\boldsymbol{q}}^{\nu},$$

where the first term is the contribution from the parent (this includes the contribution of boxes farther away than the interaction list  $\mathcal{L}_{\sigma}^{\text{int}}$  of  $\sigma$ ), and the second term correspond to the contribution of the interaction list  $\mathcal{L}_{\sigma}^{\text{int}}$  of  $\sigma$ . The linear operator  $\mathbf{T}_{\sigma,\tau}^{\text{iff}}$  is the  $P \times P$  incoming-from-incoming translation operator, which maps the incoming expansion  $\hat{\boldsymbol{v}}^{\tau}$  of the parent  $\tau$  into the child  $\sigma$  expansion  $\hat{\boldsymbol{v}}^{\sigma}$ . Using the expression (2.5) and the same binomial expansion tricks as used in (3.8), it can be derived that

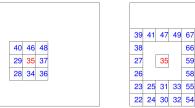
(3.11) 
$$\mathbf{T}_{\sigma,\tau}^{\text{ifi}} = \begin{cases} \binom{p}{r} (\boldsymbol{c}_{\sigma} - \boldsymbol{c}_{\tau})^{p-r}, & \text{for } r \leq p, \\ 0, & \text{for } r > p, \end{cases}$$

for  $p, r = 0, 1, \dots, P - 1$ .

For a box  $\tau$  let us define the following:

• The parent of  $\tau$  is the box on the next coarsest level that contains  $\tau$ .

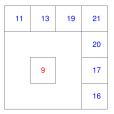
- The children list of τ is the set L<sub>τ</sub><sup>child</sup> of boxes whose parent is τ.
  The neighbor list of τ is the set L<sub>τ</sub><sup>nei</sup> of boxes on the same level that directly
- The interaction list of  $\tau$  is the set  $\mathcal{L}_{\tau}^{\rm int}$  of all boxes  $\sigma$  such that (1)  $\sigma$  and  $\tau$  are on the same level, (2)  $\sigma$  and  $\tau$  do not touch, and (3) the parents of  $\sigma$  and
- Denote  $A(I_{\tau}, I_{\sigma})$  as the (dense) operator that maps the charges  $q^{\sigma}$  at box  $\sigma$ into the potentials  $\boldsymbol{u}^{\tau}$  at box  $\tau$ , i.e.,  $\boldsymbol{u}^{\tau} = A(I_{\tau}, I_{\sigma})\boldsymbol{q}^{\sigma}$ .



For  $\tau = 35$  (red), the neighbor list  $\mathcal{L}_{\tau}^{\mathrm{nei}}$  is shown in blue.

66 68 59 61 22 24 30 32 54 56

For  $\tau = 35$  (red), the interaction list  $\mathcal{L}_{\tau}^{\mathrm{int}}$  is shown in blue.



For  $\tau = 9$  (red), the interaction list  $\mathcal{L}_{\tau}^{\mathrm{int}}$  is shown in blue.

Figure 4. Examples of interaction and neighbor lists.

A diagram that shows an example of interaction and neighbor lists is presented in Figure 4. Now we have all the tools to describe the FMM algorithm. In the precomputation step we first assemble the quadtree, the boxes lists and precompute all of the translation operators. Now, to compute the potentials u for a given set of charges q we follow three steps:

- 1. Upward pass: compute the outgoing expansions of all boxes. Start at the finest level computing the expansions directly from the charges. For parent boxes, compute their expansions using their children's expansions.
- 2. Downward pass: compute the incoming expansions of all boxes. From level l=2 to the finest, compute the *incoming expansions* using the parent's contribution and the interaction list's contribution.
- 3. Compute potentials: Each leaf computes the potentials at its points, using the incoming expansions and direct evaluations of the potentials for neighboring boxes and self-interactions.

The FMM algorithm is detailed in Figure 5, Figure 6 and Figure 7.

```
loop over levels \ell, from the finest to the coarsest
         loop over all boxes \tau on level \ell
                  if (\tau is a leaf)
                           \hat{\mathbf{q}}_{	au} = \mathbf{T}_{	au}^{	ext{ofs}} \, \mathbf{q}(I_{	au})
                          \hat{\hat{\mathbf{q}}}_{	au} = \sum_{\sigma \in \mathcal{L}_{	au}^{	ext{child}}} \mathsf{T}_{	au,\sigma}^{	ext{ofo}} \, \hat{\mathbf{q}}_{\sigma}
                  end if
         end loop
end loop
```

FIGURE 5. FMM upward pass.

```
Set \hat{\mathbf{u}}_{\tau} = \mathbf{0} for every box \tau on level 1. 

loop over levels \ell, from level \ell = 2 to the finest 

loop over all boxes \tau on level \ell 

Let \nu denote the parent of \tau. 

\hat{\mathbf{u}}_{\tau} = \mathbf{T}_{\tau,\nu}^{\mathrm{ifi}} \hat{\mathbf{u}}_{\nu} + \sum_{\sigma \in \mathcal{L}_{\tau}^{\mathrm{int}}} \mathbf{T}_{\tau,\sigma}^{\mathrm{ifo}} \hat{\mathbf{q}}_{\sigma}. 

end loop 

end loop
```

Figure 6. FMM downward pass.

$$\begin{array}{l} \textbf{loop} \text{ over every leaf box } \tau \\ \textbf{u}(I_\tau) = \textbf{A}_\tau^{\mathrm{tfi}} \, \hat{\textbf{u}}_\tau + \textbf{A}(I_\tau,I_\tau) \, \textbf{q}(I_\tau) + \sum_{\sigma \in \mathcal{L}_\tau^{\mathrm{nei}}} \, \textbf{A}(I_\tau,I_\sigma) \, \textbf{q}(I_\sigma) \\ \textbf{end loop} \end{array}$$

Figure 7. FMM compute potentials step.