

Variational Formulation

inner product : $(u, v) = \int_{\Omega} uv^* d\Omega$

Hilbert Space : $u, v \in \mathcal{H}$

1) $(u, v) = (v, u)^*$

2) $(a_1 u_1 + a_2 u_2, v) = a_1 (u_1, v) + a_2 (u_2, v)$

3) $(u, u) \geq 0$

4) if $(u, u) = 0$ then $u = 0$

Note: $(au, v) = a(u, v)$
 $(u, av) = a^*(v, u)^* = a^*(u, v)$

$\therefore (au, au) = a(u, au) = aa^*(u, u) = |a|(u, u) > 0$

let L denote an operator over a domain \mathcal{D}
with boundary $\partial\mathcal{D}$

eg: $Lu = f$ on \mathcal{D}

if $L = -\nabla^2$, $f = \frac{\rho}{\epsilon} \Rightarrow -\nabla^2 u = \frac{\rho}{\epsilon}$

on the boundary $\partial\mathcal{D}$ we may have

$u(s) = g(s)$ $s \in \partial\mathcal{D}$ (Dirichlet)

$\frac{\partial u}{\partial n} \Big|_s = p(s)$ $s \in \partial\mathcal{D}$ (Neumann)

$\frac{\partial u}{\partial n} \Big|_s + \sigma(s)u(s) = q(s)$ $s \in \partial\mathcal{D}$ (mixed or Robin)

we seek solutions to the operator equation from the class of functions which are in the domain of L and satisfy the boundary conditions.

Two other properties are required of the operator:

$$\left\{ \begin{array}{l} \text{self-adjoint: } (Lu, v) = (u, Lv) \\ \text{positive definite: } (Lu, u) > 0 \quad \text{if } u \neq 0 \end{array} \right.$$

Now consider the functional:

$$F = (Lu, u) - 2(u, F) \quad (\text{Real Hilbert Space})$$

which maps a function u into a number $\in \mathbb{R}$

assume u_0 is the solution of $Lu_0 = F$

$$\text{then } F = (Lu, u) - 2(u, Lu_0)$$

$$= (Lu, u) - 2(u, Lu_0) + (Lu_0, u_0) - (Lu_0, u_0)$$

$$= (Lu, u) - 2(Lu_0, u) + (Lu_0, u_0) - (Lu_0, u_0)$$

$$\stackrel{\substack{\text{self-adjointness} \\ \text{(by symmetry)}}}{=} (L(u - u_0), u) - (Lu_0, u) + (Lu_0, u_0) - (Lu_0, u_0)$$

$$= (L(u - u_0), u) - (Lu, u_0) + (Lu_0, u_0) - (Lu_0, u_0)$$

$$= (L(u - u_0), (u - u_0)) - (Lu_0, u_0)$$

since L is positive definite $(Lu_0, u_0) > 0$

$$(L(u - u_0), (u - u_0)) > 0$$

$\therefore F$ is minimum at $u = u_0$

\therefore we have the following "minimal Functional theorem":

if a linear operator L is self-adjoint and positive definite under the stated boundary conditions then the Functional

$$F = (Lu, u) - 2(u, f)$$

assumes its minimum value at $u_0 \in \mathcal{D}_L$ (domain of L) iff u_0 is a solution of

$$Lu_0 = f$$

this minimal value is

$$F_{\min} = - (Lu_0, u_0)$$

Note also that if u_0 is a solution, then since L is positive definite it is the unique solution:

say: $Lu_1 = f$ $Lu_2 = f$ are two solutions

since L is linear: $Lu_1 - Lu_2 = L(u_1 - u_2) = Lw = 0$

$$\therefore \langle Lw, w \rangle = 0$$

but since L is positive definite this can only happen if $w \equiv 0 \equiv u_1 - u_2$

\therefore the solution is unique!

example : Poisson's equation.

$$L = -\nabla^2 \quad F = \frac{\rho}{\epsilon} \Rightarrow \boxed{-\nabla^2 u = \frac{\rho}{\epsilon}} \quad \text{on } V$$

$(V)_{\partial S}$

First we must check for self-adjointness and positive definite properties.

$$(Lu, v) = - \int_V v \nabla^2 u \, dV \quad u, v \in \mathcal{D}_L$$

using Green's Identity:

$$\int_S v \frac{\partial u}{\partial n} \, ds = \int_V \nabla u \cdot \nabla v \, dV + \int_V v \nabla^2 u \, dV$$

$$\therefore (Lu, v) = \int_V \nabla u \cdot \nabla v \, dV - \int_S v \frac{\partial u}{\partial n} \, ds$$

similarly :

$$(u, Lv) = \int_V \nabla u \cdot \nabla v \, dV - \int_S u \frac{\partial v}{\partial n} \, ds$$

under homogeneous Dirichlet or Neumann boundary conditions the above surface integrals vanish and we have $(u, Lv) = (Lu, v)$ i.e. self-adjoint.

with "mixed - boundary" conditions
the two surface integrals become

$$\left. \frac{\partial u}{\partial n} \right|_S - \sigma(s) u(s) = 0$$

$$\left. \frac{\partial v}{\partial n} \right|_S + \sigma(s) v(s) = 0$$

$$\left. \frac{\partial u}{\partial n} \right|_S = -\sigma(s) u(s)$$

$$\left. \frac{\partial v}{\partial n} \right|_S = -\sigma(s) v(s)$$

$$\begin{aligned} \therefore - \int_S v \frac{\partial u}{\partial n} ds &= \int_S \sigma(s) u(s) v(s) ds \\ &= - \int_S u \frac{\partial v}{\partial n} ds \end{aligned}$$

\therefore again we get self-adjointness.

For positive definiteness, let $v = u$

$$\therefore (Lu, u) = \int_V (\nabla u)^2 dV - \int_S u \frac{\partial u}{\partial n} ds$$

which is > 0 for $u(s) = 0$ or $\left. \frac{\partial u}{\partial n} \right|_S = 0$

if we have mixed boundary conditions
then $\sigma(s) > 0$ is required for
positive definite property to hold.

$$F = (-\nabla^2 u, u) - 2(u, \frac{p}{\epsilon})$$

$$= \int_V \nabla u \cdot \nabla u dV - \int_S u \frac{\partial u}{\partial n} ds - 2 \int_V \frac{up}{\epsilon} dV$$

which for homogeneous Dirichlet or Neumann boundary conditions becomes:

$$F = \int_V |\nabla u|^2 dV - 2 \int_V \frac{up}{\epsilon} dV \quad (1)$$

for the homogeneous mixed boundary conditions

$$\left. \frac{\partial u}{\partial n} \right|_S = -\sigma(s)u(s)$$

$$\therefore F = \int_V |\nabla u|^2 dV + \int_S \sigma(s)u^2 ds - 2 \int_V \frac{up}{\epsilon} dV \quad (2)$$

Since $\sigma(s) > 0$ $\int_S \sigma(s)u^2 ds > 0$

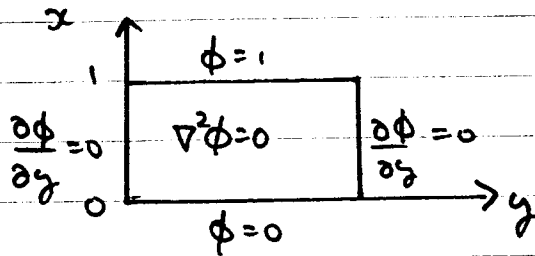
and therefore minimizing (2) is the

same as minimizing (1).

Variational Approach to a Solution

The Rayleigh - Ritz Method

Example:



because of the B.C.'s this problem is independent of y
i.e. it is a 1-D problem:

$$\frac{d^2 \phi}{dx^2} = 0$$

$$\phi(1) = 1 \quad \phi(0) = 0$$

Since we are solving Laplace's equation the appropriate Functional to minimize is:

$$\mathcal{J}(\phi) = \int_0^1 \left(\frac{d\phi}{dx} \right)^2 dx$$

this represents the electrostatic energy in the region.

we choose a "trial" function to minimize the Functional with. The trial function must satisfy Dirichlet boundary conditions:

$$\phi(x) = ax^2 + bx + c \quad (\text{trial function})$$

$$\text{B.C.'s: } \left. \begin{array}{l} \phi(1) = a + b + c \\ \phi(0) = c = 0 \end{array} \right\} \begin{array}{l} c = 0 \\ b = 1 - a \end{array}$$

$$\begin{aligned}\therefore \phi(x) &= ax^2 + (1-a)x \\ &= a(x^2 - x) + x\end{aligned}$$

where a is the "variational parameter" we will use to minimize the Functional:

$$J(a) = \int_0^1 \{a(2x-1) + 1\}^2 dx$$

set:

$$\begin{aligned}0 &= \frac{\partial J(a)}{\partial a} = 2 \int_0^1 (2x-1) [a(2x-1) + 1] dx \\ &= 2 \int_0^1 a(2x-1)^2 + (2x-1) dx \\ &= 2a \int_0^1 (2x-1)^2 dx + 2 \int_0^1 (2x-1) dx \\ &= 2a \times 0 + 0 \\ &\Rightarrow a = 0\end{aligned}$$

$\therefore \phi(x) = x$ which just happens to be the exact answer.

This method of choosing a trial Function with parameters and then choosing the parameters so as to minimize the Functional is known as the Rayleigh-Ritz Method.

One - Dimensional Linear Interpolation Fcns.

Consider the one-dimensional element with labeled "local" node numbers:



the "global" coordinates are x_1 and x_2

we approximate a fcn $F(x)$ over this element by the linear approximation:

$$F(x) = c_1 + c_2 x \quad (1)$$

where c_1 and c_2 are constants to be determined. If we know the value of $F(x)$ at x_1 and x_2 then:

$$F(x_1) \equiv F_1 = c_1 + c_2 x_1$$

$$F(x_2) \equiv F_2 = c_1 + c_2 x_2$$

$$\text{or } \begin{bmatrix} F_1 \\ F_2 \end{bmatrix} = \begin{bmatrix} 1 & x_1 \\ 1 & x_2 \end{bmatrix} \begin{bmatrix} c_1 \\ c_2 \end{bmatrix} \quad (2)$$

$$\left\{ \begin{array}{l} c_1 = \frac{1}{A_e} [F_1 x_2 - F_2 x_1] \\ c_2 = \frac{1}{A_e} [F_2 - F_1] \\ A_e = (x_2 - x_1) \end{array} \right. \quad (3)$$

$$\therefore F(x) = F_1 \left[\frac{x_2 - x}{A_e} \right] + F_2 \left[\frac{x - x_1}{A_e} \right] \quad (4)$$

$$F(x) = F_1 \psi_1(x) + F_2 \psi_2(x) = \sum_{i=1}^2 F_i \psi_i(x) = [F_i]^T [\psi_i] \quad (5)$$

$$= [\psi_i]^T [F_i]$$

$[\alpha_i]$ — represents a column vector

$$\psi_1 = \frac{x_2 - x}{x_2 - x_1} \quad \psi_2 = \frac{x - x_1}{x_2 - x_1} \quad (\text{shape Fcn's}) \quad (6)$$

Properties of linear shape Fcn's:

$$\psi_i(x_j) = \delta_{ij} \quad i, j = 1, 2$$

(i.e. equal to 1 at end-point)

$$\sum_{i=1}^2 \psi_i(x) = 1 \quad (\text{sum produces a constant of 1})$$

$\sum_{i=1}^2 F_i \psi_i(x)$ represents a line through (x_1, F_1) and (x_2, F_2) .

example

For the sample problem

$$\begin{cases} -\frac{d^2 F}{dx^2} = 1 + 4x^2 & 0 \leq x \leq 1 \\ F(0) = F(1) = 0 \end{cases}$$

we minimize the quadratic functional over elements:

$$I(F) = \int_{x_1}^{x_2} \left(\frac{dF}{dx} \right)^2 dx - 2 \int_{x_1}^{x_2} F(1+4x^2) dx$$

$$F^{(e)} = [F_i]^T [\psi_i] = [\psi_i]^T [F_i]$$

$$\frac{dF^{(e)}}{dx} = [F_i]^T [\psi_i'] = [\psi_i']^T [F_i]$$

$$[\psi_i'] = \begin{bmatrix} \frac{1}{x_1 - x_2} \\ \frac{1}{x_2 - x_1} \end{bmatrix}$$

$$\therefore I(F) = [F_i]^T \int_{x_1}^{x_2} [\psi_i][\psi_i]^T dx [F_i] - 2 \int_{x_1}^{x_2} [\psi_i]^T (1+4x^2) dx [F_i]$$

$$= [F_i]^T S [F_i] - 2 [F_i]^T \underline{b}$$

$$S = \int_{x_1}^{x_2} [\psi_i][\psi_i]^T dx = \int_{x_1}^{x_2} \begin{bmatrix} \psi_1' \psi_1' & \psi_1' \psi_2' \\ \psi_2' \psi_1' & \psi_2' \psi_2' \end{bmatrix} dx$$

$$\underline{b} = \int_{x_1}^{x_2} [\psi_i] (1+4x^2) dx = \int_{x_1}^{x_2} \begin{bmatrix} \psi_1 \\ \psi_2 \end{bmatrix} (1+4x^2) dx$$

minimizing the integral: denoting $[F_i] = \underline{F}$

$$\frac{\partial I(\underline{F})}{\partial \underline{F}} = \frac{\partial}{\partial \underline{F}} \left[\underline{F}^T S \underline{F} - 2 \underline{F}^T \underline{b} \right]$$

$$= S \underline{F} + \left[\frac{\partial (S \underline{F})^T}{\partial \underline{F}} \right] \underline{F} - 2 \frac{\partial \underline{F}^T \underline{b}}{\partial \underline{F}}$$

$$= S \underline{F} + \frac{\partial \underline{F}^T}{\partial \underline{F}} S^T \underline{F} - 2 \underline{b}$$

$$= 2 S \underline{F} - 2 \underline{b} = 0 \quad (\text{since } S \text{ is symmetric})$$

\therefore we have the matrix equation

$$S \underline{F} = \underline{b} \quad \text{for every element}$$

$$S = \int_{x_1}^{x_2} \begin{bmatrix} \frac{1}{(x_1-x_1)^2} & \frac{-1}{(x_2-x_1)} \\ \frac{-1}{(x_2-x_1)} & \frac{1}{(x_1-x_2)^2} \end{bmatrix} dx = \frac{1}{x_2-x_1} \begin{bmatrix} 1 & -1 \\ -1 & 1 \end{bmatrix}$$

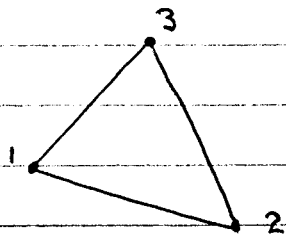
$$\underline{b} = \int_{x_1}^{x_2} \begin{bmatrix} \frac{x_2-x}{x_2-x_1} \\ \frac{x-x_1}{x_2-x_1} \end{bmatrix} (1+4x^2) dx = \frac{1}{(x_2-x_1)} \begin{bmatrix} -\frac{1}{2}x^2 + \frac{4}{3}x_2x^3 - x^4 \\ \frac{1}{2}x^2 - \frac{4}{3}x_1x^3 + x^4 \end{bmatrix}_{x_1}^{x_2}$$

$$= \begin{bmatrix} -\frac{1}{2}(x_2-x_1) + \frac{4}{3}x_2(x_2-x_1)^2 - (x_2-x_1)^3 \\ \frac{1}{2}(x_2-x_1) - \frac{4}{3}x_1(x_2-x_1)^2 + (x_2-x_1)^3 \end{bmatrix}$$

etc...

Two - Dimensional Interpolation Functions

Consider the triangular element where the "local" node numbers are labeled counterclockwise with "global" coordinates:



$$(x_1, y_1), (x_2, y_2), (x_3, y_3)$$

now say we want to approximate a fcn $F(x, y)$ over the domain of the triangle by the linear fcn:

$$F(x, y) = c_1 + c_2 x + c_3 y \quad (1)$$

where $c_1, c_2,$ and c_3 are constants to be determined.

now say we knew the value of $F(x, y)$ at the three vertices of the triangle, that is:

$$F_1 \equiv F(x_1, y_1) = c_1 + c_2 x_1 + c_3 y_1$$

$$F_2 \equiv F(x_2, y_2) = c_1 + c_2 x_2 + c_3 y_2$$

$$F_3 \equiv F(x_3, y_3) = c_1 + c_2 x_3 + c_3 y_3$$

or

$$\begin{bmatrix} F_1 \\ F_2 \\ F_3 \end{bmatrix} = \begin{bmatrix} 1 & x_1 & y_1 \\ 1 & x_2 & y_2 \\ 1 & x_3 & y_3 \end{bmatrix} \begin{bmatrix} c_1 \\ c_2 \\ c_3 \end{bmatrix} \quad (2)$$

by Cramer's Rule we can solve for the constants c_1, c_2, c_3 :

$$c_1 = \frac{\begin{vmatrix} F_1 & x_1 & y_1 \\ F_2 & x_2 & y_2 \\ F_3 & x_3 & y_3 \end{vmatrix}}{2Ae}$$

$$c_2 = \frac{\begin{vmatrix} 1 & F_1 & y_1 \\ 1 & F_2 & y_2 \\ 1 & F_3 & y_3 \end{vmatrix}}{2Ae}$$

$$c_3 = \frac{\begin{vmatrix} 1 & x_1 & F_1 \\ 1 & x_2 & F_2 \\ 1 & x_3 & F_3 \end{vmatrix}}{2Ae}$$

$$2Ae = \begin{vmatrix} 1 & x_1 & y_1 \\ 1 & x_2 & y_2 \\ 1 & x_3 & y_3 \end{vmatrix}$$

$$c_1 = \frac{1}{2Ae} \left[F_1 (x_2 y_3 - x_3 y_2) + F_2 (x_3 y_1 - x_1 y_3) + F_3 (x_1 y_2 - x_2 y_1) \right]$$

$$c_2 = \frac{1}{2Ae} \left[F_1 (y_2 - y_3) + F_2 (y_3 - y_1) + F_3 (y_1 - y_2) \right] \quad (3)$$

$$c_3 = \frac{1}{2Ae} \left[F_1 (x_3 - x_2) + F_2 (x_1 - x_3) + F_3 (x_2 - x_1) \right]$$

permutation

$$2Ae = (x_2 y_3 - x_3 y_2) + (x_3 y_1 - x_1 y_3) + (x_1 y_2 - x_2 y_1) \quad (4)$$

substituting (3) and (4) into (1):

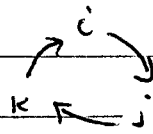
$$\begin{aligned} F(x, y) &= F_1 \psi_1(x, y) + F_2 \psi_2(x, y) + F_3 \psi_3(x, y) \\ &= \sum_{i=1}^3 F_i \psi_i = [F_i]^T [\psi_i] = [\psi_i]^T [F_i] \end{aligned} \quad (5)$$

$[\psi_i]$ - represents a column vector

the ψ_i are called "basis", "shape" or "interpolation" fcn's.

$$\psi_i = \frac{1}{2Ae} (\alpha_i + \beta_i x + \gamma_i y) \quad i=1,2,3 \quad (6)$$

$$\left\{ \begin{array}{l} \alpha_i = (x_j y_k - x_k y_j) \\ \beta_i = (y_j - y_k) \\ \gamma_i = (x_k - x_j) \end{array} \right.$$



"natural" permutation cycle (7)

eg: $\alpha_2 \Rightarrow i=2 \quad j=3 \quad k=1$ in (7)

properties of linear shape fcn's:

$$\psi_i(x_j, y_j) = \delta_{ij} \quad i,j=1,2,3 \quad (8)$$

(i.e. equal to 1 at triangle nodes)

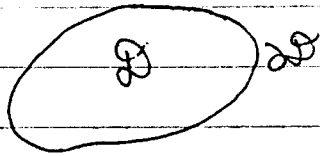
$$\sum_{i=1}^3 \psi_i = 1 \quad (9)$$

(i.e. adding them with equal weight produces a constant 1 over the triangle).

$\sum_{i=1}^3 F_i \psi_i$ represents a plane passing through (x_1, y_1, F_1) , (x_2, y_2, F_2) and (x_3, y_3, F_3) .

Laplace's Equation : $\nabla^2 \phi = 0$ on Ω ①

$\phi(s) = g(s)$ $s \in \partial\Omega$ ②



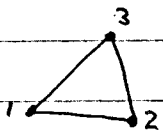
(Dirichlet problem)

solution of ① is equivalent to minimization of quadratic Functional

$$I(\phi) = \frac{1}{2} \int_{\Omega} |\nabla \phi|^2 dV \quad ③$$

we now approximate ϕ on each triangular element as

$$\phi^{(e)}(x,y) = \underline{\phi}^T \underline{\alpha} = \underline{\alpha}^T \underline{\phi} \quad ④$$



$$\underline{\phi} = \begin{bmatrix} \phi_1 \\ \phi_2 \\ \phi_3 \end{bmatrix} \quad \underline{\alpha} = \begin{bmatrix} \alpha_1(x,y) \\ \alpha_2(x,y) \\ \alpha_3(x,y) \end{bmatrix}$$

$\underline{\phi}$ is column vector of vertex values

$\underline{\alpha}$ is column vector of "shape fcn's"

aside: vector differentiation notation

$$\underline{a}^T = [a_1, a_2, \dots, a_n]$$

$$\frac{\partial \underline{a}^T}{\partial a_1} = [1, 0, \dots, 0]$$

$$\frac{\partial \underline{a}^T}{\partial a_2} = [0, 1, \dots, 0]$$

$$\frac{\partial \underline{a}^T}{\partial a_n} = [0, 0, \dots, 1]$$

$$\frac{\partial \underline{a}^T}{\partial \underline{a}} = \underline{I}$$

$$\text{if: } y = \underline{b}^T \underline{c}$$

$$\text{then: } \frac{\partial y}{\partial \underline{a}} = \frac{\partial (\underline{b}^T \underline{c})}{\partial \underline{a}}$$

$$\frac{\partial y}{\partial \underline{a}} = \frac{\partial \underline{b}^T}{\partial \underline{a}} \underline{c} + \frac{\partial \underline{c}^T}{\partial \underline{a}} \underline{b}$$

→ prove this to yourself.

substituting ④ into ③:

$$I(\phi) = \frac{1}{2} \int_{\Omega} \underline{\phi}^T (\nabla \underline{\alpha} \cdot \nabla \underline{\alpha}^T) \underline{\phi} \, dV$$

$$I(\phi) = \frac{1}{2} \underline{\phi}^T \int_{\Omega} (\nabla \underline{\alpha} \cdot \nabla \underline{\alpha}^T) \, dV \, \underline{\phi}$$

$$I(\phi) = \frac{1}{2} \underline{\phi}^T \underline{S} \, \underline{\phi} \quad \underline{S} = \int_{\Omega} (\nabla \underline{\alpha} \cdot \nabla \underline{\alpha}^T) \, dV \quad (5)$$

minimizing $I(\phi)$:

$$0 = \frac{\partial I}{\partial \underline{\phi}} = \frac{1}{2} \left[\frac{\partial \underline{\phi}^T}{\partial \underline{\phi}} (\underline{S} \underline{\phi}) + \frac{\partial (\underline{\phi}^T \underline{S})}{\partial \underline{\phi}} \underline{\phi} \right]$$

$$= \frac{1}{2} \left[\underline{S} \underline{\phi} + \frac{\partial (\underline{\phi}^T \underline{S}^T)}{\partial \underline{\phi}} \underline{\phi} \right]$$

$$= \frac{1}{2} \left[\underline{S} \underline{\phi} + \frac{\partial \underline{\phi}^T \underline{S}^T}{\partial \underline{\phi}} \underline{\phi} \right] = \frac{1}{2} \left[\underline{S} \underline{\phi} + \underline{S}^T \underline{\phi} \right]$$

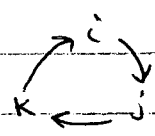
now since \underline{S} is symmetric

$$\frac{\partial I}{\partial \underline{\phi}} = \underline{S} \underline{\phi} = 0 \quad (6)$$

⑥ defines the contribution of each triangle to the final matrix equation.

For first order "shape" funcs $\alpha_i(x, y)$

$$\alpha_i = \frac{1}{2A_e} (a_i + b_i x + c_i y) \quad i = 1, 2, 3$$

$$\begin{cases} a_i = (x_j y_k - x_k y_j) \\ b_i = (y_j - y_k) \\ c_i = (x_k - x_j) \end{cases}$$


$$\begin{aligned} \nabla \alpha_i &= \frac{\partial \alpha_i}{\partial x} \hat{x} + \frac{\partial \alpha_i}{\partial y} \hat{y} \\ &= \frac{1}{2A_e} [b_i \hat{x} + c_i \hat{y}] \end{aligned}$$

$$\nabla \alpha \cdot \nabla \alpha^T = \frac{1}{4A_e^2} \begin{bmatrix} \nabla \alpha_1 \cdot \nabla \alpha_1 & \nabla \alpha_1 \cdot \nabla \alpha_2 & \nabla \alpha_1 \cdot \nabla \alpha_3 \\ \nabla \alpha_2 \cdot \nabla \alpha_1 & \nabla \alpha_2 \cdot \nabla \alpha_2 & \nabla \alpha_2 \cdot \nabla \alpha_3 \\ \nabla \alpha_3 \cdot \nabla \alpha_1 & \nabla \alpha_3 \cdot \nabla \alpha_2 & \nabla \alpha_3 \cdot \nabla \alpha_3 \end{bmatrix}$$

$$= \frac{1}{4A_e^2} \begin{bmatrix} b_1 b_1 + c_1 c_1 & b_1 b_2 + c_1 c_2 & b_1 b_3 + c_1 c_3 \\ b_2 b_1 + c_2 c_1 & b_2 b_2 + c_2 c_2 & b_2 b_3 + c_2 c_3 \\ b_3 b_1 + c_3 c_1 & b_3 b_2 + c_3 c_2 & b_3 b_3 + c_3 c_3 \end{bmatrix}$$

$$2A_e = (x_2 y_3 - x_3 y_2) + (x_3 y_1 - x_1 y_3) + (x_1 y_2 - x_2 y_1)$$

$$\therefore S = \int_{\Delta} \nabla \alpha \cdot \nabla \alpha^T dV = \nabla \alpha \cdot \nabla \alpha^T \int_{\Delta} dV = A_e \nabla \alpha \cdot \nabla \alpha^T$$

$$\therefore S = \frac{1}{4A_e} \begin{bmatrix} b_1 b_1 + c_1 c_1 & b_1 b_2 + c_1 c_2 & b_1 b_3 + c_1 c_3 \\ b_2 b_1 + c_2 c_1 & b_2 b_2 + c_2 c_2 & b_2 b_3 + c_2 c_3 \\ b_3 b_1 + c_3 c_1 & b_3 b_2 + c_3 c_2 & b_3 b_3 + c_3 c_3 \end{bmatrix}$$

Eigenvalue Problem:

$$\nabla^2 \phi + k_c^2 \phi = 0 \quad \text{on } \mathcal{D}$$

$$\phi(s) = 0 \quad s \text{ on } \partial \mathcal{D} \quad (\text{TM modes})$$

$$\left. \frac{\partial \phi}{\partial n} \right|_s = 0 \quad s \text{ on } \partial \mathcal{D} \quad (\text{TE modes})$$

$$k_c = \frac{2\pi}{\lambda_c}$$

$\lambda_c = \text{cut-off wavelength}$

Functional is:

$$I(\phi) = \frac{1}{2} \int_{\mathcal{D}} [(\nabla \phi)^2 - k_c^2 \phi^2] dx dy$$

again letting $\phi^{(e)}(x, y) = \underline{\phi}^T \underline{\alpha} = \underline{\alpha}^T \underline{\phi}$ (over a triangle)

$$I(\phi) = \frac{1}{2} \int_{\Delta} [\underline{\phi}^T (\nabla \underline{\alpha} \cdot \nabla \underline{\alpha}^T) \underline{\phi} - k_c^2 \underline{\phi}^T \underline{\alpha} \underline{\alpha}^T \underline{\phi}] dx dy$$

$$= \underline{\phi}^T \frac{1}{2} \int_{\Delta} \nabla \underline{\alpha} \cdot \nabla \underline{\alpha}^T dx dy - \frac{k_c^2}{2} \underline{\phi}^T \int_{\Delta} \underline{\alpha} \underline{\alpha}^T dx dy \underline{\phi}$$

$$= \frac{1}{2} [\underline{\phi}^T S \underline{\phi} - k_c^2 \underline{\phi}^T T \underline{\phi}]$$

$$S = \int_{\Delta} \nabla \underline{\alpha} \cdot \nabla \underline{\alpha}^T dx dy \quad T = \int_{\Delta} \underline{\alpha} \underline{\alpha}^T dx dy$$

S is the same as for the Laplace equation.

minimizing $\Gamma(\phi)$:

$$\frac{\partial \Gamma}{\partial \underline{\phi}} = 0 = S \underline{\phi} - \frac{\partial}{\partial \underline{\phi}} \left(\frac{\kappa_c^2}{2} \underline{\phi}^T T \underline{\phi} \right)$$

$$= S \underline{\phi} - \frac{\kappa_c^2}{2} \left[\frac{\partial \underline{\phi}^T}{\partial \underline{\phi}} (T \underline{\phi}) + \frac{\partial (T \underline{\phi})^T}{\partial \underline{\phi}} \underline{\phi} \right]$$

$$= S \underline{\phi} - \frac{\kappa_c^2}{2} \left[T \underline{\phi} + \frac{\partial (\underline{\phi}^T T^T)}{\partial \underline{\phi}} \underline{\phi} \right]$$

$$= S \underline{\phi} - \frac{\kappa_c^2}{2} \left[T \underline{\phi} + T^T \underline{\phi} \right]$$

$$0 = S \underline{\phi} - \kappa_c^2 T \underline{\phi} \quad (T \text{ is symmetric})$$

$$\boxed{S \underline{\phi} = \kappa_c^2 T \underline{\phi}} \quad (\text{Matrix eigenvalue equation})$$

we now must evaluate the matrix T :

$$T = \int_{\Delta} \underline{\alpha} \cdot \underline{\alpha}^T dV = \frac{Ae}{12} \begin{bmatrix} 2 & 1 & 1 \\ 1 & 2 & 1 \\ 1 & 1 & 2 \end{bmatrix}$$