

PHY604 Lecture 16

October 21, 2025

Today's lecture: Elliptical PDEs

- Relaxation methods
- Spectral methods

Elliptical equations: e.g., Laplace equation

- The PDEs we will discuss here represent boundary-value problems
 - Solution is a static field

- Consider Laplace's equation:
$$\frac{\partial^2 \Phi(x, y)}{\partial x^2} + \frac{\partial^2 \Phi(x, y)}{\partial y^2} = 0$$

- Φ is the electrostatic potential
- As usual it is useful to solve a simple problem analytically so that we can benchmark numerical methods

Separation of variables for Laplace's equation

- Write Φ as the product: $\Phi(x, y) = X(x)Y(y)$
- Insert into Laplace's equation and divide by Φ :

$$\frac{1}{X(x)} \frac{d^2 X}{dx^2} + \frac{1}{Y(y)} \frac{d^2 Y}{dy^2} = 0$$

- This equation should hold for all x and y , so each term must be a constant:

$$\frac{1}{X(x)} \frac{d^2 X}{dx^2} = -k^2, \quad \frac{1}{Y(y)} \frac{d^2 Y}{dy^2} = k^2$$

- k is a complex constant
 - Writing constant as k^2 to simplify notation later
 - Signs can be switched
- Now we have two ODEs

Solution of Laplace's eq. ODEs

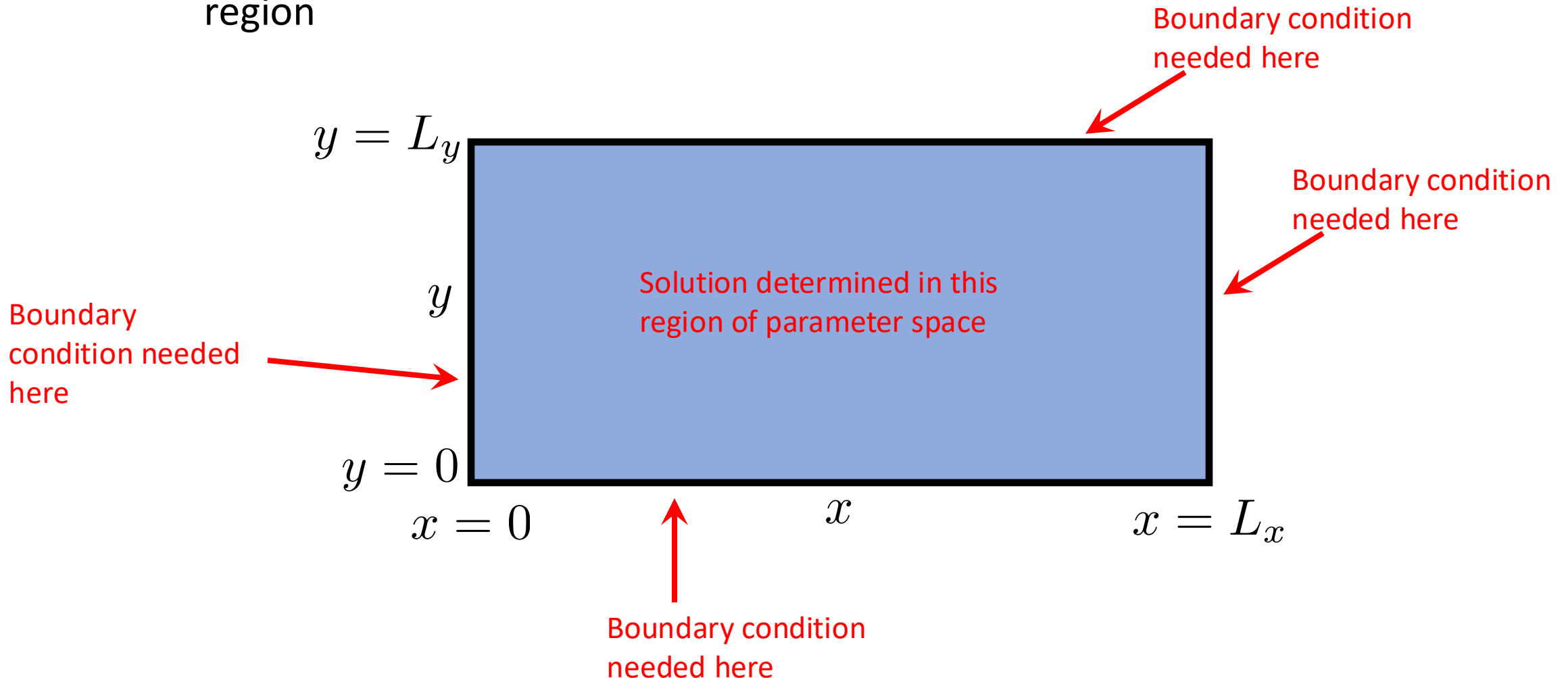
- Solution of these equations are well known:

$$X(x) = C_s \sin(kx) + C_c \cos(kx), \quad Y(y) = C'_s \sinh(ky) + C'_c \cosh(ky)$$

- Recall that k is complex, so solutions are “symmetric”
- To get the coefficients, we need to specify the boundary conditions

Boundary value problems

- All boundary values are specified at the outset
 - E.g., Laplace's equation in electrostatics, potential fixed on for sides of spatial region



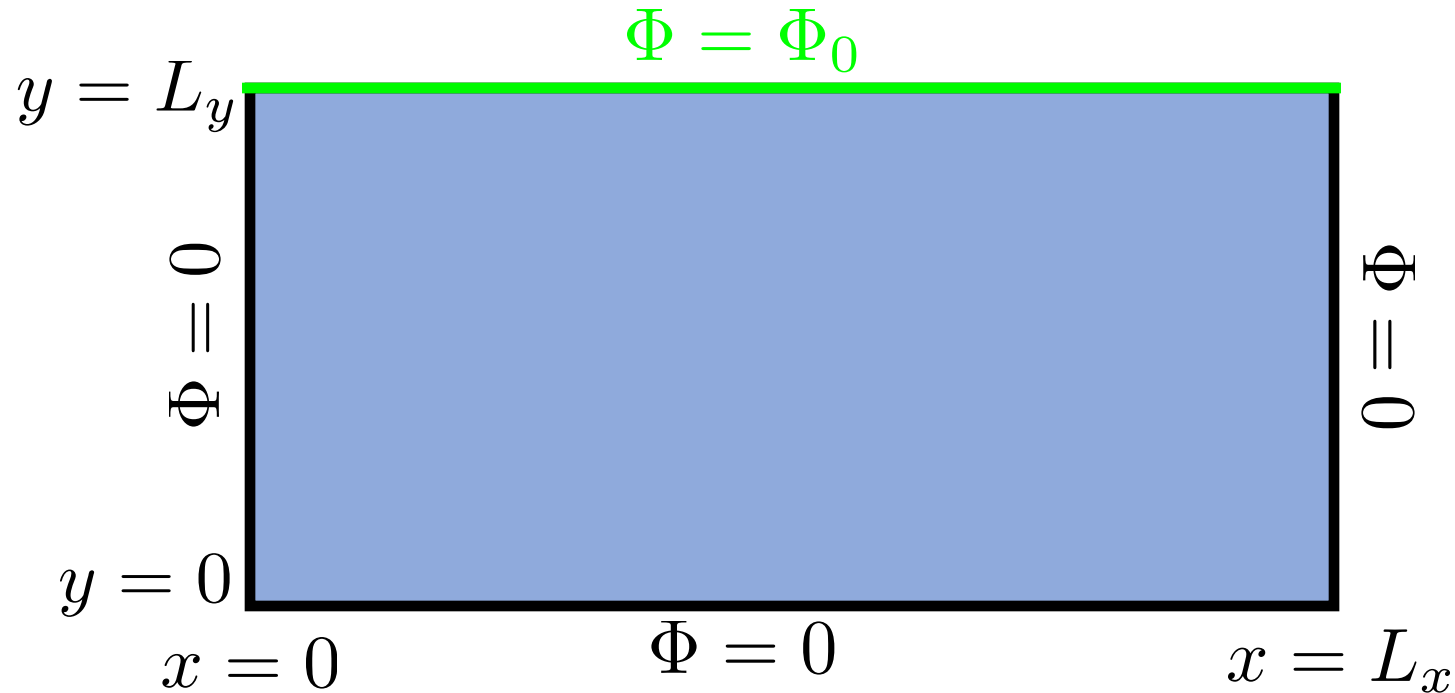
Solution of Laplace's eq. ODEs

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- Recall that k is complex, so solutions are “symmetric”
- To get the coefficients, we need to specify the boundary conditions

$$\Phi(x = 0, y) = \Phi(x = L_x, y) = \Phi(x, y = 0) = 0, \quad \Phi(x, y = L_y) = \Phi_0$$



Solution of Laplace's eq. ODEs

$$X(x) = C_s \sin(kx) + C_c \cos(kx), \quad Y(y) = C'_s \sinh(ky) + C'_c \cosh(ky)$$

- Use our boundary conditions:

$$\Phi(x = 0, y) = 0 \implies C_c = 0$$

$$\Phi(x, y = 0) = 0 \implies C'_c = 0$$

$$\Phi(x = L_x, y) = 0 \implies k = \frac{n\pi}{L_x}, \quad n = 1, 2, \dots$$

- So, we have solutions of the form:

$$c_n \sin\left(\frac{n\pi x}{L_x}\right) \sinh\left(\frac{n\pi y}{L_x}\right)$$

- Any linear combination is also a solution, so:

$$\Phi(x, y) = \sum_{n=1}^{\infty} c_n \sin\left(\frac{n\pi x}{L_x}\right) \sinh\left(\frac{n\pi y}{L_x}\right)$$

Solution of Laplace's equation

- Now we use our last boundary condition:

$$\Phi_0 = \sum_{n=1}^{\infty} c_n \sin \left(\frac{n\pi x}{L_x} \right) \sinh \left(\frac{n\pi L_y}{L_x} \right)$$

- To solve the equation, multiply both sides by $\sin(m\pi x/L_x)$ and integrate from 0 to L_x :

$$\int_0^{L_x} dx \Phi_0 \sin \left(\frac{m\pi x}{L_x} \right) = \sum_{n=1}^{\infty} c_n \sinh \left(\frac{n\pi L_y}{L_x} \right) \int_0^{L_x} dx \sin \left(\frac{m\pi x}{L_x} \right) \sin \left(\frac{n\pi x}{L_x} \right)$$

- Left-hand side integral:

$$\int_0^{L_x} dx \sin \left(\frac{m\pi x}{L_x} \right) = \begin{cases} 2L_x/m\pi, & m \text{ odd} \\ 0, & m \text{ even} \end{cases}$$

Solution of Laplace's equation

- Sum on the right-hand side simplifies because:

$$\int_0^{L_x} dx \sin\left(\frac{m\pi x}{L_x}\right) \sin\left(\frac{n\pi x}{L_x}\right) = \frac{L_x}{2} \delta_{n,m}$$

- So, we have:

$$\Phi_0 \frac{2L_x}{\pi m} = c_m \sinh\left(\frac{m\pi L_y}{L_x}\right) \frac{L_x}{2}, \quad m = 1, 3, 5, \dots$$

- So:
$$c_m = \frac{4\Phi_0}{\pi m \sinh\left(\frac{m\pi L_y}{L_x}\right)}, \quad m = 1, 3, 5, \dots$$

Solution of Laplace's equation

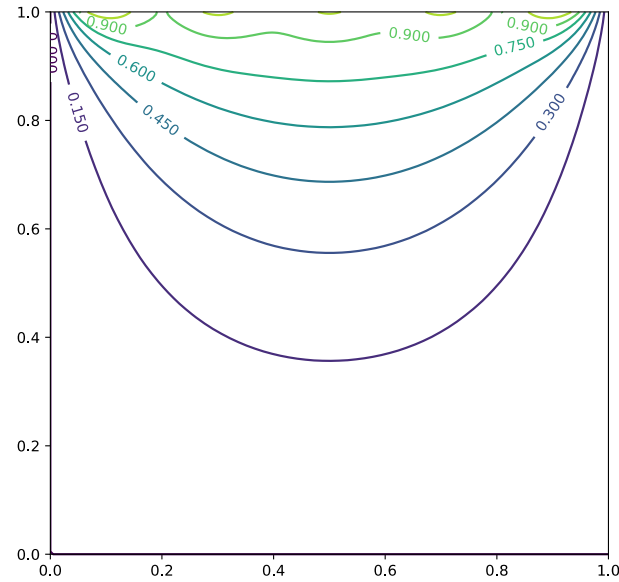
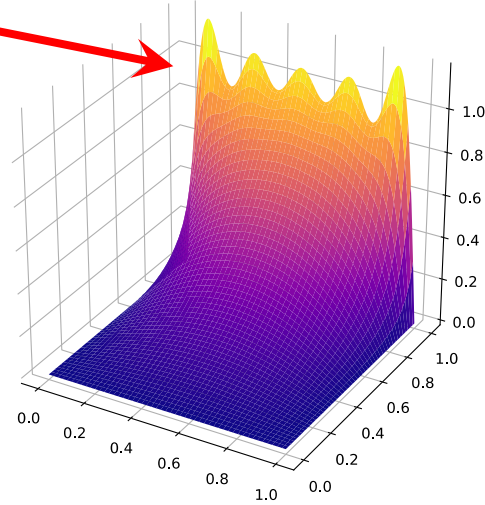
- Our final solution of Laplace's equation with our chosen boundary conditions:

$$\Phi(x, y) = \Phi_0 \sum_{n=1,3,5,\dots}^{\infty} \frac{4}{\pi n} \sin\left(\frac{n\pi x}{L_x}\right) \frac{\sinh\left(\frac{n\pi y}{L_x}\right)}{\sinh\left(\frac{n\pi L_y}{L_x}\right)}$$

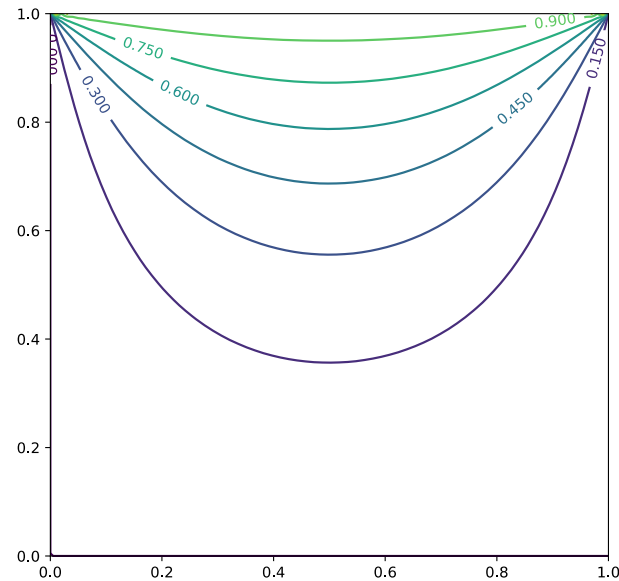
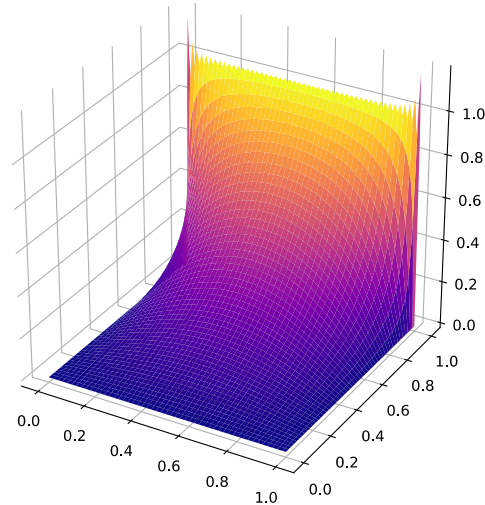
Analytical solution to Laplace equation

“Gibbs phenomenon,”
oscillations of Fourier series for
discontinuous function

5 terms in the sum:



50 terms in the sum:



Numerical solution of the Laplace equation

- To do this, we'll go back to the *diffusion* equation we have solved previously, this time in two spatial dimensions:

$$\frac{\partial T(x, y, t)}{\partial t} = \kappa \left(\frac{\partial^2 T(x, y, t)}{\partial x^2} + \frac{\partial^2 T(x, y, t)}{\partial y^2} \right)$$

- Given an initial temperature profile and stationary boundary conditions, the solution will eventually relax to some steady state:

$$\lim_{t \rightarrow \infty} T(x, y, t) = T_s(x, y)$$

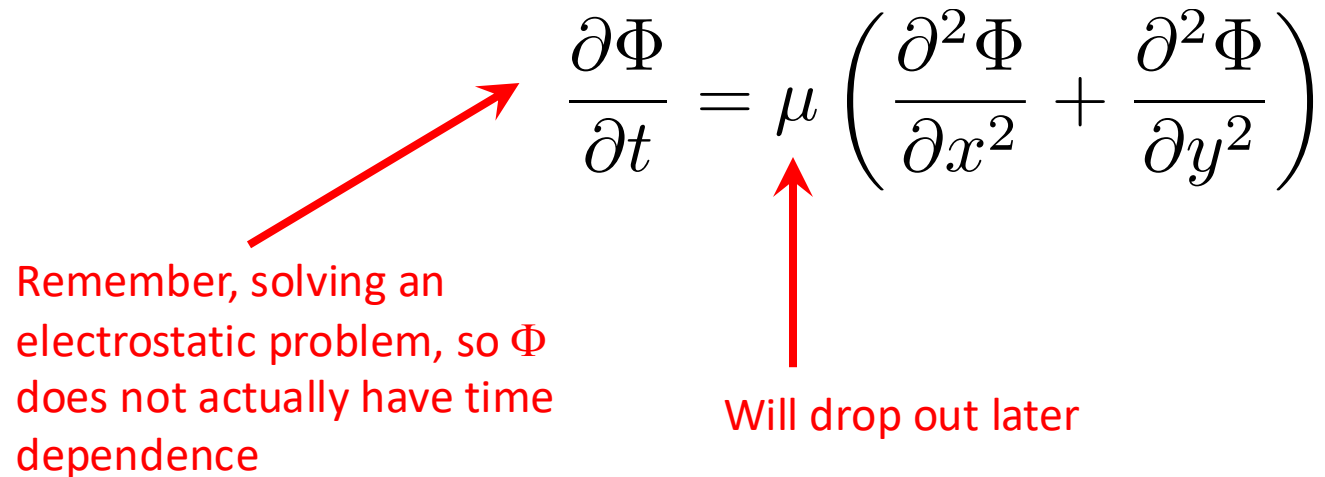
- In this state $\partial T / \partial t = 0$, so:

$$\frac{\partial^2 T_s}{\partial x^2} + \frac{\partial^2 T_s}{\partial y^2} = 0$$

- We can think of the Laplace equation as the steady-state of the diffusion equation

Relaxation methods

- Methods based on this physical intuition are called relaxation methods
- We can use the FTCS method that we have used previously for the diffusion equation
- Start with the 2D “diffusion” equation:



Remember, solving an electrostatic problem, so Φ does not actually have time dependence

$$\frac{\partial \Phi}{\partial t} = \mu \left(\frac{\partial^2 \Phi}{\partial x^2} + \frac{\partial^2 \Phi}{\partial y^2} \right)$$

Will drop out later

Relaxation methods

- Methods based on this physical intuition are called relaxation methods
- We can use the FTCS method that we have used previously for the diffusion equation

- Start with the 2D “diffusion” equation:

$$\frac{\partial \Phi}{\partial t} = \mu \left(\frac{\partial^2 \Phi}{\partial x^2} + \frac{\partial^2 \Phi}{\partial y^2} \right)$$

- Discretize:

$$\begin{aligned} \Phi_{i,j}^{n+1} = & \Phi_{i,j}^n + \frac{\mu\tau}{h_x^2} (\Phi_{i+1,j}^n + \Phi_{i-1,j}^n - 2\Phi_{i,j}^n) \\ & + \frac{\mu\tau}{h_y^2} (\Phi_{i,j+1}^n + \Phi_{i,j-1}^n - 2\Phi_{i,j}^n) \end{aligned}$$

- n here is not really time, more an improved guess for the solution

Jacobi relaxation method

- Recall that FTCS is stable for $\mu\tau/h^2 \leq 1/2$

- In 2D the stability criteria is :

$$\frac{\mu\tau}{h_x^2} + \frac{\mu\tau}{h_y^2} \leq \frac{1}{2}$$

- If $h_x = h_y = h$, then the criterion is

$$\frac{\mu\tau}{h^2} \leq \frac{1}{4}$$

- Since we want to take n to infinity, we choose the largest timestep:

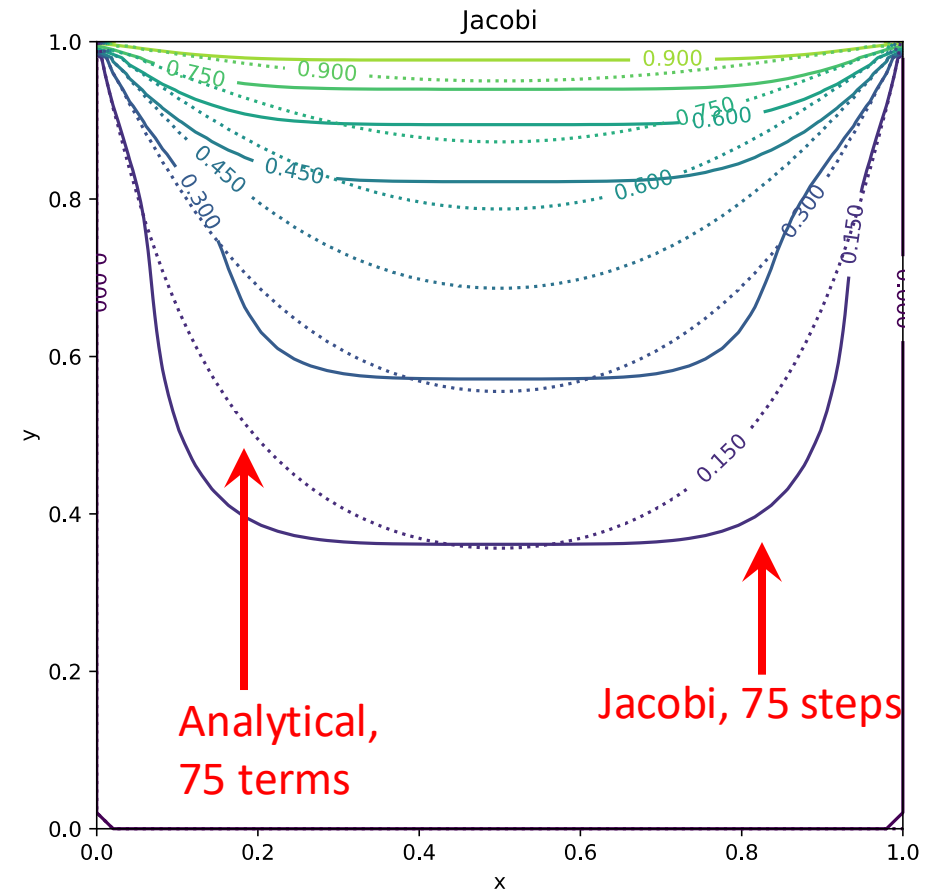
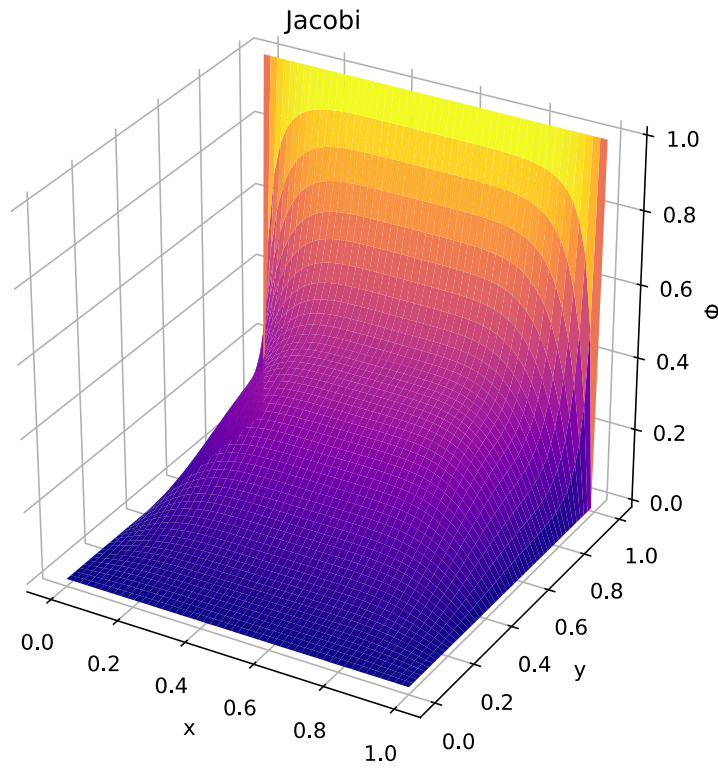
$$\Phi_{i,j}^{n+1} = \frac{1}{4}(\Phi_{i+1,j}^n + \Phi_{i-1,j}^n + \Phi_{i,j+1}^n + \Phi_{i,j-1}^n)$$

Jacobi method for Laplace equation

$$\Phi_{i,j}^{n+1} = \frac{1}{4}(\Phi_{i+1,j}^n + \Phi_{i-1,j}^n + \Phi_{i,j+1}^n + \Phi_{i,j-1}^n)$$

- Note that the μ has dropped out
- Involves replacing the value of the potential at a point with the average value of the four nearest neighbors
 - Discrete version of mean-value theorem for the electrostatic potential
- This equation is for the interior points (exterior are set by boundary conditions)
- Simple to generalize for Poisson equation

Jacobi method for Laplace equation



Gauss-Seidel and simultaneous overrelaxation

- **Gauss-Seidel**: We can improve the convergence over the Jacobi method by using updated values of the potential as they are calculated:

$$\Phi_{i,j}^{n+1} = \frac{1}{4} (\Phi_{i+1,j}^n + \Phi_{i-1,j}^{n+1} + \Phi_{i,j+1}^n + \Phi_{i,j-1}^{n+1})$$

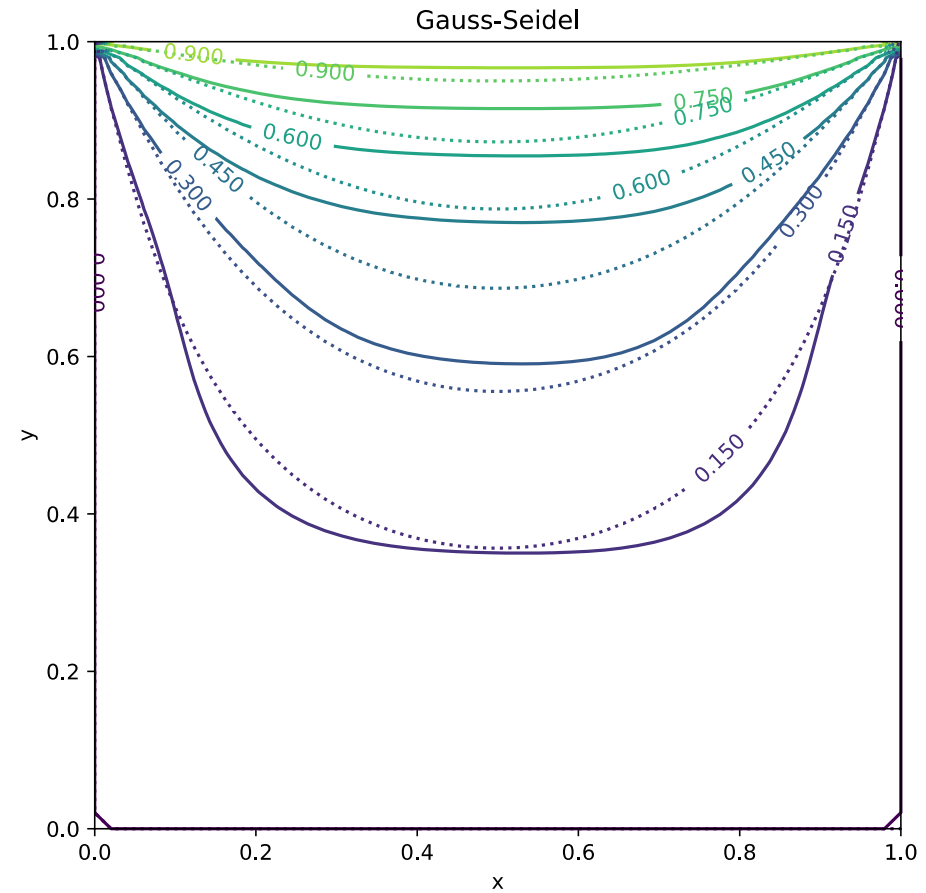
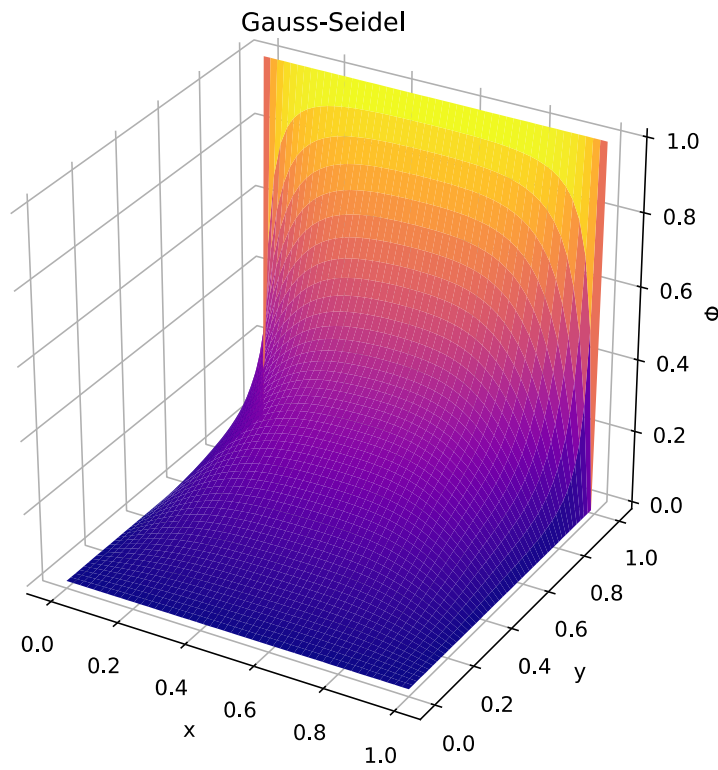
- **Simultaneous overrelaxation**: Choose a mixing parameter ω :

$$\Phi_{i,j}^{n+1} = (1 - \omega) \Phi_{i,j}^n + \frac{\omega}{4} (\Phi_{i+1,j}^n + \Phi_{i-1,j}^{n+1} + \Phi_{i,j+1}^n + \Phi_{i,j-1}^{n+1})$$

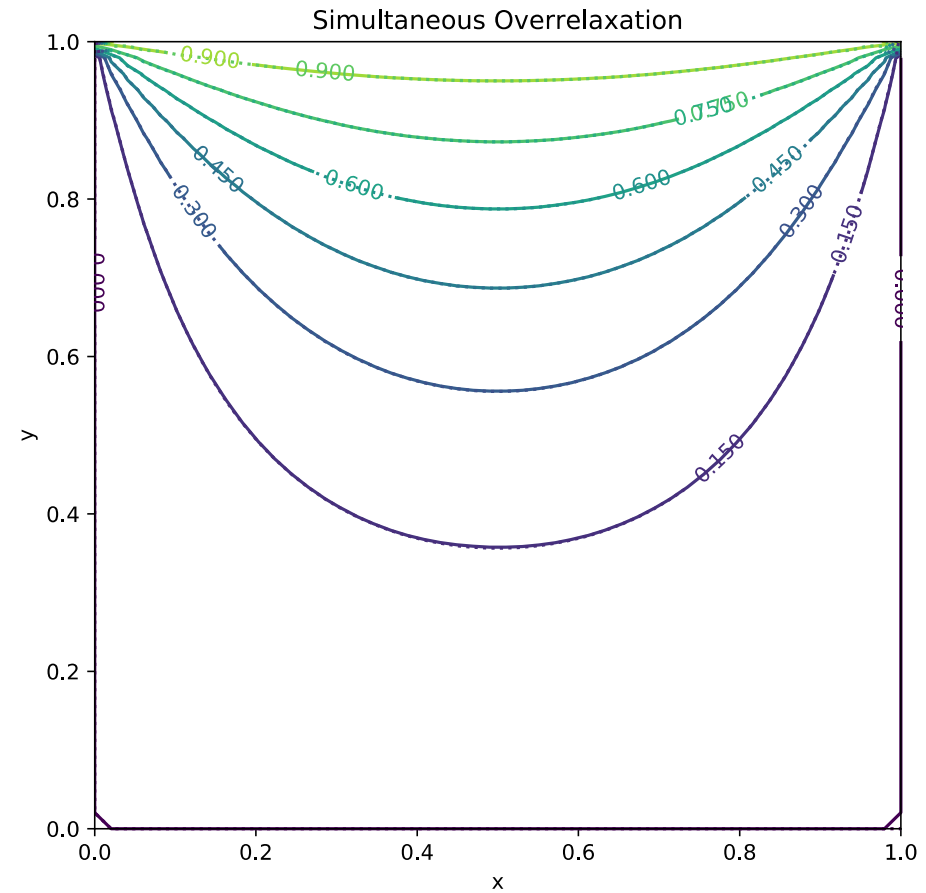
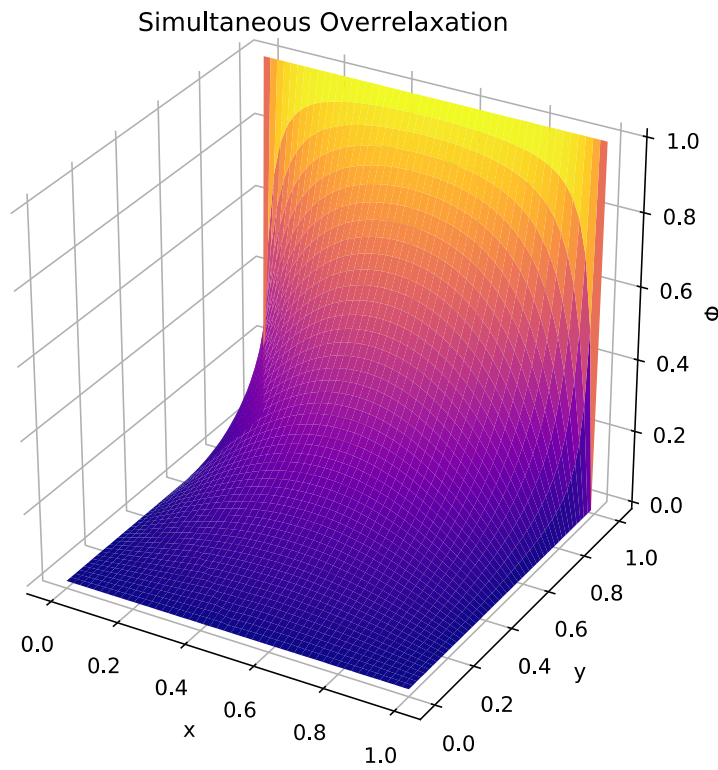
- $\omega < 1$ slows convergence, $\omega > 2$ is unstable
- Often chosen by trial and error
- E.g., for a square geometry with equal discretization, often a good choice:

$$\omega_{\text{opt}} = \frac{2}{1 + \sin(\pi/N)}$$

Gauss-Seidel for Laplace equation



Simultaneous overrelaxation for Laplace eq.



Recall: Jacobi iterative method

- Starting with a linear system: $a_{11}x_1 + a_{12}x_2 + \cdots + a_{1n}x_n = b_1$

$$a_{21}x_1 + a_{22}x_2 + \cdots + a_{2n}x_n = b_2$$

$$\vdots \quad \quad \quad \vdots \quad \quad \quad \vdots \quad \quad \quad \vdots$$

$$a_{n1}x_1 + a_{n2}x_2 + \cdots + a_{nn}x_n = b_n$$

- Pick initial guesses $\mathbf{x}^k$, solve equation i for i th unknown to get an improved guess:

$$x_1^{k+1} = -\frac{1}{a_{11}}(a_{12}x_1^k + a_{13}x_2^k + \cdots + a_{1n}x_n^k - b_1)$$

$$x_2^{k+1} = -\frac{1}{a_{22}}(a_{21}x_1^k + a_{23}x_2^k + \cdots + a_{2n}x_n^k - b_2)$$

$$\vdots \quad \quad \quad \vdots \quad \quad \quad \vdots \quad \quad \quad \vdots$$

$$x_n^{k+1} = -\frac{1}{a_{nn}}(a_{n1}x_1^k + a_{n2}x_2^k + \cdots + a_{n,n-1}x_{n-1}^k - b_n)$$

Recall: Jacobi iterative method

- We can write an element-wise formula for $\mathbf{x}$:

$$x_i^{k+1} = \frac{1}{a_{ii}} \left(b_i - \sum_{j \neq i} a_{ij} x_j^k \right)$$

- Or:

$$\mathbf{x}_i^{k+1} = \mathbf{D}^{-1} (\mathbf{b} - (\mathbf{A} - \mathbf{D})\mathbf{x}^k)$$

- Where $\mathbf{D}$ is a diagonal matrix constructed from the diagonal elements of $\mathbf{A}$
- Convergence is guaranteed if matrix is diagonally dominant (but works in other cases):

$$a_{ii} > \sum_{j=1, j \neq i}^N |a_{ij}|$$

The iterative methods discussed here are the same as we used to solve linear systems

- Can interpret Φ as a vector, so are solving $\mathbf{A}\Phi=\mathbf{b}$
- Going back to our initial discretization of the Laplace equation (for $h_x=h_y$):

$$\frac{1}{h^2} (\Phi_{i+1,j}^n + \Phi_{i-1,j}^n + \Phi_{i,j+1}^n + \Phi_{i,j-1}^n - 4\Phi_{i,j}^n) = 0$$

- Note that $\mathbf{A}$ is a banded matrix with 4's on the diagonal, 1's on off-diagonal elements
- This is when the Jacobi method is guaranteed to be accurate (diagonally dominated)!
- Same holds for Gauss-Seidel and SOR

Today's lecture: Elliptical PDEs

- Relaxation methods
- Spectral methods

A different way to represent the potential

- Consider again the Poisson equation:

$$\nabla^2 \Phi(\mathbf{r}) = -\frac{1}{\epsilon_0} \rho(\mathbf{r})$$

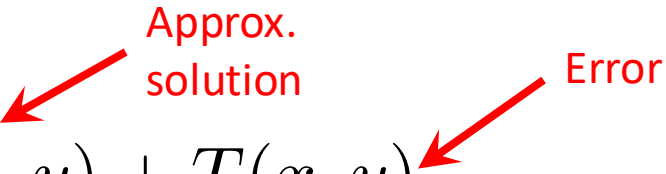
- For simplicity, square geometry: $0 \leq x \leq L, \quad 0 \leq y \leq L$
- Relaxation methods discretize space and solve for $\Phi_{i,j}$
- We constructed out analytical solution as in **infinite sum of trigonometric functions**
- Let's build an approximate solution as a **finite sum**:

$$\Phi(x, y) = a_1 f_1(x, y) + a_2 f_2(x, y) + \cdots + a_K f_K(x, y) + T(x, y)$$

$$= \sum_{k=1}^K a_k f_k(x, y) + T(x, y)$$

$$= \Phi_a(x, y) + T(x, y)$$

Approximate solution

$$\Phi(x, y) = \Phi_a(x, y) + T(x, y)$$


The diagram shows two red arrows pointing to the terms in the equation. One arrow points from the text 'Approx. solution' to the term $\Phi_a(x, y)$. The other arrow points from the text 'Error' to the term $T(x, y)$.

- To simplify the approximate solution, we take orthogonal trial functions:

$$\int_0^L dx \int_0^L dy f_k(x, y) f_{k'}(x, y) = A_k \delta_{k, k'}$$

- Insert into the Poisson equation:

$$\nabla^2 \left[\sum_k a_k f_k(x, y) \right] + \frac{1}{\epsilon_0} \rho(x, y) = R(x, y)$$

- Where the residual R is:

$$R(x, y) = -\nabla^2 T(x, y)$$

Obtain coefficients with Galerkin method

- Next step is to obtain coefficients a_k
- Galerkin method imposes the condition that the residual is orthogonal to all of the trial functions:

$$\int_0^L dx \int_0^L dy f_k(x, y) R(x, y) = 0$$

- Choice of trial functions motivated by geometry and boundary conditions
- Let's take **Neumann** boundary conditions:

$$\left. \frac{\partial \Phi}{\partial x} \right|_{x=0} = \left. \frac{\partial \Phi}{\partial x} \right|_{x=L} = \left. \frac{\partial \Phi}{\partial y} \right|_{y=0} = \left. \frac{\partial \Phi}{\partial y} \right|_{y=L} = 0$$

- Normal component of electric field zero at the boundaries

Trial functions for our geometry and BCs

- Natural set of trial functions:

$$f_{m,n}(x, y) = \cos \left[\frac{m\pi x}{L} \right] \cos \left[\frac{n\pi y}{L} \right]$$

- Can confirm that these functions are orthogonal:

$$\int_0^L dx \int_0^L dy f_{m,n}(x, y) f_{m',n'}(x, y) = \frac{L^2}{4} (1 + \delta_{m,0})(1 + \delta_{n,0}) \delta_{m,m'} \delta_{n,n'}$$

- Inserting into Poisson equation

$$\nabla^2 \left[\sum_k a_k f_k(x, y) \right] + \frac{1}{\epsilon_0} \rho(x, y) = R(x, y)$$

- Gives:

$$- \sum_{m=0}^{M-1} \sum_{n=0}^{M-1} a_{m,n} \frac{\pi^2(m^2 + n^2)}{L^2} f_{m,n}(x, y) + \frac{1}{\epsilon_0} \rho(x, y) = R(x, y)$$

Now we need so solve for coefficients

- Apply to both sides of the equation:

$$\int_0^L dx \int_0^L dy f_{m',n'}(x, y)$$

- And use “Galerkin condition”:

$$\int_0^L dx \int_0^L dy f_k(x, y) R(x, y) = 0$$

- Which gives:

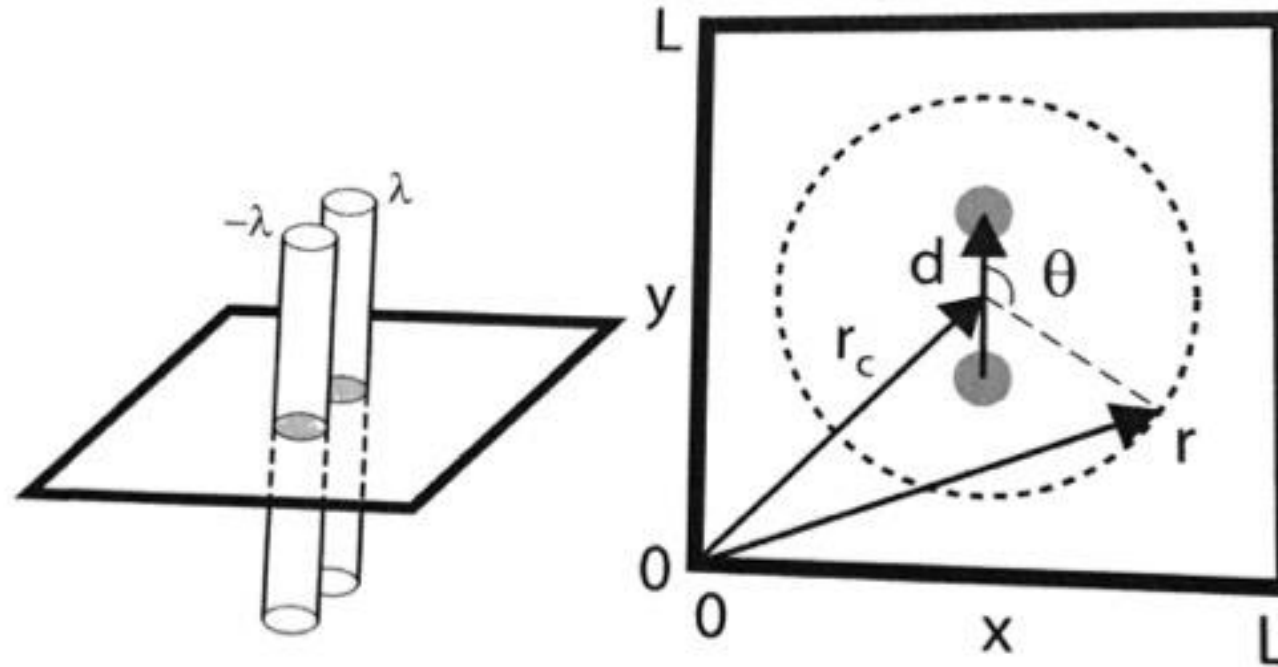
$$a_{m,n} = \frac{4}{\pi^2 \epsilon_0 (m^2 + n^2) (1 + \delta_{m,0}) (1 + \delta_{n,0})} \int_0^L dx \int_0^L dy \rho(x, y) \cos\left(\frac{m\pi x}{L}\right) \cos\left(\frac{n\pi y}{L}\right)$$

Final solution with Galerkin method:

$$\Phi_a(x, y) = \sum_{m=0}^{M-1} \sum_{n=0}^{M-1} a_{m,n} \cos\left(\frac{m\pi x}{L}\right) \cos\left(\frac{n\pi y}{L}\right)$$

$$a_{m,n} = \frac{4}{\pi^2 \epsilon_0 (m^2 + n^2) (1 + \delta_{m,0}) (1 + \delta_{n,0})} \int_0^L dx \int_0^L dy \rho(x, y) \cos\left(\frac{m\pi x}{L}\right) \cos\left(\frac{n\pi y}{L}\right)$$

Ex: charge distribution of 2D dipoles (Garcia Sec. 8.2)



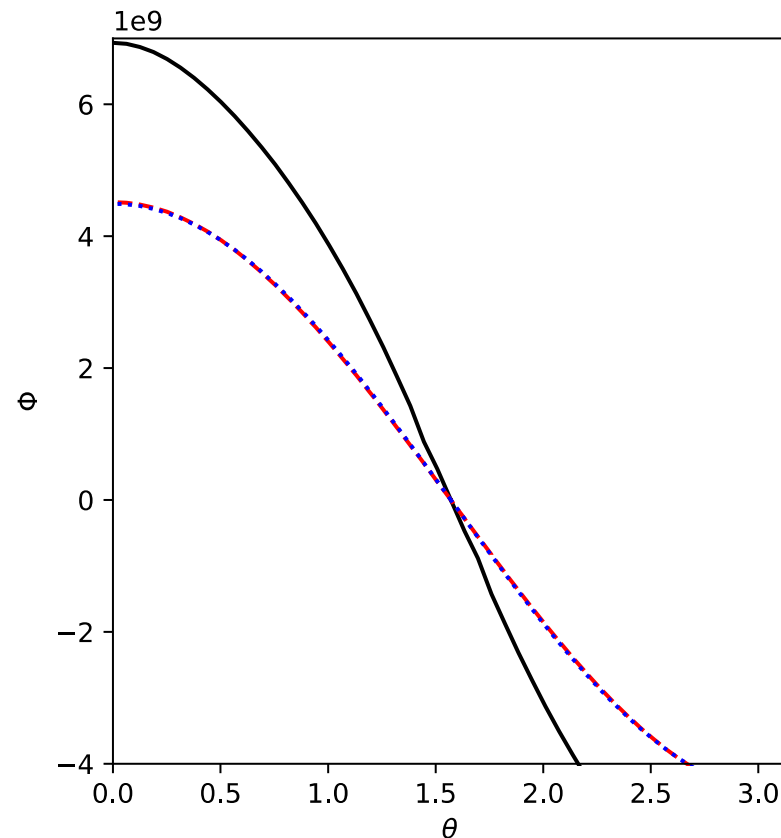
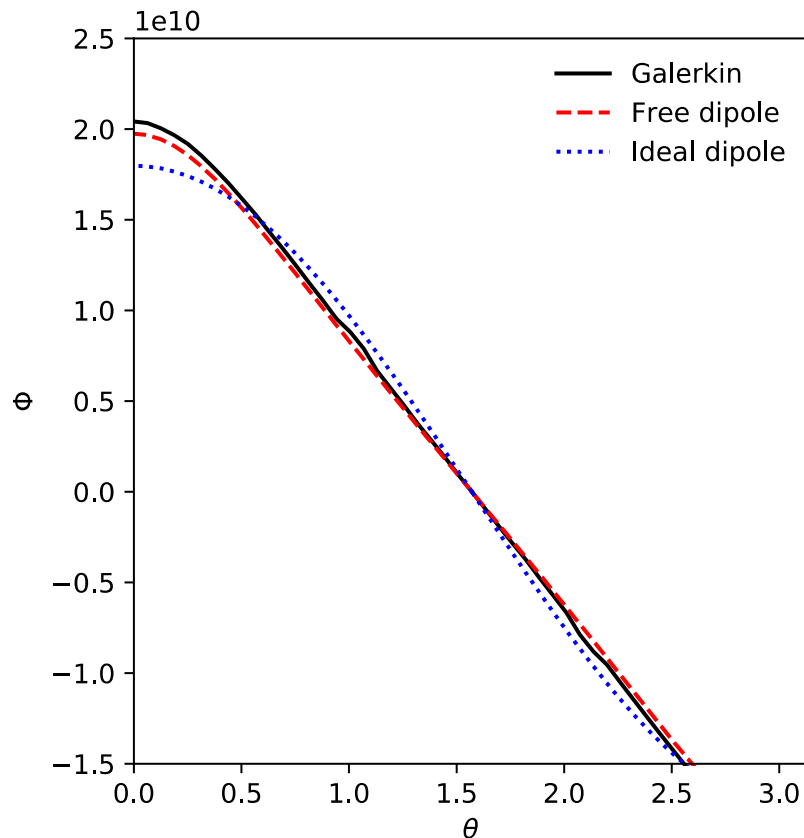
$$\rho(\mathbf{r}) = \lambda[\delta(\mathbf{r} - \mathbf{r}_+) - \delta(\mathbf{r} - \mathbf{r}_-)]$$

- Where:

$$\mathbf{r}_{\pm} = \mathbf{r}_c \pm \frac{1}{2}\mathbf{d}$$

Galerkin solution to the dipole potential

- Compare to free dipole: $\Phi^{\text{free}}(\mathbf{r}) = -\frac{\lambda}{2\pi\epsilon_0} [\ln |\mathbf{r} - \mathbf{r}_+| - \ln |\mathbf{r} - \mathbf{r}_-|]$
- Or “ideal” dipole potential (far away): $\Phi^{\text{ideal}}(\mathbf{r}) = \frac{\lambda}{2\pi\epsilon_0} \frac{|\mathbf{d}|}{|\mathbf{r} - \mathbf{r}_c|} \cos \theta$



Comments on the Galerkin method

- Can choose any trial functions that are orthogonal and obey the boundary conditions
 - In contrast to the separation of variables, where we first found general solutions to PDE, the imposed boundary conditions
- Should be interpreted as a **spectral transform approach**, i.e., representing the solution as a Fourier series
 - In our example, it was a cosine series because of our boundary conditions
- Did not use a spatial grid
 - Convenient if only need the answer at specific points
 - Inefficient if we want to map the potential over the whole range, because of the computation of the prefactors, especially for a more complex potential

After class tasks

- Homework 3 due tomorrow Oct. 22
- Homework 4 will be posted soon
- Readings
 - Garcia Chapters 7 and 8
 - [Mike Zingale's notes on computational hydrodynamics](#)