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**WINTER COLLEGE ON "MULTILEVEL TECHNIQUES IN
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Algebraic Multigrid Methods

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Algebraic Multigrid
Methods

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This talk covers:

- description of multigrid principles
- AMG approach

- scalar problems
- system problems
- "element" interpolation
- future challenges
- conclusions

Algebraic multigrid (AMG):

- uses multigrid principles
- automatically defines mg components
- solves problems hard for standard mg
 - irregular domains
 - unstructured grids
 - discontinuous coefficients
 - discrete problems
- suited to structural & flow problems
- requires no multigrid expertise

AMG is not a general matrix solver

- applies to classes of problems
- research is application driven
- sense of smoothness
- use available information

Multigrid Principles

Problem: $LU = f$ on Ω

Components required

- A sequence of grids $\Omega^1, \Omega^2, \dots, \Omega^m$
 Ω^1 — finest grid
 Ω^m — coarsest grid
- Operators L^1, L^2, \dots, L^m
- Grid transfer operators:
 I_{k+1}^k Interpolation
 I_k^{k+1} Restriction
- Relaxation method for each level.

Multigrid cycle (v_1, v_2) V-cycle $MG^k(u^k; f^k)$

If $k=m$, set $u^k = (L^k)^{-1} f^k$

Otherwise

Relax v_1 times on $L^k u^k = f^k$

Set $u^{k+1} = 0$, $f^{k+1} = I_k^{k+1} (f^k - L^k u^k)$

Perform $MG^{k+1}(u^{k+1}; f^{k+1})$

Set $u^k \leftarrow u^k + I_{k+1}^k u^{k+1}$

Relax v_2 times on $L^k u^k = f^k$

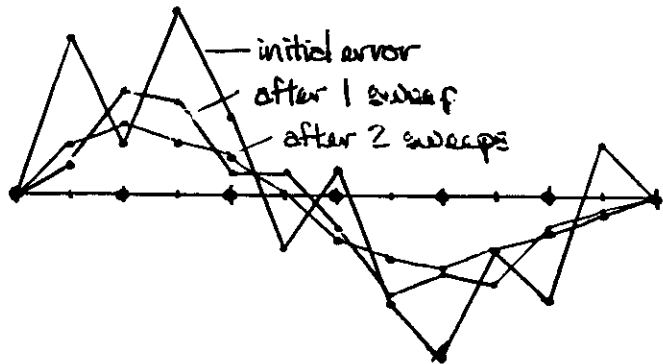
} Coarse-grid correction

The Cycle - why it works -

$$\text{let } L = \frac{d^2}{dx^2}$$

Solve $L^k U^k = F^k$ on $[0, 1]$ $U(0) = U(1) = 0$

$$e^k = U^k - u^k$$



Relaxation smooths error - quickly (converges slowly on smooth error)

$$e^k = U^k - u^k \quad \text{and} \quad r^k = F^k - L^k u^k \quad \text{smooth}$$

Residual equation

$$L^k e^k = r^k$$

Transfer to Ω^{kn}

$$L^{kn} U^{kn} = I_k^{kn} r^k$$

U^{kn} approximates e^k , so

$$u^k + I_k^{kn} U^{kn} \text{ approximates } U^k.$$

Reasons for efficiency:

- Relaxation smooths error
 \Rightarrow residual equation can be solved on a coarser grid.

General Conditions for an efficient MG solution -

- (1) Error not reduced by relaxation must be approximated by some function in $R(I_k^{kn})$
- (2) Coarse-grid correction must eliminate error in $R(I_k^{kn})$.

Relaxation in "Geometric" multigrid must be chosen carefully -

Algebraic Multigrid -

- Fix relaxation (point Gauss-Seidel)
- Choose coarse grids and interpolation to satisfy (1)
- Define other components to satisfy (2).

Setup & Solution phases separate

Outline of Setup Phase for $L'u = f$ on Ω^k

1) Set $k=1$

2) Partition Ω^k into C^k and F^k . Set $\Omega^{k+1} = C^k$.

Define I_{k+1}^k

Set $I_{k+1}^k = (I_{k+1}^k)^T$
 $L^{k+1} = I_{k+1}^k L^k I_{k+1}^k$ } Variational or Galerkin formulation

3) If Ω^{k+1} is "small", Stop

Otherwise, set $k \leftarrow k+1$ & go to 2.

The Energy Functional and the Goal of Computation

$$Au = f \quad A \text{ s.p.d. } U \text{ exact sol. } e = U - u$$

$$\text{Energy norm } \|u\|_A = \langle Au, u \rangle^{1/2} \quad \langle u, v \rangle = \sum_j u_j v_j$$

- Solving $Au = f \Leftrightarrow$ minimizing $\langle Au - 2f, u \rangle$

$$\langle Au - 2f, u \rangle = \sum_i (\sum_j a_{ij} u_j - 2f_i) u_i$$

$$\text{To minimize } 0 = \frac{\partial}{\partial u_i} \quad i=1, \dots, n$$

$$0 = \sum_j a_{ij} u_j + \sum_j a_{ji} u_j - 2f_i$$

$$\text{or, } A \text{ symmetric} \Rightarrow \sum_j a_{ij} u_j = f_i \Rightarrow Au = f$$

- Minimizing $\langle Au - 2f, u \rangle \Leftrightarrow \min. \|e\|_A$

$$\langle Au - 2f, u \rangle + \langle AU, U \rangle = \langle Au - 2AU, u \rangle + \langle AU, U \rangle$$

$$= \langle Au, u \rangle - 2\langle AU, u \rangle + \langle AU, U \rangle$$

$$= \langle Au, u \rangle - \langle Au, U \rangle - \langle AU, u \rangle + \langle AU, U \rangle$$

$$= -\langle Au, e \rangle + \langle AU, e \rangle$$

$$= \langle Ae, e \rangle = \|e\|_A^2$$

Energy Minimization and the Coarse Grid Problem

Given I_c^f (interpolation)

Goal: Eliminate error in $\mathcal{R}(I_c^f)$ How?

$$\begin{aligned} \min_{v^c} \langle A(u + I_c^f v^c) - 2f, u + I_c^f v^c \rangle \\ = \underbrace{\langle Au - 2f, u \rangle}_{\text{constant}} + \langle Au - 2f, I_c^f v^c \rangle \\ + \underbrace{\langle AI_c^f v^c, u \rangle}_{\downarrow = \langle Au, I_c^f v^c \rangle} + \langle AI_c^f v^c, I_c^f v^c \rangle \end{aligned}$$

$$= c + 2 \langle Au - f, I_c^f v^c \rangle + \langle AI_c^f v^c, I_c^f v^c \rangle$$

$$= c + \langle AI_c^f v^c - 2(f - Au), I_c^f v^c \rangle$$

$$= c + \langle (I_c^f)^T A I_c^f v^c - 2(I_c^f)^T (f - Au), v^c \rangle$$

\Leftrightarrow solve

$$\underbrace{(I_c^f)^T A I_c^f}_{A^c} v^c = \underbrace{(I_c^f)^T}_{I_c^c} (f - Au) \quad (\text{restriction})$$

Note: $\langle Bu, v \rangle = \langle u, B^T v \rangle$

$$\langle u, v \rangle = \langle v, u \rangle$$

$$\langle u, v+w \rangle = \langle u, v \rangle + \langle u, w \rangle$$

Energy Minimization and Gauss-Seidel

Let $d \in \mathbb{R}^n$. Consider the problem

$$\begin{aligned} \min_{\alpha} \langle A(u + \alpha d) - 2f, u + \alpha d \rangle \\ = \langle Au - 2f, u \rangle + \alpha \langle Au - 2f, d \rangle \\ + \alpha \langle Ad, u \rangle + \alpha^2 \langle Ad, d \rangle \\ = c + 2\alpha \langle Au - f, d \rangle + \alpha^2 \langle Ad, d \rangle \end{aligned}$$

To minimize, set $\frac{d}{d\alpha} = 0$

$$0 = \frac{d}{d\alpha} = 2 \langle Au - f, d \rangle + 2\alpha \langle Ad, d \rangle$$

$$\alpha = - \frac{\langle Au - f, d \rangle}{\langle Ad, d \rangle} = \frac{\langle r, d \rangle}{\langle Ad, d \rangle}$$

Let $\delta_i^f = [0, 0, \dots, 1, 0, \dots, 0]^T$ 1 in i th place

- For $d = \delta_i^f$, $i = 1, 2, \dots, n$ This is Gauss-Seidel
- For $d = I_c^f \delta_i^c$, $i = 1, 2, \dots, n_c$.

This is Gauss-Seidel on the coarse grid

⑤

Given L^k - How do we choose Ω^{kh} and I_{kh}^k ?

Assume L^k is "positive type" -

i.e. $l_{ii} > 0$, $l_{ij} \leq 0$ for $i \neq j$ and $\sum_j l_{ij} \geq 0$.
(not necessary, but convenient)

For each $i \in \Omega^k$, define

$$S_i = \{j \in \Omega^k : -l_{ij} \geq \alpha \max_{k \neq i} -l_{ik}\} \text{ "strong" connections}$$

(usually $\alpha = .25$)

Choose C^k, F^k so that:

* For each $i \in F^k$, each $j \in S_i$ is either in C or $S_j \cap C_i (= S_j \cap C) \neq \emptyset$.

2-stage process -

- Quick C/F choice

- 2nd pass - enforce * by introducing C-points.

Choosing the coarse grid.

To Partition Ω into C and F - 2 part process

1. For each $i \in \Omega$, let λ_i be the number of points j which strongly depend on i .
2. Pick i with maximal λ_i (i not already in C or F .) Put i in C .
3. For each j which strongly depends on i (j not already in C or F), put j in F , then increment λ_k for each k on which j strongly depends.
4. If $\Omega = C \cup F$ stop. Otherwise go to 2.

Definition of interpolation:

$$\text{Let } C_i = S_i \cap C$$

$$D_i^S = S_i \cap F$$

$$D_i^W = \text{everything else (i.e. "weak" connections)}$$

Form of interpolation

Let v^{kn} be defined on SL^{kn}

$$(I_{kn}^k v^{kn})_i \begin{cases} v_i^{kn} & \text{if } i \in C \\ \sum_{j \in C_i} \omega_{ij} v_j^{kn} & \text{if } i \in F \end{cases}$$

Relaxation is slow if and only if the residual is small compared to the error.

i.e. if $i \in F$

$$\begin{aligned} L_i e_i &\approx - \sum_{j \neq i} L_{ij} e_j \\ &= - \underbrace{\sum_{j \in C_i} L_{ij} e_j}_{C\text{-points}} - \underbrace{\sum_{j \in D_i^S} L_{ij} e_j}_{\text{Strong}} - \underbrace{\sum_{j \in D_i^W} L_{ij} e_j}_{\text{Weak}} \end{aligned}$$

For weak connections - set $e_j = e_i$

For strong connections, set

$$e_j = \left(\sum_{k \in C_i} L_{jk} e_k \right) / \sum_{k \in C_i} L_{jk}$$

Solve for e_i to get ω_{ij} 's

Results

Laplace Operator

Stencil	Convergence per cycle	Complexity	Times
$\frac{1}{h^2} \begin{bmatrix} & -1 & \\ -1 & 4 & -1 \\ & -1 & \end{bmatrix}$.054	2.21	.29 - Cycle 1.63 - Setup
$\frac{1}{2h^2} \begin{bmatrix} & -1 & -1 \\ & 4 & \\ -1 & & -1 \end{bmatrix}$.067	2.12	.27 - Cycle 1.52 - Setup
$\frac{1}{8h^2} \begin{bmatrix} -1 & -1 & -1 \\ -1 & 8 & -1 \\ -1 & -1 & -1 \end{bmatrix}$.078	1.30	.26 - Cycle 1.83 - Setup
$\frac{1}{20h^2} \begin{bmatrix} & -1 & -1 & \\ -1 & 4 & -1 & \\ & -1 & 4 & -1 \\ -1 & -1 & -1 & \end{bmatrix}$.109	1.30	.26 - Cycle 1.83 - Setup

Results Finite difference - symmetric %/t

$-ε u_{xx} - u_{yy}$	$ε =$	0.001	.084 / cycle
	.01		.093
	.1		.098
	.5		.069
	1		.056
	2		.079
	10		.087
	100		.093
	1000		.083
$-∇(d∇u)$.069

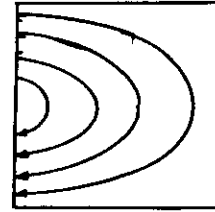
10	100
1	1000

Finite Element - Symmetric

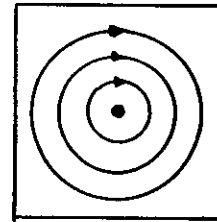
$Δu$ with local refinements	.124
$Δu$ with	.063
BINACAD001Z	.18

Non-symmetric problems (upwind differencing)

$$-ε Δu + a(x,y) u_x + b(x,y) u_y = f$$

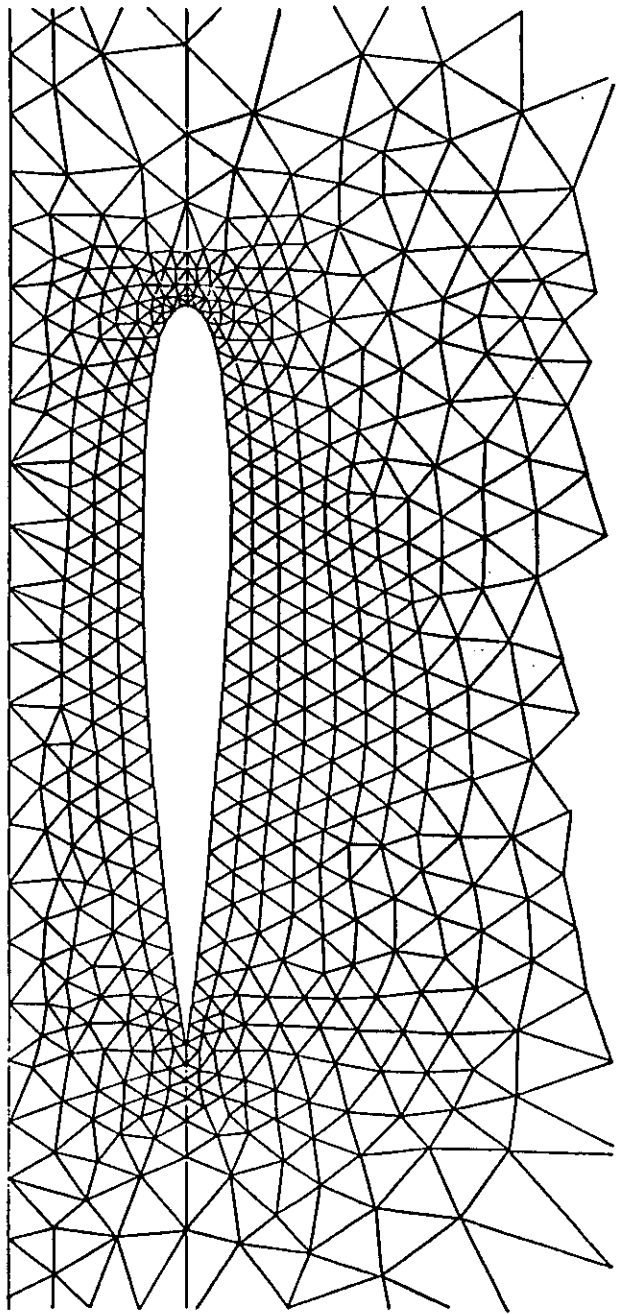


$ε = .1$.069
$ε = .001$.050
$ε = .00001$.029

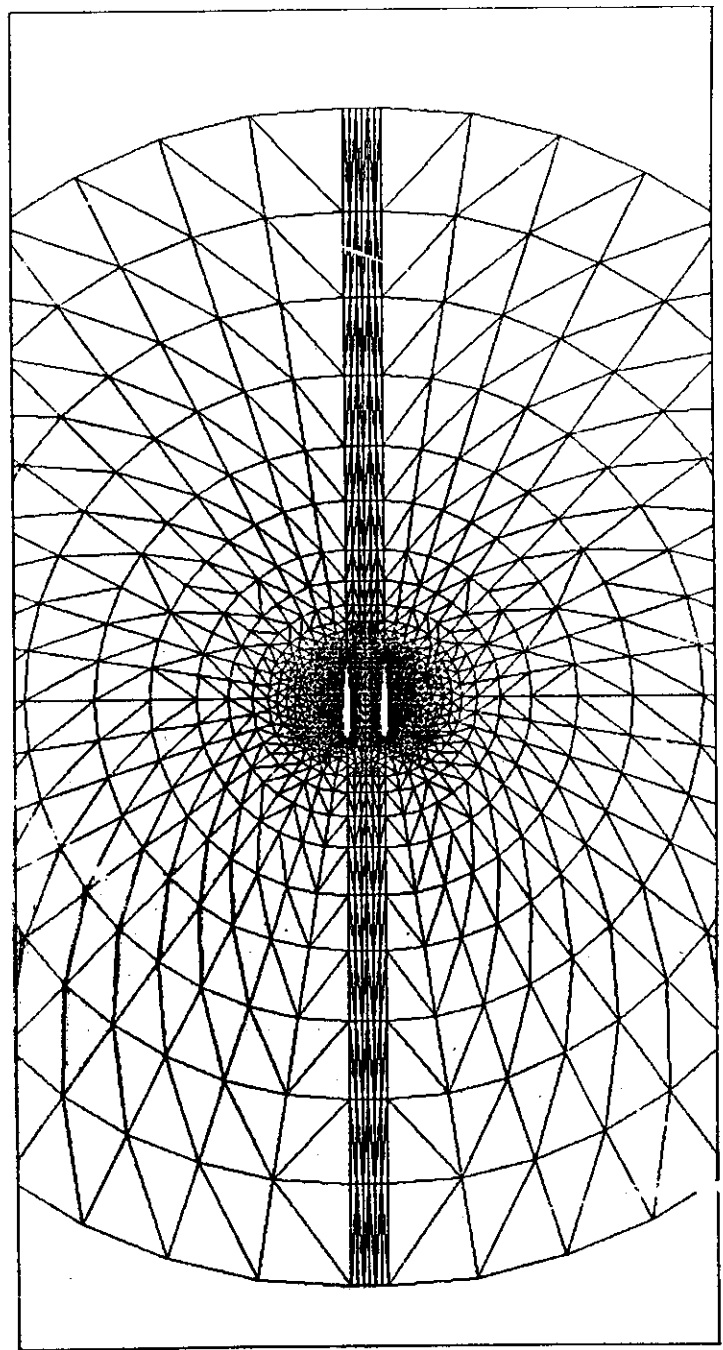


$ε = .1$.062
$ε = .001$.076
$ε = .00001$.102

$a = \cos θ$	$θ = 0$.00002
$b = \sin θ$	$π/8$.0003
	$π/4$.00008
	$3π/8$.0007
	$π/2$.00002



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Problems

Solve Poisson equation on unstructured finite element discretizations of:

- (1) Domain around NACA0012 airfoil
- (2) Domain around 2 NACA0012 airfoils
- (3) Domain between 2 spheres
 $\{\bar{x} \in \mathbb{R}^3: 2 \leq \|\bar{x}\| \leq 10\}$

	Problem			
	1	2	3	
*of points	1020	1702	2052	
grid complexity	1.81	1.74	1.61	
L-complexity	2.67	2.63	3.01	
Convergence Factor	.094	.167	.128	(asymptotic)
Time/cycle*	.10	.16	.34	

* On an IBM 3085.

Last problem - cycle time estimated

(8)

Approaches for "System" Problems (>1 unknown)

Example: Plane-stress

$$u_{xx} + \frac{1-\nu}{2} u_{yy} + \frac{1+\nu}{2} v_{xy} = f \quad \text{on } \Omega \subset \mathbb{R}^2$$

$$\frac{1+\nu}{2} u_{xy} + \frac{1-\nu}{2} v_{xx} + v_{yy} = g$$

yields discrete system

$$\begin{bmatrix} A'_{11} & A'_{12} \\ A'_{21} & A'_{22} \end{bmatrix} \begin{bmatrix} u' \\ v' \end{bmatrix} = \begin{bmatrix} f' \\ g' \end{bmatrix}$$

- AMG as described does not work well. (Relaxation does not make error in u and v "equal".)

"Block" Approach

Block Gauss-Seidel

$$u' \leftarrow (A'_{11})^{-1} (f' - A'_{12} v')$$

$$v' \leftarrow (A'_{22})^{-1} (g' - A'_{21} u')$$

Repeat until convergence

Use "scalar" AMG to "solve" (1 or more cycles)

- separate coarsening $I_{k+1}^{k(1)}$ $I_{k+1}^{k(2)}$
- separate interpolation

⑥

Problem: This does not work. Outer iteration is very slow (v=3 1x2 problem .75)

Solution: Allow interaction between u and v on all levels.

$$A^k = \begin{pmatrix} A_{11}^k & A_{12}^k \\ A_{21}^k & A_{22}^k \end{pmatrix} \quad I_{km}^k = \begin{pmatrix} I_{km}^{k(1)} & 0 \\ 0 & I_{km}^{k(2)} \end{pmatrix}$$

Continue as usual

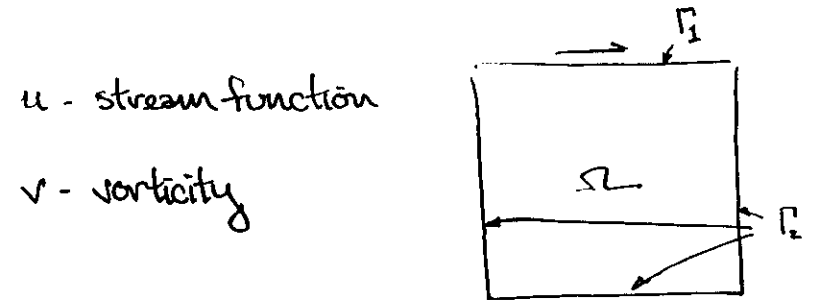
$$I_k^{km} = (I_{km}^k)^T \quad A^{km} = I_k^{km} A^k I_{km}^k \quad k=1,2,\dots$$

This is called the "unknown" approach.



The Model Problem

- Stokes flow in a driven cavity
- vorticity - stream function formulation



$$\begin{aligned} -\Delta u - v &= 0 \\ -\Delta v &= 0 \end{aligned} \quad \text{on } \Omega$$

$$u = 0 \quad \text{on } \Gamma = \Gamma_1 \cup \Gamma_2$$

$$\vec{\nabla} u \cdot \vec{n} = 1 \quad \text{on } \Gamma_1, \quad \vec{\nabla} u \cdot \vec{n} = 0 \quad \text{on } \Gamma_2$$

Discretization of the Problem

- uniform quadrilateral mesh
 $h = 1/8, 1/16, 1/32, 1/64$
- bilinear finite element

Application of AMG

- Separate coarsening for u, v
- Gauss-Seidel relaxation
- $(1,1)$ V-cycles

Results (asymptotic convergence)

$1/8$.22
$1/16$.39
$1/32$.42
$1/64$.44

Approaches for Systems

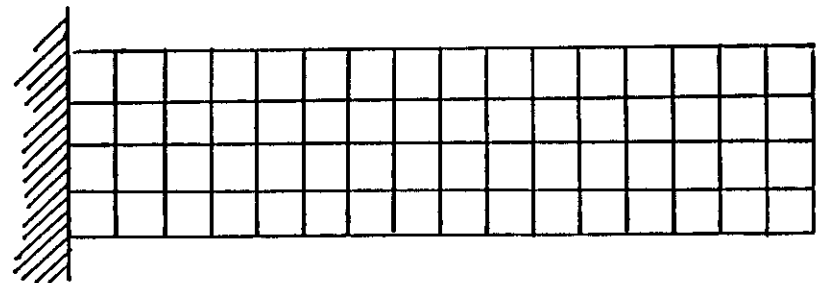
- Elasticity and structures
- Fluid flow problems

"Standard" approach

- Separate interpolation for functions
- Separate coarsening

Works well on

- VLSI design problem
- 2 & 3-D Elasticity (some domains)



Fails on cantilever beam

Why?

- Very ill-conditioned problem
 - Has very slowly converging error
- ⇒ Separate interpolation was not accurate enough.

"Block" approach

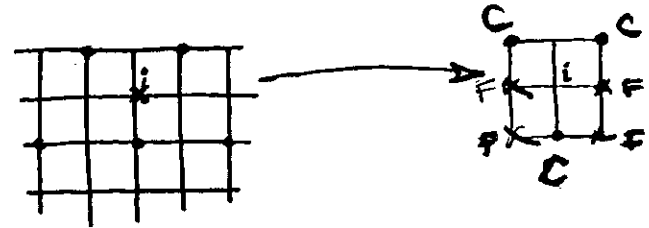
- Blocks correspond to points
- Block Gauss-Seidel
- Point coarsening
- Block interpolation

Problem: Proper distribution difficult

$$A_{ii} \vec{e}_i = - \sum_{k \in \mathcal{C}} A_{ik} \vec{e}_k - \underbrace{\sum_{j \in \mathcal{F}} A_{ij} \vec{e}_j}_{?}$$

Element Interpolation

Consider the cantilever beam discretized using finite elements:



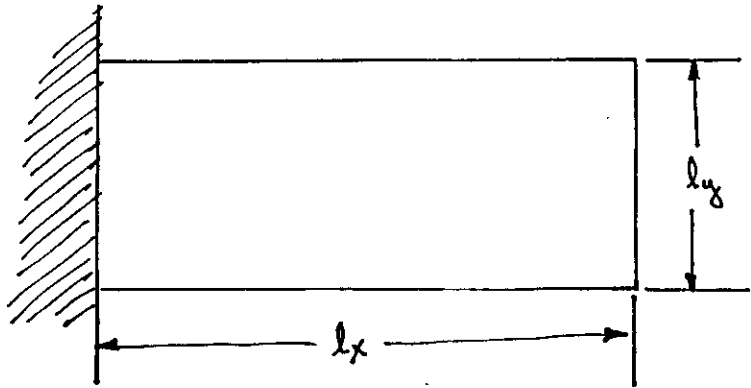
To compute interpolation to point i :

- Consider only elements adjacent to i
- For fixed (but arbitrary) values at C -points, minimize the energy.

2-D Elasticity - Plane Stress

$$u_{xx} + \frac{1-\nu}{2} u_{yy} + \frac{1+\nu}{2} v_{xy} = f_1$$

$$\frac{1+\nu}{2} u_{xy} + \frac{1-\nu}{2} v_{xx} + v_{yy} = f_2$$

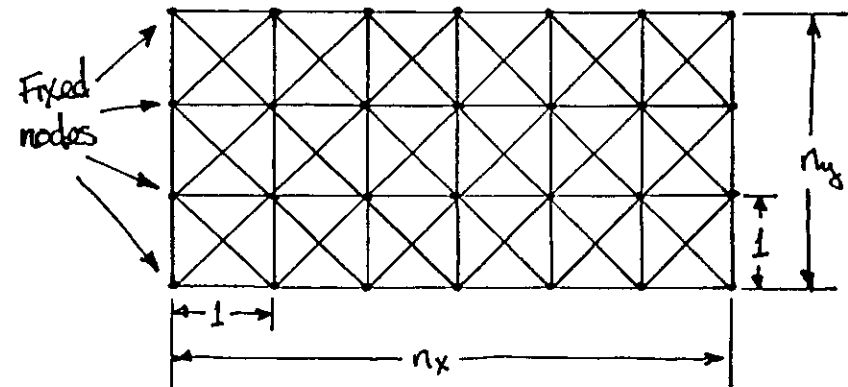


- Finite element discretization (bilinear test functions)

l_x	l_y	h	Rates (V-cycle)	
			element int.	linear int.
1	1	$1/32$.20	.20
1	$1/4$	$1/8$.25	.86
1	$1/8$	$1/16$.26	.96
1	$1/32$	$1/64$.26	.98

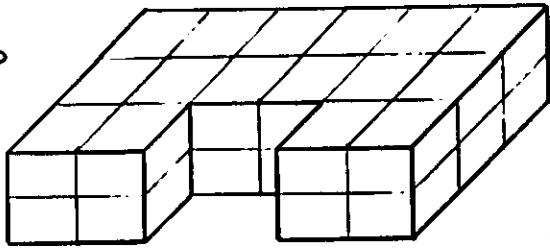
2-D Beam Structures

- Hinged joints
- Beams subject to compression/extension
- Equations derived from Hooke's Law.

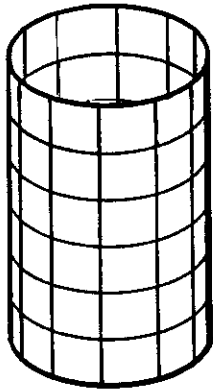


n_x	n_y	V-cycle Rates	
		element int.	linear int.
16	16	.26	.20
32	8	.27	.64
64	4	.28	.97
128	2	.28	.93 +

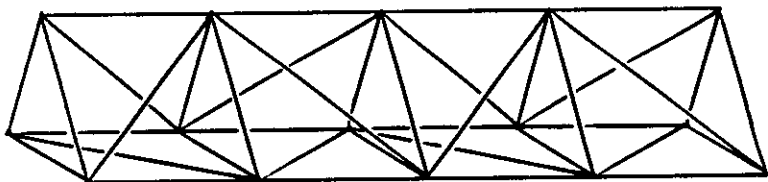
• Solids



• Shells



• Beam Structures

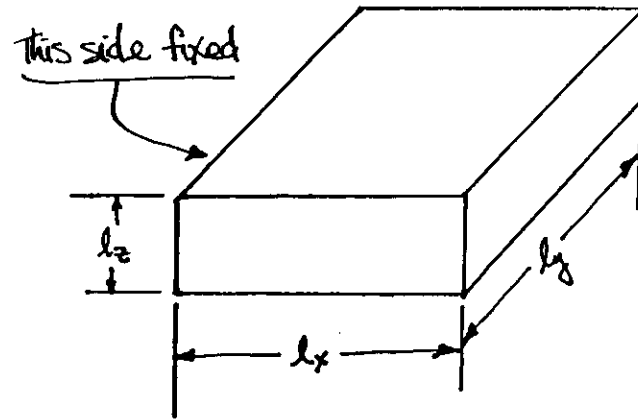


3-D Elasticity - Solids

$$u_{xx} + \frac{1-\nu}{2} (u_{yy} + u_{zz}) + \frac{1+\nu}{2} (v_{xy} + w_{xz}) = f_1$$

$$\frac{1+\nu}{2} u_{xy} + \frac{1-\nu}{2} (v_{xx} + v_{zz}) + v_{yy} + \frac{1+\nu}{2} w_{yz} = f_2$$

$$\frac{1+\nu}{2} (u_{xz} + v_{yz}) + \frac{1-\nu}{2} (w_{xx} + w_{yy}) + w_{zz} = f_3$$

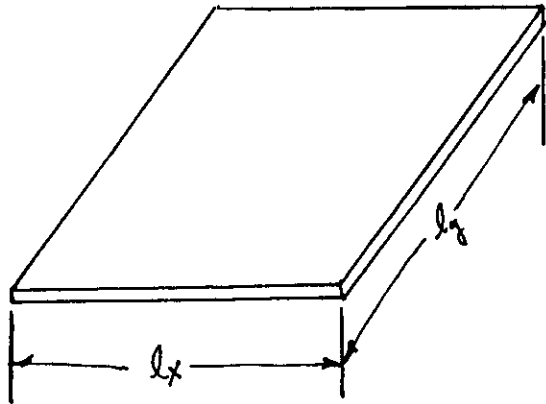


• Finite elements (trilinear test functions)

l_x	l_y	l_z	h	ν -cycle (element int.)
1	1	1	$\frac{1}{8}$.28
1	1	$\frac{1}{4}$	$\frac{1}{4}$.41
1	1	$\frac{1}{8}$	$\frac{1}{8}$.41
1	1	$\frac{1}{16}$	$\frac{1}{16}$.42

3-D Elasticity - Shells

- Equations modified to model physics on thin plates.



- Finite elements - 6 degrees of freedom per node.

Preliminary results (element int.)

$$l_x = l_y = 1 \quad h = 1/16 \quad \nu\text{-cycle rate} = .41$$

Remarks

- Must save A_α 's
 - Extends to coarser grids
- $$A_\alpha^c = (I_c^f)^T A_\alpha I_c^f$$
- Easy to use for systems
 - Can make computation cheaper

$$A^c = (I_c^f)^T A I_c^f$$

