

# Numerical Modelling in **FORTRAN** day 6

Paul Tackley, 2014

# Today' s Goals

1. Learn about **iterative** solvers for boundary value problems, including the **multigrid** method
2. Practice, practice! This time programming an iterative solver for the **Poisson** equation.
  - Useful for e.g., gravitational potential, electromagnetism, convection (streamfunction-vorticity formulation)

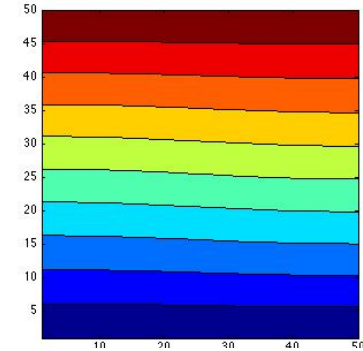
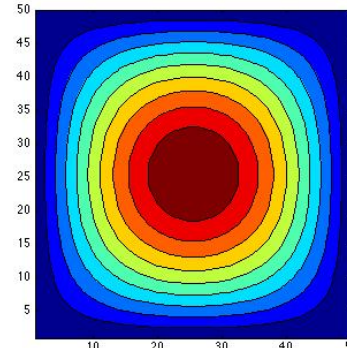
# Review last week: Advection-diffusion for fixed flow field

- (i) Calculate velocity at each point using centered derivatives
- $$(v_x, v_y) = \left( \frac{\partial \psi}{\partial y}, -\frac{\partial \psi}{\partial x} \right)$$
- (ii) Take timesteps to integrate the advection-diffusion equation for the specified length of time using UPWIND finite-differences for  $dT/dx$  and  $dT/dy$

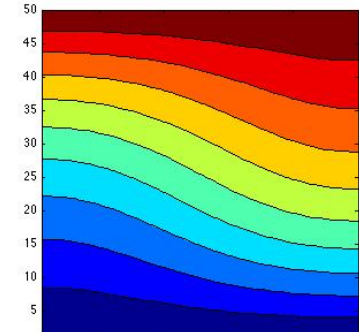
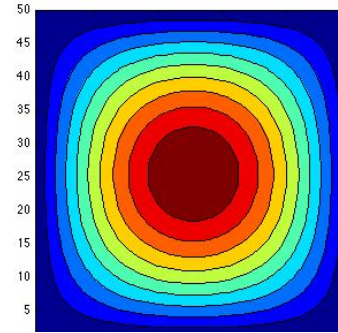
$$\frac{\partial T}{\partial t} = -v_x \frac{\partial T}{\partial x} - v_y \frac{\partial T}{\partial y} + \nabla^2 T$$

This is what it  
should look  
like

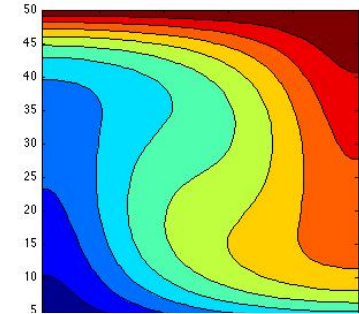
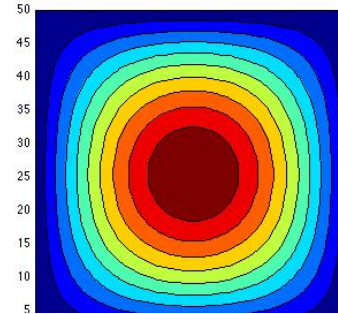
$B=0.1$



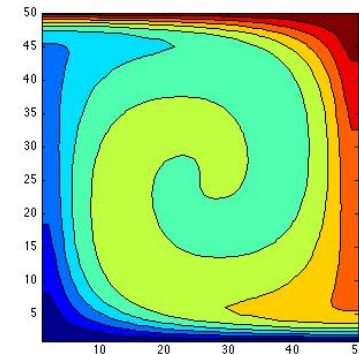
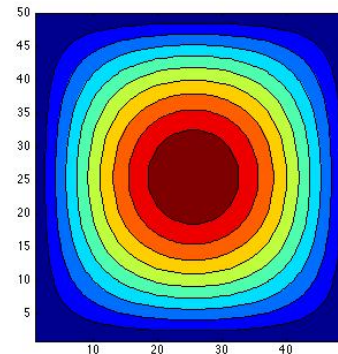
$B=1$



$B=10$



$B=100$



The next step: Calculate velocity field from temperature field (=>**convection**)

e.g., for highly viscous flow (e.g., Earth's mantle) with constant viscosity (P=pressure, Ra=Rayleigh number):

$$-\nabla P + \nabla^2 \vec{v} = -Ra T \hat{y}$$

Substituting the streamfunction for velocity, we get:

$$\nabla^4 \psi = -Ra \frac{\partial T}{\partial x}$$

writing as 2 Poisson equations:

$$\nabla^2 \psi = -\omega \qquad \nabla^2 \omega = Ra \frac{\partial T}{\partial x}$$

the **streamfunction-vorticity** formulation

# we need a Poisson solver

- An example of a **boundary value problem** (uniquely determined by interior equations and values at boundaries), as compared to
- **initial value problems** (depend on initial conditions as well as boundary conditions, like the diffusion equation)

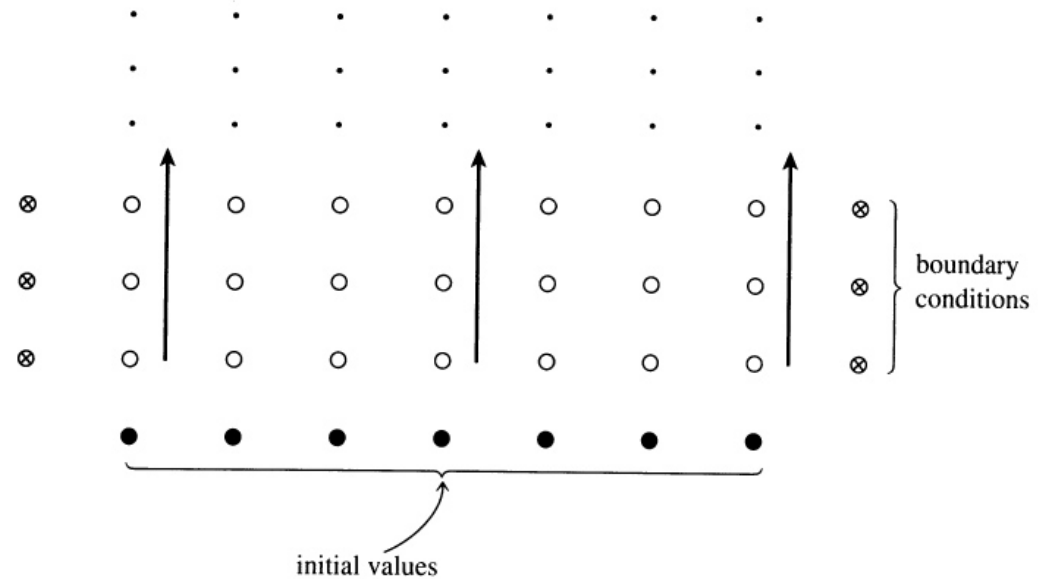
Siméon Denis Poisson (1781-1840)



# Initial value vs boundary value problems

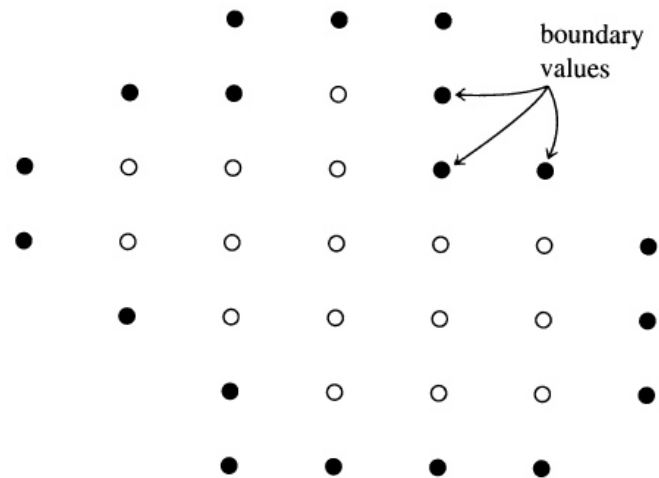
...often, the problem is both

- (a) initial value problem



(a)

- (b) Boundary value problem



(b)

# Example: 1D Poisson

Poisson:  $\nabla^2 u = f$       In 1-D:  $\frac{\partial^2 u}{\partial x^2} = f$

**Solve for**  $\nabla^2$       **fixed**  $f$

Finite-difference form:  $\frac{1}{h^2} (u_{i-1} - 2u_i + u_{i+1}) = f_i$

Example with 5 grid points:

$$u_0 = 0$$

$$u_0 - 2u_1 + u_2 = h^2 f_1$$

$$u_1 - 2u_2 + u_3 = h^2 f_2$$

$$u_2 - 2u_3 + u_4 = h^2 f_3$$

$$u_4 = 0$$

**Problem:**  
simultaneous  
solution needed



# Ways to solve Poisson equation

- **Problem:** A large number of finite-difference equations must be solved simultaneously
- **Method 1. Direct**
  - Put finite-difference equations into a matrix and call a subroutine to find the solution
  - Pro: get the answer in one step
  - Cons: for large problems
    - matrix very large  $(nx*ny)^2$
    - solution very slow:  $time \sim (nx*ny)^3$
- **Method 2. Iterative**
  - Start with initial guess and keep improving it until it is “good enough”
  - Pros: for large problems
    - Minimal memory needed.
    - Fast if use multigrid method:  $time \sim (nx*ny)$
  - Cons: Slow if don't use multigrid method

# Iterative (Relaxation) Methods

- An alternative to using a direct matrix solver for sets of coupled PDEs
- Start with ‘guess’, then iteratively improve it
- Approximate solution ‘relaxes’ to the correct numerical solution
- Stop iterating when the error ( ‘residue’ ) is small enough

# Why?

- Storage:
  - Matrix method has large storage requirements:  $(\text{\#points})^2$ . For large problems, e.g.,  $1\text{e}6$  grid points, this is impossible!
  - Iterative method just uses  $\text{\#points}$
- Time:
  - Matrix method takes a long time for large  $\text{\#points}$ : scaling as  $N^3$  operations
  - The iterative **multigrid** method has  $\text{\#operations}$  scaling as  $N$

# Example: 1D Poisson

Poisson:  $\nabla^2 u = f$       In 1-D:  $\frac{\partial^2 u}{\partial x^2} = f$

Finite-difference form:  $\frac{1}{h^2}(u_{i-1} - 2u_i + u_{i+1}) = f_i$

Assume we have an approximate solution  $\tilde{u}_i$

The error or residue:  $R_i = \frac{1}{h^2}(\tilde{u}_{i-1} - 2\tilde{u}_i + \tilde{u}_{i+1}) - f_i$

Now calculate correction to  $\tilde{u}_i$  to reduce residue

# Correcting $\tilde{u}_i$

From the residue equation note that:  $\frac{\partial R_i}{\partial \tilde{u}_i} = \frac{-2}{h^2}$

So adding a correction  $+\frac{1}{2}h^2 R_i$  to  $\tilde{u}_i$  should zero R

$$\text{i.e.,} \quad \tilde{u}_i^{n+1} = \tilde{u}_i^n + \alpha \frac{1}{2} h^2 R_i$$

Unfortunately it doesn't zero R because the surrounding points also change, but it does reduce R

$\alpha$  is a 'relaxation parameter' of around 1:

$\alpha > 1 \Rightarrow$  'overrelaxation'

$\alpha < 1 \Rightarrow$  'underrelaxation'

# 2 types of iterations: Jacobi & Gauss-Seidel

- Gauss-Seidel: update point at same time as residue calculation
  - do (all points)
  - residue=...
  - $u(i,j) = u(i,j) + f(\text{residue})$
  - end do
- Jacobi: calculate all residues first then update all points
  - do (all points)
  - residue(i,j)=...
  - end do
  - do (all points)
  - $u(i,j) = u(i,j) + f(\text{residue}(i,j))$
  - end do

# Use Gauss-Seidel or Jacobi ?

- Gauss-Seidel converges faster and does not require storage of all points' residue
- $\alpha > 1$  (over-relaxation) can be used for GS, but  $< 1$  required for J to be stable.
- For our multigrid program, optimal  $\alpha$  about 1 for GS, 0.7 for Jacobi
- Only problem: GS not possible on multiple CPUs. Solution: red-black iterations.
- Conclusion: **use Gauss-Seidel iterations**

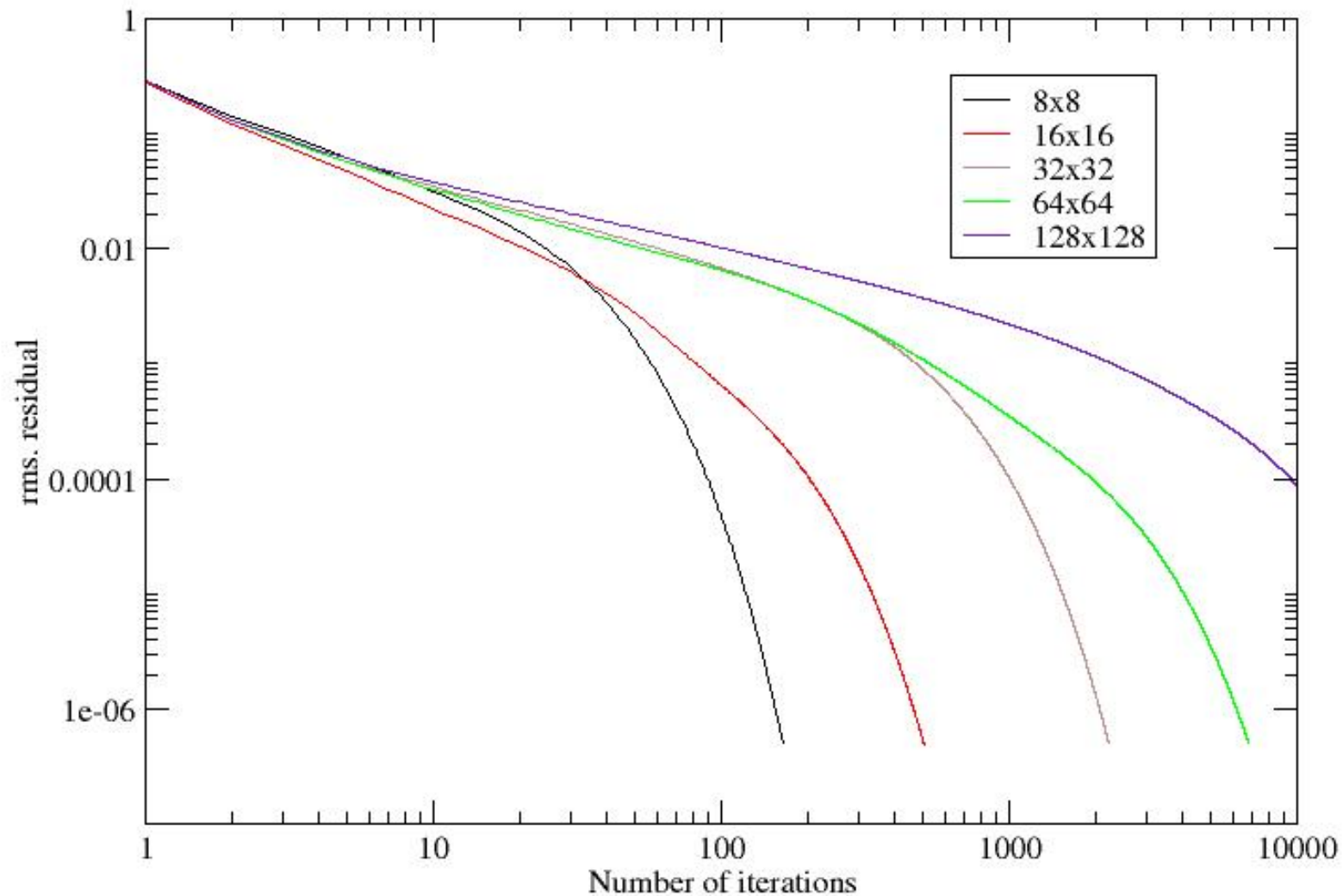
# Demonstration of 1-D relaxation using Matlab:

## Things to note

- The residue becomes smooth as well as smaller (=>short-wavelength solution converges fastest)
- #iterations increases with #grid points. How small must  $R$  be for the solution to be ‘good enough’ (visually)?
- Effect of  $\alpha$  :
  - smaller => slower convergence, smooth residue
  - larger => faster convergence
  - too large => unstable



## Scalar Poisson problem - fine grid iters



- Higher  $N \Rightarrow$  slower convergence

# Now 2D Poisson eqn.

$$\nabla^2 u = f$$

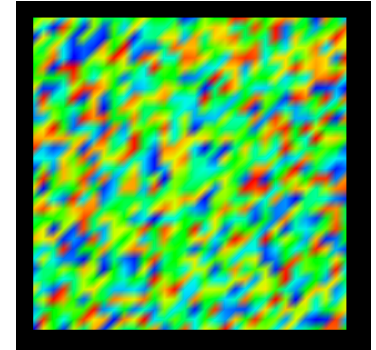
Finite-difference approximation:

$$\frac{1}{h^2} (u_{i,j+1} + u_{i,j-1} + u_{i+1,j} + u_{i-1,j} - 4u_{i,j}) = f_{ij}$$

Use iterative approach=>start with  $u=0$ , sweep through grid updating  $u$  values according to:

$$\tilde{u}_{ij}^{n+1} = \tilde{u}_{ij}^n + \alpha R_{ij} \frac{h^2}{4}$$

Where  $R_{ij}$  is the **residue** (“error”):  $R = \nabla^2 \tilde{u} - f$



# Residue after repeated iterations

Start

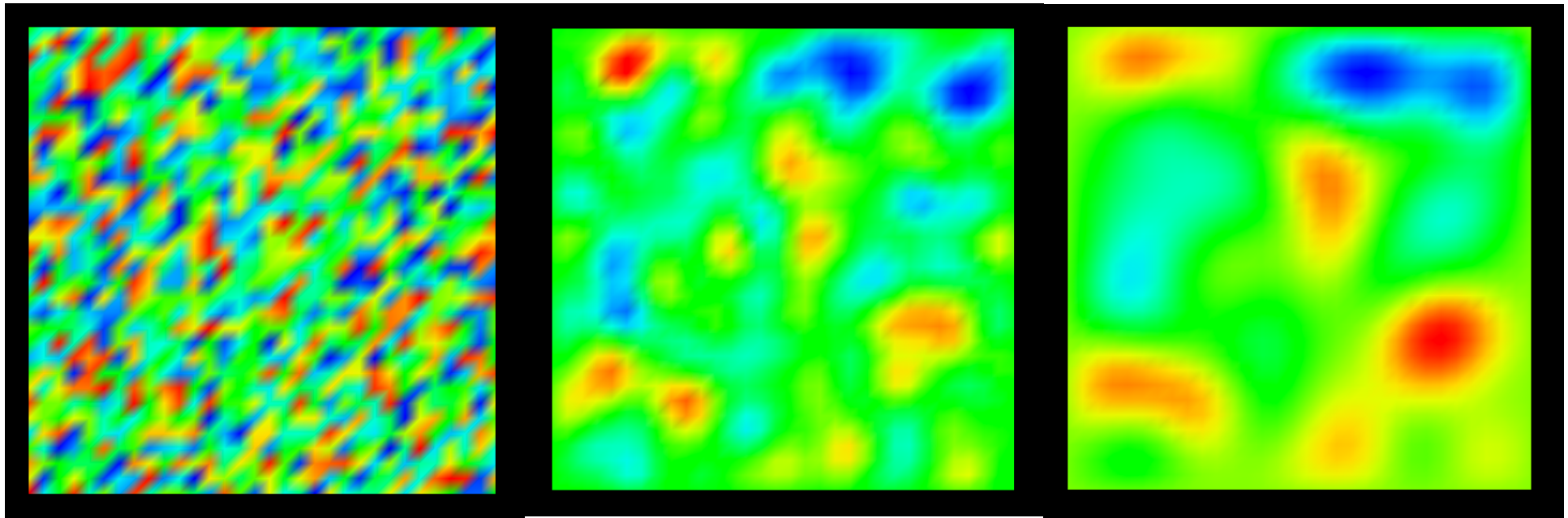
rms residue=0.5

5 iterations

Rms residue=0.06

20 iterations

Rms residue=0.025



Residue gets **smoother** => iterations are like a diffusion process

Iterations smooth the residue  
=>**solve R on a coarser grid**  
=>faster convergence

Start

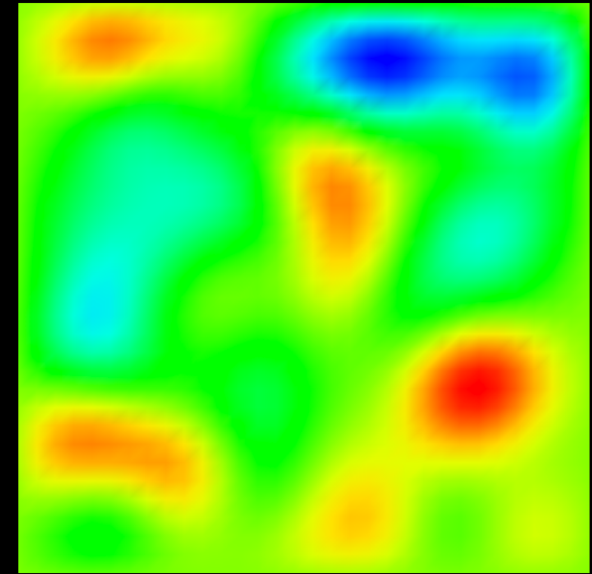
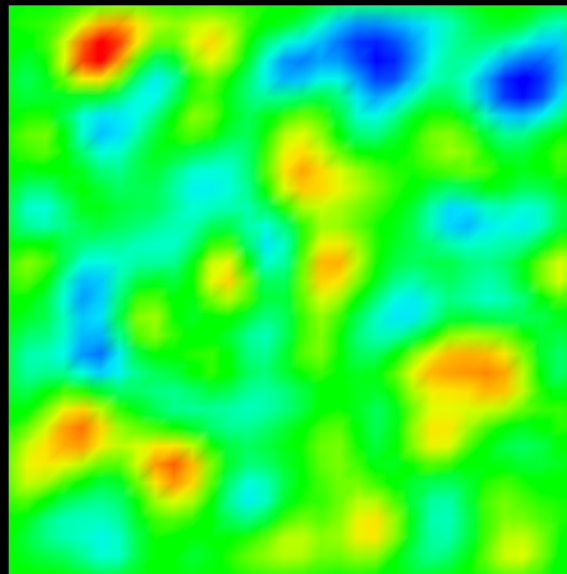
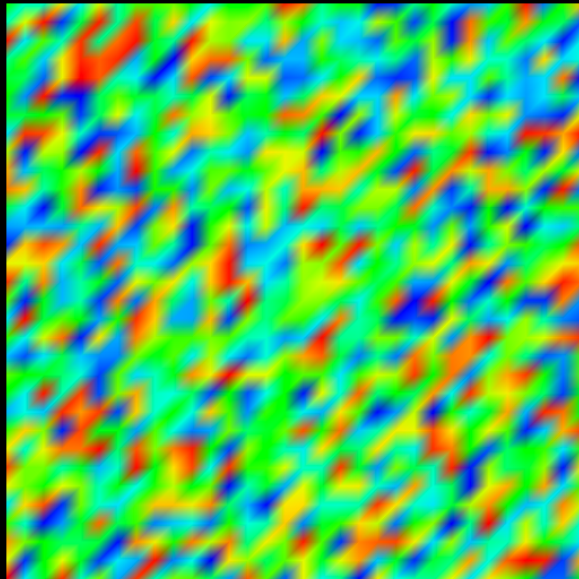
rms residue=0.5

5 iterations

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20 iterations

Rms residue=0.025

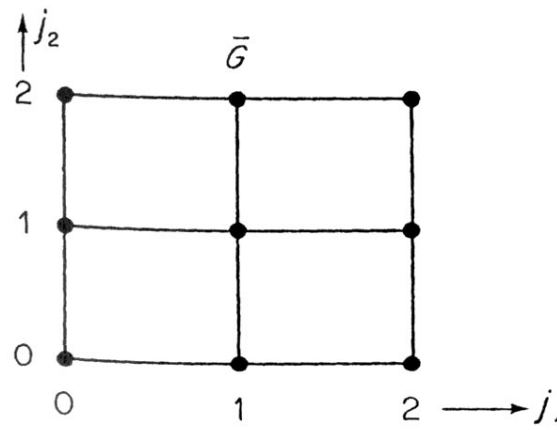
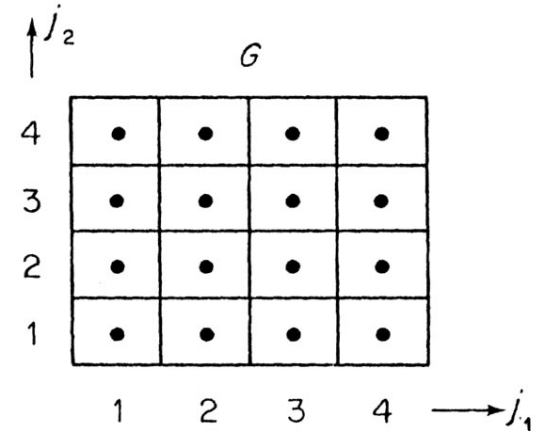
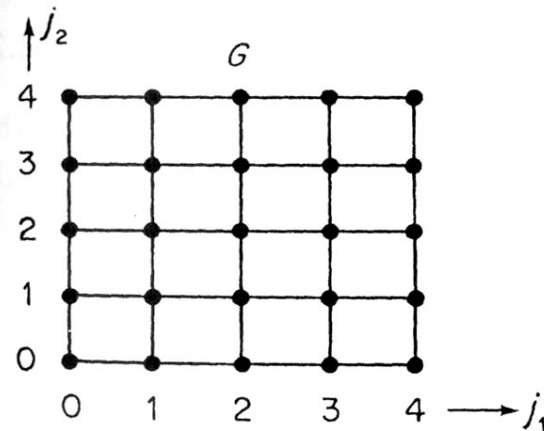


# 2-grid Cycle

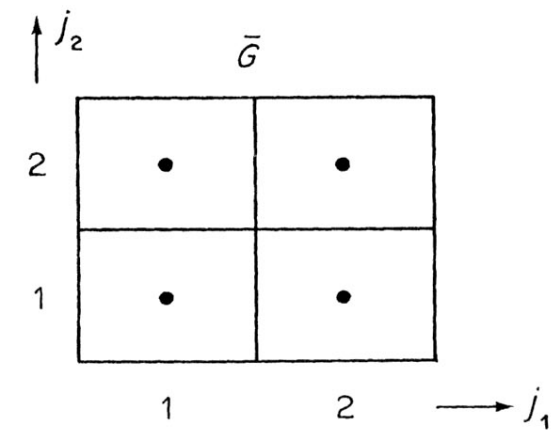
- Several iterations on the fine grid
- Approximate (“**restrict**”)  $R$  on coarse grid
- Find coarse-grid solution to  $R$  (=correction to  $u$ )
- Interpolate (“**prolongate**”) correction=>fine grid and add to  $u$
- Repeat until low enough  $R$  is obtained

# Coarsening: vertex-centered vs. cell-centered

We will use the grid on the left. Number of points has to be a **power-of-two plus 1**, e.g., 5, 9, 17, 33, 65, 129, 257, 513, 1025, 2049, 4097, ...



Vertex-centred



Cell-centred

Figure 5.1.2 Vertex-centred and cell-centred coarsening in two dimensions. (● grid points.)

# Multigrid cycle

- Start as 2-grid cycle, but keep going to coarser and coarser grids, on each one calculating the correction to the residue on the previous level
- Exact solution on coarsest grid (~ few points in each direction)
- Go from coarsest to finest level, at each step interpolating the correction from the next coarsest level and taking a few iterations to smooth this correction
- All lengthscales are relaxed @ the same rate!

# V-cycles and W-cycles

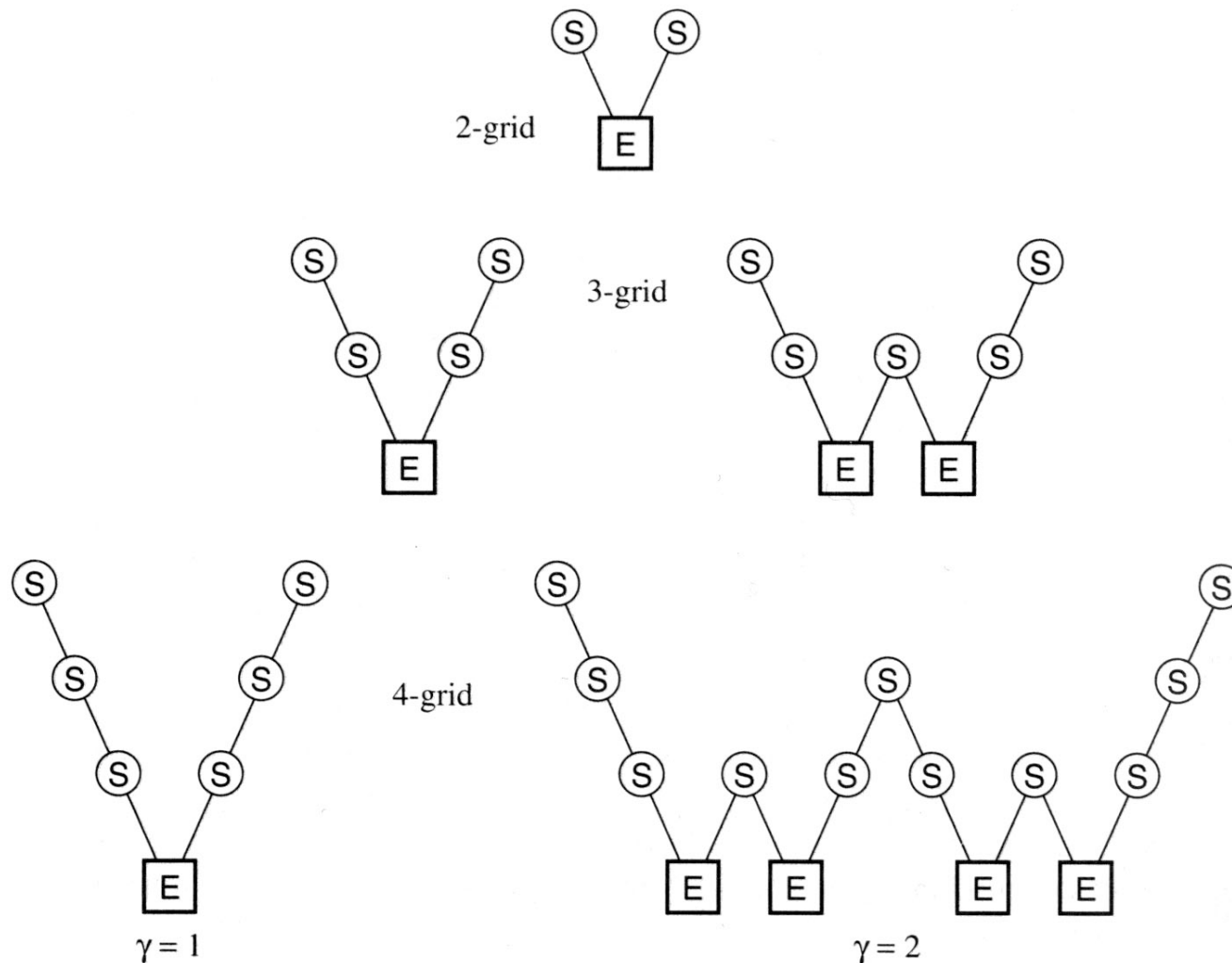
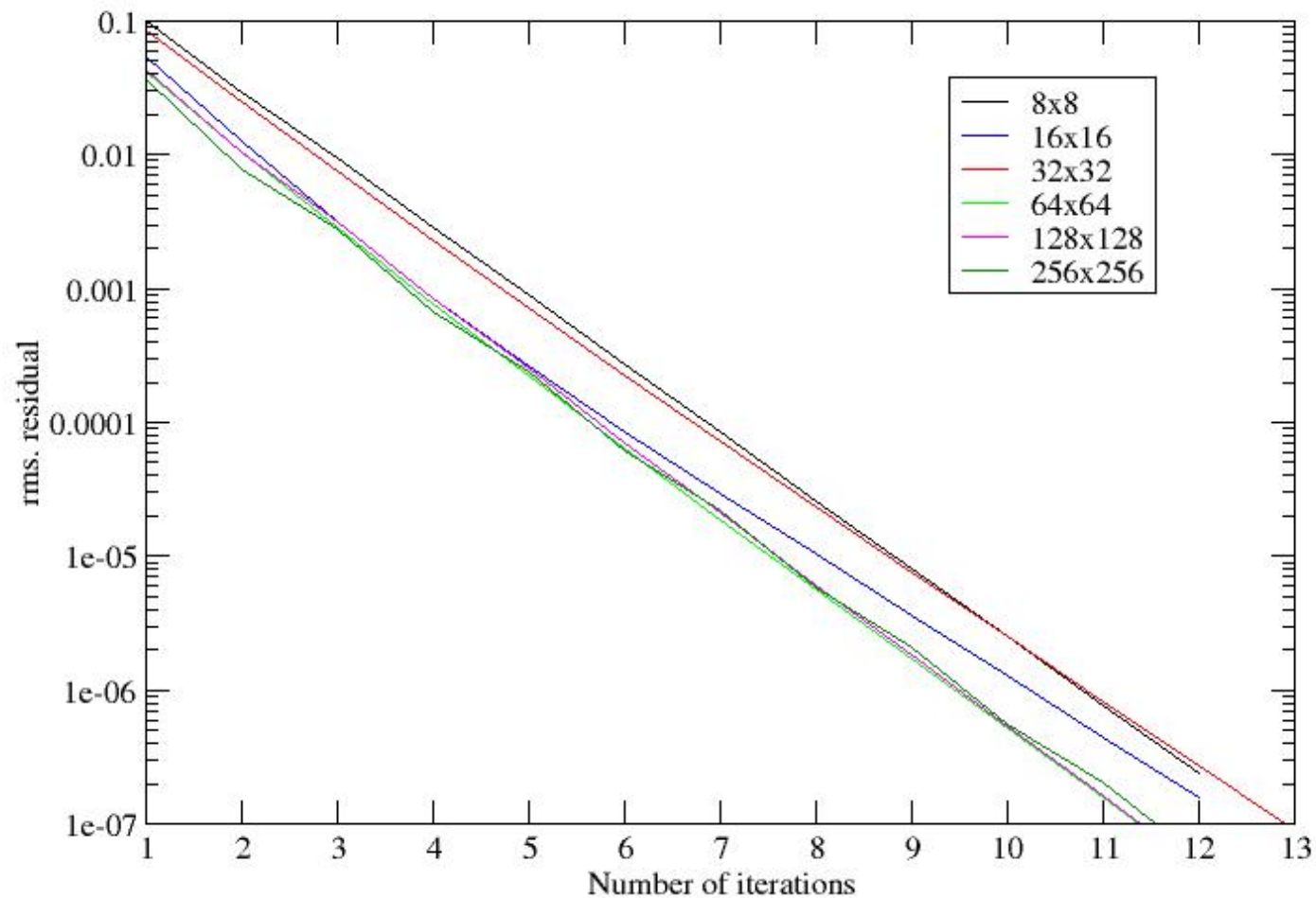


Figure 19.6.1. Structure of multigrid cycles. S denotes smoothing, while E denotes exact solution on the coarsest grid. Each descending line \ denotes restriction ( $\mathcal{R}$ ) and each ascending line / denotes prolongation ( $\mathcal{P}$ ). The finest grid is at the top level of each diagram. For the V-cycles ( $\gamma = 1$ ) the E step is replaced by one 2-grid iteration each time the number of grid levels is increased by one. For the W-cycles ( $\gamma = 2$ ), each E step gets replaced by two 2-grid iterations.



## Scalar Poisson problem - MULTIGRID



- Convergence rate **independent of grid size**
- $\Rightarrow$  #operations scales as #grid points

# Programming multigrid V-cycles

- You are given a function that does the steps in the V-cycle
- The function is **recursive**, i.e., it **calls itself**. It calls itself to find the correction at the next coarser level.
- It calls various functions that you need to write: doing an iteration, calculating the residue, restrict or prolongate a field
- Add these functions and make it into a module
- Boundary conditions: zero

```

recursive function Vcycle_2DPoisson(u_f,rhs,h) result (resV)
  implicit none
  real resV
  real,intent(inout):: u_f(:,,:) ! arguments
  real,intent(in)   :: rhs(:,,:),h
  integer          :: nx,ny,nxc,nyc, i,j ! local variables
  real,allocatable:: res_c(:,:),corr_c(:,:),res_f(:,:),corr_f(:,:)
  real             :: alpha=1.0, res_rms

  nx=size(u_f,1); ny=size(u_f,2) ! must be power of 2 plus 1
  nxc=1+(nx-1)/2; nyc=1+(ny-1)/2 ! coarse grid size

  if (min(nx,ny)>5) then ! not the coarsest level

    allocate(res_f(nx,ny),corr_f(nx,ny), &
             corr_c(nxc,nyc),res_c(nxc,nyc))

    !----- take 2 iterations on the fine grid-----
    res_rms = iteration_2DPoisson(u_f,rhs,h,alpha)
    res_rms = iteration_2DPoisson(u_f,rhs,h,alpha)

    !----- restrict the residue to the coarse grid -----
    call residue_2DPoisson(u_f,rhs,h,res_f)
    call restrict(res_f,res_c)

    !----- solve for the coarse grid correction -----
    corr_c = 0.
    res_rms = Vcycle_2DPoisson(corr_c,res_c,h*2) ! *RECURSIVE CALL*

    !---- prolongate (interpolate) the correction to the fine grid
    call prolongate(corr_c,corr_f)

    !----- correct the fine-grid solution -----
    u_f = u_f - corr_f

    !----- two more smoothing iterations on the fine grid---
    res_rms = iteration_2DPoisson(u_f,rhs,h,alpha)
    res_rms = iteration_2DPoisson(u_f,rhs,h,alpha)

    deallocate(res_f,corr_f,res_c,corr_c)

  else []

```

```

!----- solve for the coarse grid correction -----
corr_c = 0.
res_rms = Vcycle_2DPoisson(corr_c,res_c,h*2) ! *RECURSIVE CALL*

!---- prolongate (interpolate) the correction to the fine grid
call prolongate(corr_c,corr_f)

!----- correct the fine-grid solution -----
u_f = u_f - corr_f

!----- two more smoothing iterations on the fine grid---
res_rms = iteration_2DPoisson(u_f,rhs,h,alpha)
res_rms = iteration_2DPoisson(u_f,rhs,h,alpha)

deallocate(res_f,corr_f,res_c,corr_c)

else

!----- coarsest level (ny=5): iterate to get 'exact' solution
do i = 1,100
    res_rms = iteration_2DPoisson(u_f,rhs,h,alpha)
end do

end if

resV = res_rms ! returns the rms. residue

end function Vcycle_2DPoisson

```

# The use of **result** in functions

- Avoids use of the function name in the code. Instead, another variable name is used to set the result
- Required by some compilers for recursive functions
- Example see this code

```

recursive function Vcycle_2DPoisson(u_f,rhs,h) result (resV)
  implicit none
  real resV
  real,intent(inout):: u_f(:,,:) ! arguments
  real,intent(in)   :: rhs(:,,:),h
  integer          :: nx,ny,nxc,nyc, i,j ! local variables
  real,allocatable:: res_c(:,:),corr_c(:,:),res_f(:,:),corr_f(:,:)
  real            :: alpha=1.0, res_rms

  nx=size(u_f,1); ny=size(u_f,2) ! must be power of 2 plus 1
  nxc=1+(nx-1)/2; nyc=1+(ny-1)/2 ! coarse grid size

  if (min(nx,ny)>5) then ! not the coarsest level

    allocate(res_f(nx,ny),corr_f(nx,ny), &
             corr_c(nxc,nyc),res_c(nxc,nyc))

    !----- take 2 iterations on the fine grid-----
    res_rms = iteration_2DPoisson(u_f,rhs,h,alpha)
    res_rms = iteration_2DPoisson(u_f,rhs,h,alpha)

    !----- restrict the residue to the coarse grid -----
    call residue_2DPoisson(u_f,rhs,h,res_f)
    call restrict(res_f,res_c)

```

```

    !----- coarsest level (ny=5): iterate to get 'exact' solution
    do i = 1,100
      res_rms = iteration_2DPoisson(u_f,rhs,h,alpha)
    end do

  end if

  resV = res_rms ! returns the rms. residue
end function Vcycle_2DPoisson

```

# Exercise

- Make a Poisson solver module by adding necessary functions to the provided V-cycle function
- Write a main program that tests this, taking multiple iterations until the residue is less than about  $1e-5$  of the right-hand side  $f$ 
  - **NOTE**: may need **64-bit** precision to obtain this accuracy
- The program should have the option of calling the iteration function directly, or the V-cycle function, so that you can compare their performance

# Functions to add

- function **iteration\_2DPoisson**(u,f,h,alpha)
  - does one iteration on u field as detailed earlier
  - is a function that returns the rms. residue
- subroutine **residue\_2DPoisson**(u,f,h,res)
  - calculates the residue in array res
- subroutine **restrict**(fine,coarse)
  - Copies every other point in fine into coarse
- subroutine **prolongate**(coarse,fine)
  - Copies coarse into every other point in fine
  - Does linear interpolation to fill the other points



# Convergence criterion

Equation to solve:  $\nabla^2 u = f$

Residue:  $R = \nabla^2 \tilde{u} - f$

Iterate until  $\frac{R_{rms}}{f_{rms}} < err$       Where  $err$  is e.g.  $10^{-5}$

Rms=root-mean-square:  $f_{rms} = \sqrt{\frac{\sum (A_{ij})^2}{n_x n_y}}$

# Test program details

- Using namelist input, read in nx,ny, flags to indicate what type of source and iterations, and alpha
- Initialise:
  - $h=1/(ny-1)$
  - source field f to random, spike, etc.
  - $u(:,:)=0$
- Repeatedly call either iteration\_2DPoisson or Vcycle\_2DPoisson until “converged” (according to rms. residue compared to rms. source field f)
- Write f and u to files for visualisation!

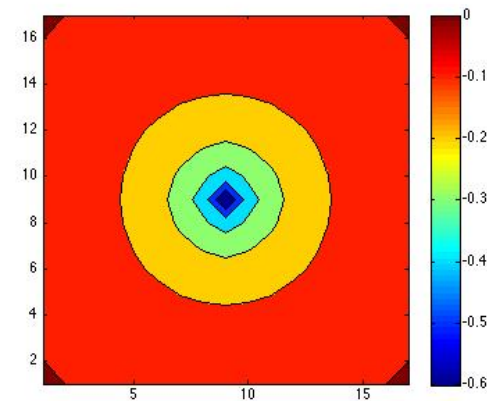
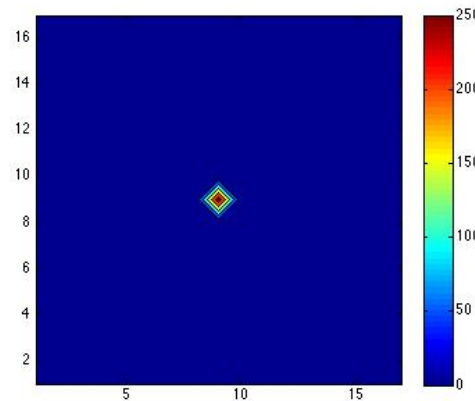
# Tests

- Record number of iterations needed for different grid sizes for each scheme, from  $17 \times 17$  to at least  $257 \times 257$
- What is the effect of alpha on iterations? (for multigrid it is hard-wired)
- For multigrid, what is the maximum number of grid points that fit in your computer, and how long does it take to solve the equations?

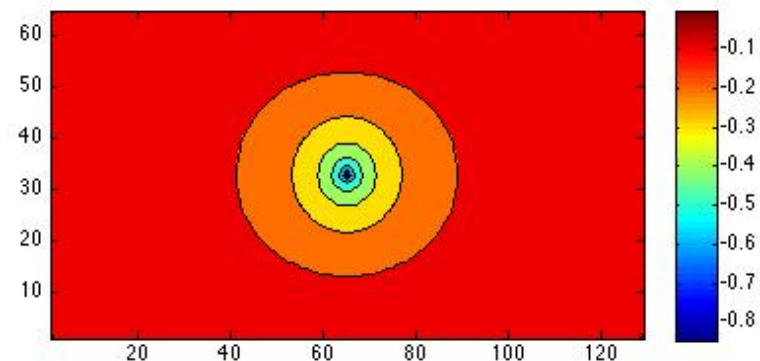
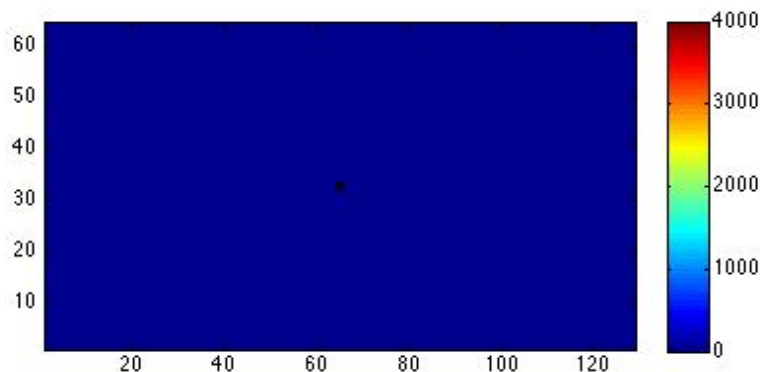
# Example solutions

- For a delta function (spike) in the center of the grid, i.e.,
  - $f(nx/2+1, ny/2+1) = 1/dy^{**2}$ , otherwise 0
  - $(1/dy^{**2})$  is so that the integral of this=1)

17x17 grid



129x65 grid



# Hand in by email

- Your complete program
- Results of your tests
  - #iterations vs. grid size
  - effect of alpha
  - maximum #points)
- Plots of your solutions for a delta-function source