

Method of Relaxation in 1D

We are trying to solve Laplace's Equation

$$\nabla^2 V = 0$$

In some (many?) cases, it makes sense to solve it numerically because the solutions are not necessarily <sup>analytic</sup> functions. However, we can often exploit general property #3:

$$V(\vec{r}) = \frac{1}{4\pi r^2} \oint V dA \quad V \text{ at a point is the average of its neighbors.}$$

We will approach this problem in 1D first, which arguably is boring b/c it's always linear, because it will help us gain intuition about the algorithm and avoid common pitfalls.

1D Laplace Equation

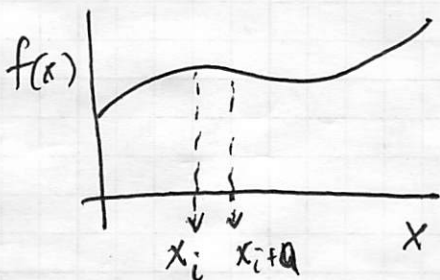
$$V(x) = ? \quad \text{when} \quad \frac{d^2 V}{dx^2} = 0$$

to develop a computational description of this problem we need to develop the concept of a numerical derivative of a function.



$$\frac{df}{dx} = ?$$

Consider some function  $f(x)$ ,



Let's cut up the space into equally-sized steps of size  $a$ .

We can approximate the slope (derivative) of that function using a line,

$$\frac{df}{dx} \approx \frac{f(x_i+a) - f(x_i)}{a} \quad \star \text{ this would be the approx } \frac{df}{dx} @ x_i + \frac{a}{2}$$

So,

$$f'(x_i + \frac{a}{2}) \approx \frac{f(x_i+a) - f(x_i)}{a} \quad \text{and}$$

$$f'(x_i - \frac{a}{2}) \approx \frac{f(x_i) - f(x_i-a)}{a}$$

What about the second derivative,  $f''(x_i)$ ?

given approximate values for  $f'$ , we can further approximate  $f''$  using the same ideas,

$$f''(x_i) \approx \frac{f'(x_i + \frac{a}{2}) - f'(x_i - \frac{a}{2})}{a}$$

$\star$  This would be  $\frac{d^2f}{dx^2} @ x_i$  b/c  $\pm \frac{a}{2}$  are used as the bounds.

Can we write that in terms of  $f(x)$  (as that's what we have)?

$$f''(x_i) \approx \frac{f'(x_i + \frac{a}{2}) - f'(x_i - \frac{a}{2})}{a}$$

with  $f'(x_i + \frac{a}{2}) \approx \frac{f(x_i + a) - f(x_i)}{a}$

$$f'(x_i - \frac{a}{2}) \approx \frac{f(x_i) - f(x_i - a)}{a}$$

$$f''(x_i) \approx \frac{[f(x_i + a) - f(x_i)]/a - [f(x_i) - f(x_i - a)]/a}{a}$$

$$= \frac{f(x_i + a) - 2f(x_i) + f(x_i - a)}{a^2}$$

What we have derived is a description of the numerical second derivative at a point  $x_i$  for the function  $f$ .

### Relating this to Laplace's Equation

$$\nabla^2 V = 0 \Rightarrow \text{in 1D} \Rightarrow \frac{d^2 V}{dx^2} = 0$$

$$\frac{V(x_i + a) - 2V(x_i) + V(x_i - a)}{a^2} = 0$$

Re writing this equation shows us precisely what we expect ~~for~~ from GP #3:

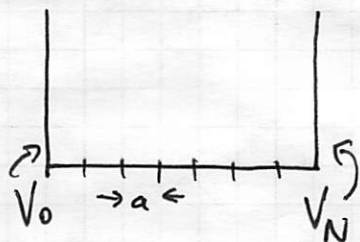
$$V(x_i) = \frac{1}{2} [V(x_i + a) + V(x_i - a)]$$

The potential at  $x_i$  is equal to the average of the potential at  $x_i + a$  and  $x_i - a$ . Note  $a$  can be reduced for better resolution.



Making this a computational task

Consider a 1D Laplace Eqn problem,  
with set Boundary Conditions,



$$V(x_0) = V_0$$

$$V(x_N) = V_N$$

We will cut up the space, called a "mesh",  
in equal size chunks,  $a$ .

Pseudocode:

let  $V(x_i) =$  some reasonable random  
#s. (e.g. between  $V_0$  &  $V_N$ )

then,

for  $i$  in range  $(1, N-1)$

$$V(x_i) = \frac{1}{2} [V(x_i - a) + V(x_i + a)]$$

→ this completes one execution,

Continue until error converges to  
selected error or to max number  
of iterations.

Convergence, when error is acceptable.

for  
kth  
run

$$\text{error} = V_{k+1} - V_k$$

$$\text{max}(\text{error}) < \text{acceptance?}$$

Stop

else run again

} just one  
possible  
way.

Let's go back to the 2D problem,

$$\nabla^2 V(x, y) = \frac{\partial^2 V}{\partial x^2} + \frac{\partial^2 V}{\partial y^2} = 0$$

We can develop an approximate form for this PDE, by approximating the derivative,

$$\frac{\partial^2 V}{\partial x^2} = \frac{V(x+a, y) - 2V(x, y) + V(x-a, y)}{a^2}$$

$$\frac{\partial^2 V}{\partial y^2} = \frac{V(x, y+a) - 2V(x, y) + V(x, y-a)}{a^2}$$

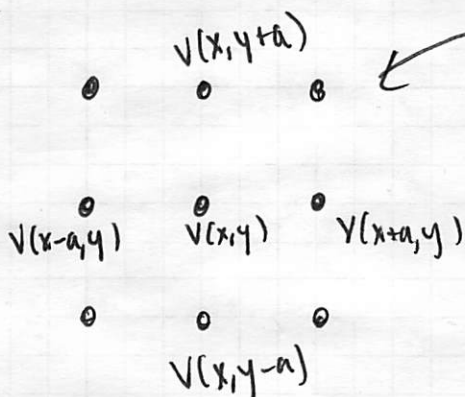
so that,

$$\frac{\partial^2 V}{\partial x^2} + \frac{\partial^2 V}{\partial y^2} = \frac{V(x+a, y) + V(x, y+a) + V(x-a, y) + V(x, y-a) - 4V(x, y)}{a^2}$$

$$\text{B/c } \nabla^2 V(x, y) = 0,$$

$$V(x, y) = \frac{1}{4} \left[ V(x+a, y) + V(x, y+a) + V(x-a, y) + V(x, y-a) \right]$$

$V$  @  $x, y$  is average of surrounding pts!



these pts. we call the mesh.

to find  $V(x, y)$ , we successively average over the surrounding mesh pts.