

Method of Moments and Finite Element Methods

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Linear Operator Equations

- A linear operator equation is represented symbolically as

$$\mathcal{L}u = f \quad (1)$$

where \mathcal{L} is a linear operator,

$u = u(\mathbf{r})$ = unknown field or current, $\mathbf{r} \in \mathcal{D}$

$f = f(\mathbf{r})$ = known source, incident field or other
forcing function, $\mathbf{r} \in \mathcal{D}$

- Operator \mathcal{L} is linear if it satisfies

$$\mathcal{L}(au_1 + bu_2) = a\mathcal{L}u_1 + b\mathcal{L}u_2$$

- \mathcal{L} is generally a differential, integral, or integro - differential operator on \mathcal{D} relating sources f to fields u in \mathcal{D} or on its boundary, $\partial\mathcal{D}$. Matrices are also linear operators.

Linear Operator Examples

- $\mathcal{L} V = \frac{d^2 V}{dx^2} + k_0^2 V + \text{BC's}$
- $\mathcal{L} q_s = \int_s \frac{q_s(\mathbf{r}')}{4\pi\epsilon |\mathbf{r} - \mathbf{r}'|} dS'$
- $\mathcal{L} \mathbf{J} = j\omega\mu \int_s G(\mathbf{r}, \mathbf{r}') \mathbf{J}(\mathbf{r}') dS' - \frac{\nabla}{j\omega\epsilon} \int_s G(\mathbf{r}, \mathbf{r}') \nabla' \cdot \mathbf{J}(\mathbf{r}') dS' + \text{BC's}$
- $\mathcal{L} \mathbf{J} = \frac{\mathbf{J}(\mathbf{r})}{2} - \hat{\mathbf{n}} \times \int_s \nabla G(\mathbf{r}, \mathbf{r}') \times \mathbf{J}(\mathbf{r}') dS'$
- $\mathcal{L} \Phi = \nabla^2 \Phi + k_0^2 \epsilon_r \Phi + \text{BC's}$
- $\mathcal{L} \mathbf{E} = \nabla \times \mu_r^{-1} \nabla \times \mathbf{E} - k_0^2 \epsilon_r \mathbf{E} + \text{BC's}$
- $\mathcal{L} [x_m] = [L_{mn}] [x_m]$

In statics :

$$G(\mathbf{r}, \mathbf{r}') = \frac{1}{4\pi R}, \quad R = |\mathbf{r} - \mathbf{r}'|, \text{ (3D)}$$

$$= -\frac{1}{2\pi} \ln D, \quad D = |\boldsymbol{\rho} - \boldsymbol{\rho}'|, \text{ (2D)}$$

In dynamics :

$$G(\mathbf{r}, \mathbf{r}') = \frac{e^{-jkR}}{4\pi R}, \quad \text{(3D)}$$

$$= -\frac{H_0^{(2)}(kD)}{4j}, \quad \text{(2D)}$$

Inner Products

- Numerical solution methods are *projection* or *moment methods* involving *inner products*.

- Inner (dot) product between pairs of N - component vectors,

$$u = (u_1, u_2, \dots, u_N), \quad v = (v_1, v_2, \dots, v_N):$$

$$u \cdot v = \langle u, v \rangle = \sum_{n=1}^N u_n v_n = [u_n]^t [v_n] \quad (3)$$

- Inner product between two scalar functions u and v :

$$\langle u, v \rangle = \int_D uv \, d\mathcal{D}, \quad \mathcal{D} = \mathcal{C}, \mathcal{S}, \text{ or } \mathcal{V}, \quad (4)$$

Projection of “ u
on v ” or “ v on u .”

\mathcal{D} = line or curve \mathcal{C} (one - dimension), surface \mathcal{S} (two dimensions), or volume \mathcal{V} (three - dimensions) (Note : Often v appears conjugated in inner product definitions!)

- (Bi-) linearity of inner product :

$$\langle au_1 + bu_2, v \rangle = a \langle u_1, v \rangle + b \langle u_2, v \rangle,$$

$$\langle u, cv_1 + dv_2 \rangle = c \langle u, v_1 \rangle + d \langle u, v_2 \rangle \quad (5)$$

Inner Products: Examples

- u and v vector functions, $u = \mathbf{u}$ and $v = \mathbf{v}$:

$$\langle u; v \rangle = \int_{\mathcal{D}} u \cdot v \, d\mathcal{D}. \quad (6)$$

- Scalars u and v convolved with scalar Green's function $G(\mathbf{r}, \mathbf{r}')$ ("kernel" of integral equation):

$$\langle u, G, v \rangle \equiv \langle u, \langle G, v \rangle \rangle = \int_{\mathcal{D}} \int_{\mathcal{D}} u(\mathbf{r}) G(\mathbf{r}, \mathbf{r}') v(\mathbf{r}') \, d\mathcal{D}' d\mathcal{D} \quad (7)$$

- Vectors u and v convolved with scalar Green's function $G(\mathbf{r}, \mathbf{r}')$:

$$\langle u; G, v \rangle = \int_{\mathcal{D}} \int_{\mathcal{D}} u(\mathbf{r}) \cdot v(\mathbf{r}') G(\mathbf{r}, \mathbf{r}') \, d\mathcal{D}' d\mathcal{D} \quad (8)$$

- Vectors u and v convolved with dyadic Green's function $\mathcal{G}(\mathbf{r}, \mathbf{r}')$:

$$\langle u; \mathcal{G}; v \rangle = \int_{\mathcal{D}} \int_{\mathcal{D}} u(\mathbf{r}) \cdot \mathcal{G}(\mathbf{r}, \mathbf{r}') \cdot v(\mathbf{r}') \, d\mathcal{D}' d\mathcal{D} \quad (9)$$

- In general, $\langle u, v \rangle = \langle v, u \rangle$

Weak and Strong Forms of Operator Equations

- *Strong form* (equality holds at every point in \mathcal{D}):

$$\mathcal{L}u = f \quad (10)$$

- *Weak form* (equality holds in a weighted average sense):

$$\langle w, Lu \rangle = \langle w, f \rangle \quad (11)$$

where $\{w\}$ is a set of *weighting functions*.

Notes:

- For differential operators, integration by parts is often used to transfer differentiability requirements from u to w .
 - If u is continuous, solutions u of (10) and (11) are identical
 - If u is infinite dimensional, then so the set of functions w
- In numerical solutions, w 's are chosen from a *finite* set of weighting or testing functions $\{w_m^{(N)}\}$, $m = 1, 2, \dots, N$.

Bases and Unknown Representations

- Approximate u as

$$u \approx u = \sum_{n=1}^N U_n u_n = [U_n]^t [u_n] \quad (12)$$

where coefficients U_n are unknown and u_n , $n = 1, \dots, N$ are known *basis functions*.

- u_n must be "independent" and capable of approximating u .
- Independence of bases is measured by their projections on one another,

$$\langle u_m, u_n \rangle \equiv \text{Gram Matrix} \quad (13)$$

Independence of Basis Functions

- Ideal are *orthonormal* bases u_n , for which

$$\langle u_m, u_n \rangle = \int_{\mathcal{D}} u_m u_n d\mathcal{D} = \delta_{mn} = \begin{cases} 1, m = n, \\ 0, m \neq n, \end{cases} \quad (14)$$

(δ_{mn} is the "Kronecker delta") but are difficult to discover for arbitrary \mathcal{D}

- Instead, first approximate \mathcal{D} by subdividing into *subdomains* or *elements* (e.g., line segments, triangles, rectangles, tetrahedrons) \mathcal{D}^e , $e = 1, 2, \dots, E$.

Then $\mathcal{D} \approx \tilde{\mathcal{D}} = \bigcup_{e=1}^E \mathcal{D}^e$.

- Then *interpolatory polynomial* bases are usually used. They satisfy the property $u_m(\mathbf{r}_j) = \delta_{mj}$ where \mathbf{r}_j , $j = 1, 2, \dots, N$ are interpolation points on $\tilde{\mathcal{D}}$. In addition, they also satisfy the following "approximation" to (14):

$$\sum_{j=1}^N u_m(\mathbf{r}_j) u_n(\mathbf{r}_j) = \delta_{mn} \quad (15)$$

Method of Moments

- Substituting representation for u into operator equation and testing with $\{w = w_m, m = 1, 2, \dots, N\}$ yields

$$\sum_{n=1}^N \langle w_m, \mathcal{L}u_n \rangle U_n = \langle w_m, f \rangle, \quad m = 1, 2, \dots, N \quad (16)$$

or in matrix form,

$$[L_{mn}][U_n] = [F_m], \quad (17)$$

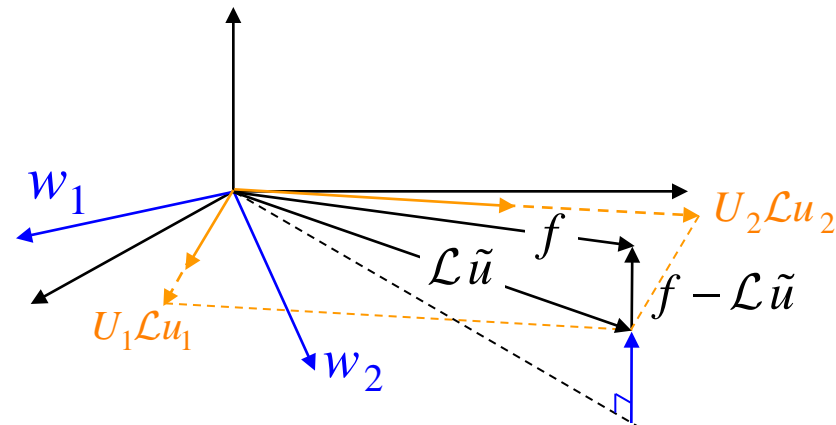
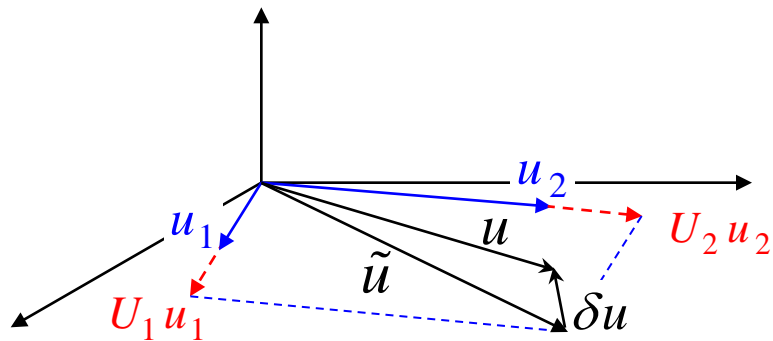
where $L_{mn} = \langle w_m, \mathcal{L}u_n \rangle$ and $F_m = \langle w_m, f \rangle$.

- Solving the linear system yields unknowns $[U_n]$ that provide an approximation to u in \mathcal{D} . The result can also be written as

$$u \approx u = [U_n]^t [u_n] = [u_n]^t [U_n], \quad (18)$$

where $[u_n]^t$ denotes transpose of $[u_n]$.

Abstract Vector Space Interpretation of the Method of Moments



- The unknown is approximated in the "subspace of basis vectors u_n " as $u \approx \tilde{u} = \sum_n U_n u_n$
- Both $\mathcal{L}\tilde{u} = \sum_n U_n \mathcal{L}u_n$ and f are *projected* onto the "subspace of testing vectors w_m "; equating the projections determines $\{U_n\}$.
- The projection both minimizes the residual error $f - \sum_n U_n \mathcal{L}u_n$ and makes it orthogonal to the testing vector subspace.

Linear Functionals

- A *linear functional* $I[u]$ is a scalar physical parameter or figure of merit that depends linearly on u (e.g., $I[au] = aI[u]$).

Examples :

- Capacitance where u is surface charge
- Input admittance where u is a surface current
- Vector component of far field where u is a surface current
- Value of $u(\mathbf{r})$ at point \mathbf{r} (may be unbounded at edge or corner!)

- **Riesz representation theorem :** For any *bounded* linear functional, a function g exists such that $I[u]$ can be represented as an inner product,

$$I[u] = - \langle u, g \rangle \quad (19)$$

- For $u \approx \tilde{u}$,

$$I[u] \approx I[\tilde{u}] = - \langle \tilde{u}, g \rangle = - \sum_{n=1}^N U_n \langle u_n, g \rangle = - [U_n]^t [\langle u_n, g \rangle] \quad (20)$$

- **Note :** Sampled values of the unknown involve unbounded functionals :

$$J(\mathbf{r}') = \int_S J(\mathbf{r}) \delta(\mathbf{r} - \mathbf{r}') dS \Rightarrow g(\mathbf{r}) = -\delta(\mathbf{r} - \mathbf{r}') \text{ is unbounded}$$

Summary of Method of Moments

- Subdivide \mathcal{D} into E subdomains or elements \mathcal{D}^e ; approximate the solution domain as $\mathcal{D} \approx \tilde{\mathcal{D}} = \bigcup_{e=1}^E \mathcal{D}^e$.

- Choose (usually interpolating) basis functions $\{u_n\}$ and approximate u as

$$u \approx \tilde{u} = \sum_n U_n u_n = [U_n]^t [u_n].$$

- Choose weighting (testing) functions $\{w_m\}$. (*Galerkin's method*: $\{w_m\} \equiv \{u_m\}$)
- Substitute \tilde{u} into operator equation and test with w_m . (For differential or integro-differential operators, integrate by parts to reduce differentiability requirements on u_n and incorporate boundary conditions.)

Summary of Method of Moments, Cont'd

- Solve the resulting linear matrix system

$$[L_{mn}][U_n] = [F_m]$$

where

$$L_{mn} = \langle w_m, \mathcal{L}u_n \rangle,$$
$$F_m = \langle w_m, f \rangle$$

for unknown coefficients U_n . A direct or iterative solution procedure may be used.

- Compute desired figure - of - merit (functional) $I[u]$ as

$$I[u] \approx I[\tilde{u}] = -\langle \tilde{u}, g \rangle = -\sum_{n=1}^N U_n \langle u_n, g \rangle = -[U_n]^t [\langle u_n, g \rangle]. \quad (21)$$

The Variational Approach

- Variational and MoM approaches appear to be quite different, but really are equivalent, as we'll show.

- As a first step, we define an adjoint operator \mathcal{L}^\dagger such that

$$\langle w, \mathcal{L}u \rangle = \langle \mathcal{L}^\dagger w, u \rangle \quad (22)$$

for arbitrary u and w .

- Adjoints exist and are unique; to find:
 - Differential operators: Successively integrate by parts
 - Integral operators: Interchange source and observation points in the kernel
 - Matrix operator: Simply transpose the original matrix

The Adjoint Problem

- The variational approach to solving $\mathcal{L}u = f$ begins by considering the linear functional

$$I[u] = - \langle u, g \rangle$$

- Next define the *adjoint* problem,

$$\mathcal{L}^\dagger w = g. \tag{23}$$

where g plays role of source or forcing function, w is solution of adjoint problem.

- Physical significance of w may not always be clear, but note it does provide an alternative means to compute the functional:

$$\begin{aligned} I[u] &= - \langle u, g \rangle = - \langle u, \mathcal{L}^\dagger w \rangle \\ &= - \langle \mathcal{L}u, w \rangle = - \langle f, w \rangle \end{aligned} \tag{24}$$

In electromagnetics, this dual representation is usually a consequence of *reciprocity*, which also often implies that $\mathcal{L} = \mathcal{L}^\dagger$ (\mathcal{L} is *self - adjoint*)

Adjoint Operator Examples

- $\mathcal{L}^\dagger V = \frac{d^2 V}{dx^2} + k_0^2 V \quad + \quad \text{BC}^\dagger \text{s}$
- $\mathcal{L}^\dagger q_s = \int_s \frac{q_s(\mathbf{r}')}{4\pi\epsilon |\mathbf{r} - \mathbf{r}'|} dS'$
- $\mathcal{L}^\dagger \mathbf{J} = j\omega\mu \int_s G(\mathbf{r}, \mathbf{r}') \mathbf{J}(\mathbf{r}') dS' - \frac{\nabla}{j\omega\epsilon} \int_s G(\mathbf{r}, \mathbf{r}') \nabla' \cdot \mathbf{J}(\mathbf{r}') dS' \quad + \quad \text{BC}^\dagger \text{s}$
- $\mathcal{L}^\dagger \mathbf{J} = \frac{\mathbf{J}(\mathbf{r})}{2} - \hat{\mathbf{n}} \times (\hat{\mathbf{n}} \times \nabla \times \text{PV} \int_s G(\mathbf{r}, \mathbf{r}') \hat{\mathbf{n}}' \times \mathbf{J}(\mathbf{r}') dS') \quad + \quad \text{BC}^\dagger \text{ (see Appendix)}$
- $\mathcal{L}^\dagger \Phi = \nabla^2 \Phi + k_0^2 \epsilon_r \Phi \quad + \quad \text{BC}^\dagger \text{s}$
- $\mathcal{L}^\dagger \mathbf{E} = \nabla \times \mu_r^{-1} \nabla \times \mathbf{E} - k_0^2 \epsilon_r \mathbf{E} \quad + \quad \text{BC}^\dagger \text{s}$
- $\mathcal{L}^\dagger [x_m] = [L_{mn}]^\dagger [x_m]$

Most of the above operators are "self - adjoint!"

Bi-Variational Functional

- Define the *bivariational* functional **

$$I[\tilde{u}, \tilde{w}] = \langle \mathcal{L}\tilde{u}, \tilde{w} \rangle - \langle \tilde{u}, g \rangle - \langle f, \tilde{w} \rangle \quad (25)$$

Note that $I[u, w] = \langle \mathcal{L}u, w \rangle - \langle u, g \rangle - \langle f, w \rangle = I[u]$.

We regard \tilde{u} and \tilde{w} as approximate or *trial* solutions to the original and adjoint problems, respectively.

- Define *solution errors* in the original and adjoint problems as

$$\delta u = \tilde{u} - u, \quad \delta w = \tilde{w} - w. \quad (26)$$

Then we can easily show that

$$I[\tilde{u}, \tilde{w}] = -\langle u, g \rangle + \langle \mathcal{L}\delta u, \delta w \rangle \quad (27)$$

or $\delta I[u, w] = \langle \mathcal{L}\delta u, \delta w \rangle$ with *second order error* in δu and δw .

(Functional is said to be *stationary* or to have only a *second order variation* about the functions u and w .)

** Other, less general functionals may actually restrict the form of the resulting linear system, e.g., to Galerkin's method!

Rayleigh-Ritz Procedure

- Approximate u and w in terms of basis sets $\{u_n\}$ and $\{w_m\}$ as

$$\tilde{u} = \sum_n U_n u_n \quad (28)$$

$$\tilde{w} = \sum_m W_m w_m . \quad (29)$$

- Substitute above expansions into the bi - variational functional,

$$I[\tilde{u}, \tilde{w}] = \sum_m \sum_n W_m U_n \langle \mathcal{L} u_n, w_m \rangle - \sum_n U_n \langle u_n, g \rangle - \sum_m W_m \langle f, w_m \rangle \quad (30)$$

and set $\partial I[\tilde{u}, \tilde{w}] / \partial W_p = \partial I[\tilde{u}, \tilde{w}] / \partial U_p = 0$ (*stationarity condition*).

- Replace dummy index p by m in first set, p by n in the second.
The surprising result is that ...

... One Obtains Independent Moment Equations for Both the Original and the Adjoint Problems!

- (Moment equations for original problem, which are independent of g)

$$\sum_n \langle w_m, \mathcal{L}u_n \rangle U_n = \langle w_m, f \rangle, \quad m = 1, 2, \dots, N \Rightarrow ([L_{mn}][U_n] = [F_m]) \quad (31)$$

- (Moment equations for original problem, which are independent of f)

$$\sum_m \langle \mathcal{L}u_n, w_m \rangle W_m = \langle u_n, g \rangle, \quad n = 1, 2, \dots, N. \quad (32)$$

$$(\text{recall } \langle \mathcal{L}u_n, w_m \rangle = \langle u_n, \mathcal{L}^\dagger w_m \rangle) \Rightarrow ([L_{mn}]^\dagger [W_n] = [G_m]) \quad (33)$$

- Note also the independence of equation sets (31) and (32), and the *reversed roles of basis and testing functions* in the adjoint problem.

Idea : Why not insert the resulting \tilde{u}, \tilde{w} into the *variational* form possibly yielding more accurate results than substituting into the *non - variational* form of the functional?

Evaluation of Functional

Write bivariational functional as

$$\begin{aligned} I(\tilde{u}, \tilde{w}) &= \langle \mathcal{L}\tilde{u}, \tilde{w} \rangle - \langle f, \tilde{w} \rangle - \langle \tilde{u}, g \rangle \\ &= \langle \mathcal{L}\tilde{u} - f, \tilde{w} \rangle - \langle \tilde{u}, g \rangle. \end{aligned} \quad (34)$$

The first term on the right hand side vanishes :

$$\begin{aligned} \langle \mathcal{L}\tilde{u} - f, \tilde{w} \rangle &= \sum_m W_m \left[\sum_n \langle \mathcal{L}u_n, w_m \rangle U_n - \langle f, w_m \rangle \right] = [W_m]^t ([L_{mn}][U_n] - [F_m]) \\ &= 0. \quad (\text{by (31)}) \end{aligned} \quad (35)$$

Hence

$$I(\tilde{u}, \tilde{w}) = -\langle \tilde{u}, g \rangle = -\sum_n U_n \langle u_n, g \rangle \quad (36)$$

i.e.,

$$I(\tilde{u}) = I(\tilde{u}, \tilde{w}), \quad (37)$$

so we obtain the same result using either the (bi-)variational or non - variational forms of the functional!

Equivalence of MoM and Variational Approach

- \tilde{u} can be determined from the MoM equations independent of w and g . The solution is same as that obtained by the (bi-)variational approach.
- $I[\tilde{u}] = I[\tilde{u}, \tilde{w}]$ independent of g .
- The variational approach (and adjoint problem) is useful in proving stationarity, but seems otherwise largely superfluous in arriving at a numerical formulation. The moment method yields the same solution but is generally simpler to apply. The variational problem tells us...
 - Moment method solutions are *automatically stationary*.
 - Error in $I[\tilde{u}, \tilde{w}]$ ($\delta I[u, w] = \langle \mathcal{L}\delta u, \delta w \rangle$) is proportional to that in both u, w ; so to reduce error, choose w_m to *well approximate* w , the solution of the adjoint problem.

Appendix: Derivation of MFIE

Adjoint Operator

- $\mathcal{L}\mathbf{J} = \frac{\mathbf{J}(\mathbf{r})}{2} - \hat{\mathbf{n}} \times \text{PV} \int_S \nabla G(\mathbf{r}, \mathbf{r}') \times \mathbf{J}(\mathbf{r}') dS' = (\hat{\mathbf{n}} \times \mathbf{H}^{\text{inc}})$

We first write the MFIE in the non - standard form

$$-\hat{\mathbf{n}} \times \mathcal{L}\mathbf{J} = -\hat{\mathbf{n}} \times \frac{\mathbf{J}(\mathbf{r})}{2} + \hat{\mathbf{n}} \times (\hat{\mathbf{n}} \times \nabla \times \text{PV} \int_S G(\mathbf{r}, \mathbf{r}') \mathbf{J}(\mathbf{r}') dS') = -\hat{\mathbf{n}} \times (\hat{\mathbf{n}} \times \mathbf{H}^{\text{inc}})$$

$$\hat{\mathbf{n}} \times (\hat{\mathbf{n}} \times \mathbf{H}^{S-}) = -\mathbf{H}_{\text{tan}}^{S-} = -\frac{1}{\mu} \lim_{\mathbf{r} \uparrow S} \nabla \times \mathbf{A}$$

Then for a tangential surface testing vector $\mathbf{M}(\mathbf{r})$, we have that

$$\langle \mathbf{M}, -\hat{\mathbf{n}} \times \frac{\mathbf{J}(\mathbf{r})}{2} + \hat{\mathbf{n}} \times (\hat{\mathbf{n}} \times \nabla \times \text{PV} \int_S G(\mathbf{r}, \mathbf{r}') \mathbf{J}(\mathbf{r}') dS') \rangle$$

which is of the form $\langle \mathbf{M}, -\lim_{\mathbf{r} \uparrow S} \mathbf{H}[\mathbf{J}] \rangle$ and which can be interpreted as a reaction integral.

Hence, by the reaction theorem, we have

$$\langle \mathbf{M}, -\lim_{\mathbf{r} \uparrow S} \mathbf{H}[\mathbf{J}] \rangle = \langle \mathbf{J}, \lim_{\mathbf{r} \downarrow S} \mathbf{E}[\mathbf{M}] \rangle = \langle \mathbf{J}, \hat{\mathbf{n}} \times \frac{\mathbf{M}(\mathbf{r})}{2} + \hat{\mathbf{n}} \times (\hat{\mathbf{n}} \times \nabla \times \text{PV} \int_S G(\mathbf{r}, \mathbf{r}') \mathbf{M}(\mathbf{r}') dS') \rangle$$

$$-\hat{\mathbf{n}} \times (\hat{\mathbf{n}} \times \mathbf{E}^{S+}) = \mathbf{E}_{\text{tan}}^{S+} = -\frac{1}{\epsilon} \lim_{\mathbf{r} \downarrow S} \nabla \times \mathbf{F}$$

Appendix: Derivation of MFIE Adjoint Operator, cont'd

Thus,

$$\langle \mathbf{M}, -\hat{\mathbf{n}} \times \frac{\mathbf{J}(\mathbf{r})}{2} + \hat{\mathbf{n}} \times (\hat{\mathbf{n}} \times \nabla \times \text{PV} \int_S G(\mathbf{r}, \mathbf{r}') \mathbf{J}(\mathbf{r}') dS') \rangle = \langle \mathbf{J}, \hat{\mathbf{n}} \times \frac{\mathbf{M}(\mathbf{r})}{2} + \hat{\mathbf{n}} \times (\hat{\mathbf{n}} \times \nabla \times \text{PV} \int_S G(\mathbf{r}, \mathbf{r}') \mathbf{M}(\mathbf{r}') dS') \rangle$$

and even though $\mathbf{M} \cdot \left(-\hat{\mathbf{n}} \times \frac{\mathbf{J}(\mathbf{r})}{2} \right) = \mathbf{J} \cdot \left(\hat{\mathbf{n}} \times \frac{\mathbf{M}(\mathbf{r})}{2} \right)$, the operator $-\hat{\mathbf{n}} \times \mathcal{L}$ is non-self-adjoint.

For the original operator, set $\mathbf{M} = -\hat{\mathbf{n}} \times \mathbf{w}$ and the above becomes

$$\langle \mathbf{w}, \frac{\mathbf{J}(\mathbf{r})}{2} - (\hat{\mathbf{n}} \times \nabla \times \text{PV} \int_S G(\mathbf{r}, \mathbf{r}') \mathbf{J}(\mathbf{r}') dS') \rangle = \langle \mathbf{J}, \frac{\mathbf{w}(\mathbf{r})}{2} - \hat{\mathbf{n}} \times (\hat{\mathbf{n}} \times \nabla \times \text{PV} \int_S G(\mathbf{r}, \mathbf{r}') \hat{\mathbf{n}}' \times \mathbf{w}(\mathbf{r}) dS') \rangle$$

and the adjoint operator is

$$\mathcal{L}^\dagger \mathbf{w} = \frac{\mathbf{w}(\mathbf{r})}{2} - \hat{\mathbf{n}} \times (\hat{\mathbf{n}} \times \nabla \times \text{PV} \int_S G(\mathbf{r}, \mathbf{r}') \hat{\mathbf{n}}' \times \mathbf{w}(\mathbf{r}) dS')$$