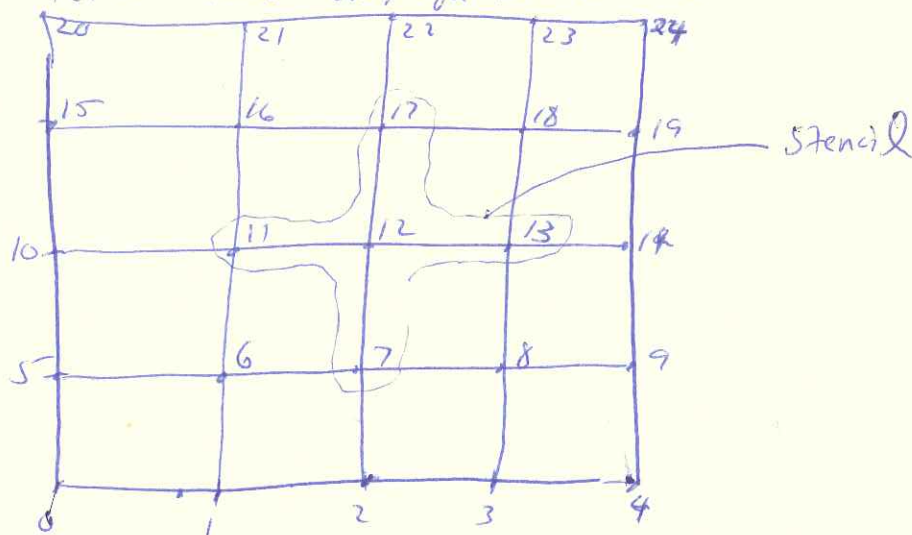


Ok, let's extend our equation to 2D:



$$\frac{\partial^2 T}{\partial x^2} + \frac{\partial^2 T}{\partial y^2} = -\frac{\dot{q}_g}{k}$$

discretization: $\frac{\partial^2 T}{\partial x^2} = \frac{T_{i+1,j} - 2T_{i,j} + T_{i-1,j}}{\Delta x^2}$

$$\frac{\partial^2 T}{\partial y^2} = \frac{T_{i,j+1} - 2T_{i,j} + T_{i,j-1}}{\Delta y^2}$$

i - x index
 j - y index
 } → Note, when doing this on computer, I still suggest using a single dimensional array for T

So: $\frac{T_{i+1,j} - 2T_{i,j} + T_{i-1,j}}{\Delta x^2} + \frac{T_{i,j+1} - 2T_{i,j} + T_{i,j-1}}{\Delta y^2} = -\frac{\dot{q}_g}{k}$

Rearrange:

$$\underbrace{\frac{1}{\Delta y^2} T_{i,j-1}}_{a_s} + \underbrace{\frac{1}{\Delta x^2} T_{i-1,j}}_{a_w} + \underbrace{\left(-\frac{2}{\Delta x^2} - \frac{2}{\Delta y^2}\right) T_{i,j}}_{a_p} + \underbrace{\frac{1}{\Delta x^2} T_{i+1,j}}_{a_E} + \underbrace{\frac{1}{\Delta y^2} T_{i,j+1}}_{a_N} = -\frac{\dot{q}_g}{k}$$

Let's setup the coefficient matrix and source vector for internal pts.
 Assume all ~~boundaries~~ boundaries are temperature boundaries:
 (Dirichlet)

East: T_{EBC}
 Points 5, 10, 15

West: T_{WBC}
 pts 9, 14, 19

South: T_{SBC}
 pts 1, 2, 3

North: T_{NBC}
 pts 21, 22, 23

~~Corners don't~~
 corners can be either
 boundary or average

$$\begin{bmatrix}
 -\frac{a_p}{\Delta x \Delta z} & a_E & 0 & a_W & 0 & 0 & 0 & 0 & 0 \\
 a_W & a_p & a_E & 0 & a_W & 0 & 0 & 0 & 0 \\
 0 & a_W & a_p & 0 & 0 & a_N & 0 & 0 & 0 \\
 a_S & 0 & 0 & a_p & a_E & 0 & a_W & 0 & 0
 \end{bmatrix}
 \begin{bmatrix}
 T_6 \\
 T_7 \\
 T_8 \\
 T_{11} \\
 T_{12} \\
 T_{13} \\
 T_{16} \\
 T_{17} \\
 T_{18}
 \end{bmatrix}
 =
 \begin{bmatrix}
 -\frac{q_0}{K} & -a_W T_{WBC} & -a_S T_{SBC} \\
 -\frac{q_0}{K} & -a_S T_{SBC} & \\
 -\frac{q_0}{K} & -a_S T_{SBC} & -a_E T_{EBC} \\
 -\frac{q_0}{K} & -a_W T_{WBC} &
 \end{bmatrix}$$

keep filling in the matrix / \vec{b} vector

Can solve via LU-Decomposition or Gauss-Seidel or Jacobi, etc.

If ^{uniform mesh} ~~boundary~~ → matrix is symmetric!

↓
 can be useful for
 b/c methods ^{can} take
 advantage of this → Cholesky Decomposition
 $A = UL = LL^T$
 → conjugate gradient
 methods

What if ~~west~~ side is Neumann condition: condition: $\frac{\partial T}{\partial x} = 0$

$$\frac{\partial T}{\partial x} = 0$$

If we use forward difference:

$$\frac{T_{i+1,j} - T_{i,j}}{\Delta x} = 0$$

$$T_{i+1,j} = T_{i,j}$$

Let's look at Pt. 6:

$$\underbrace{\frac{1}{\Delta y^2} T_1}_{a_s} + \underbrace{\frac{1}{\Delta x^2} T_5}_{a_w} + \underbrace{\left(-\frac{2}{\Delta x^2} - \frac{2}{\Delta y^2}\right) T_6}_{a_r} + \underbrace{\frac{1}{\Delta x^2} T_7}_{a_e} + \underbrace{\frac{1}{\Delta y^2} T_{11}}_{a_n} = -\frac{q_s}{k}$$

$$T_5 = T_6$$

$$T_1 = T_{wbc}$$

$$\text{So: } \begin{bmatrix} \underbrace{\left(\frac{1}{\Delta x^2} + a_r\right)}_{a_w} & a_e & 0 & a_n & 0 & 0 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} T_6 \\ T_7 \\ T_8 \\ T_{11} \\ T_{12} \\ \vdots \end{bmatrix} = \begin{bmatrix} -\frac{q_s}{k} - a_s T_{wbc} \end{bmatrix}$$

Recall that the forward difference is 1st order and hence we might not have the ~~best~~ ^{best} accuracy. However, if Δx is relatively small, effect shouldn't be too great.

~~Solving for~~

Once the ~~matrix~~ coefficient matrix and the source vector, \vec{b} , are setup, using the LU-decomposition algorithm to solve is easy \rightarrow it's just a matter of calling those functions.

Note, the $O(N^3)$ includes the factorization.

However, LU-Decomp is expensive! It is $O(N^3)$, this means that the number of calculations performed cubes as a function of the # of unknowns! Also notice that there are a lot of zeros! It's a ~~sparse~~ sparse matrix. LU-Decomposition doesn't care if it's ~~sparse~~! The coefficient matrix is sparse.

Another possible alternative choice for a solution method is Gauss-Seidel or successive over-relaxation (SOR). Of ~~course~~ ^{course}, there are better methods than this as well.

For Gauss-Seidel/SOR:

Rearrange the ~~equation~~ difference equations

$$a_p T_{ij} = -a_E T_{i+1,j} - a_S T_{i,j-1} - a_W T_{i-1,j} - a_N T_{i,j+1} - \frac{\dot{q}_g}{k}$$

$$T_{ij} = \frac{1}{a_p} \left(-a_E T_{i+1,j} - a_S T_{i,j-1} - a_W T_{i-1,j} - a_N T_{i,j+1} - \frac{\dot{q}_g}{k} \right) - \epsilon_g \quad (1)$$

Algorithm for Gauss-Seidel:

loop until ~~set T_i~~ converged or until looped over a set # of times (like 100)
loop over all pts.

if ~~at~~ East Boundary

set T_{ij} to proper BC ϵ_t

else if West Boundary

set T_{ij} to proper BC

else if North

else if South

\rightarrow continue

continue Gauss-Seidel algorithm

else

$T_{ij} = \text{equation above } (Eq. (1))$

end if

if $(|T_{ij} - T_{ij-\text{previous}}| > \text{criterion})$

not converged

end if

$T_{ij-\text{previous}} = T_{ij}$

End loop of pts

End outer convergence loop

→ Go to computer to run my program → show input file, show Tecplot file, open Tecplot, show ~~show inputs to the~~ my function parameters.

General algorithm for HW #2:

Read in input file (heatgen-k, boundary conditions, type and value of solver, etc)

Write out input data to screen to ensure correct.

Set a_p, a_w, a_e, a_s, a_n

Initialize source and Temp arrays (plus any other arrays used)

If LU-Decomp solver

call setloef function → this sets the coefficient matrix and source vector

~~call LU-decomp and LU-solve~~

Call LU-decomp functions to solve

write out Tecplot output file → (NOTE, I'm okay w/ you writing out only the internal pts)

else if Gauss-Seidel

Call Gauss-Seidel function → no need to have full coeff. matrix

Write Tecplot file → you'll probably end up including boundary pts here

* Show input file and Tecplot file *