

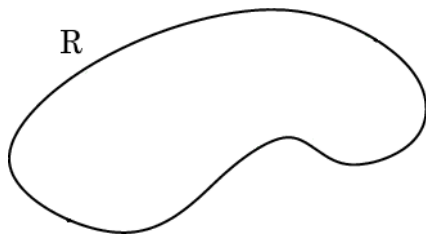
# *Brownian Motion and Harmonic Functions*

Harini Chandramouli  
Kiya Holmes  
Brandon Reeves  
Nora Stack

Michigan State University

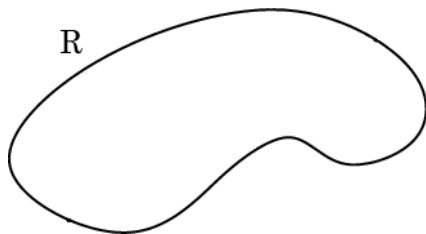
2012.08.02

# Laplace's Equation



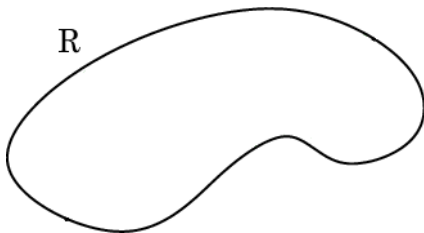
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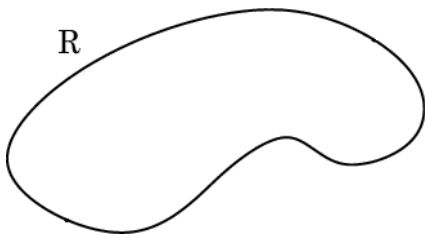
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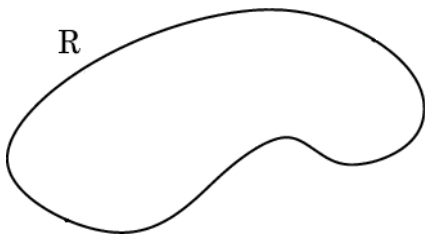
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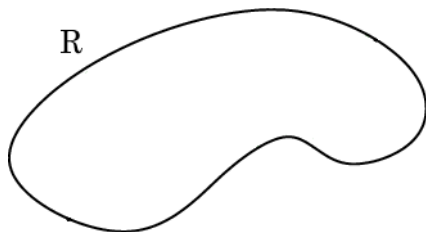


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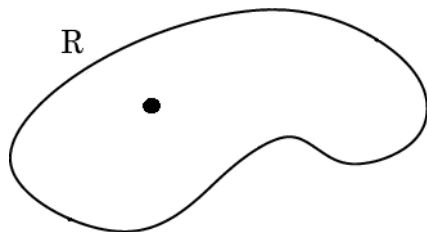
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  - $u_{xx} + u_{yy} = 0$  (Laplace's Equation)
  - In nice regions, the solution is well-known



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- We can use random walks to simulate Brownian motion
  - Specifically, the random walks on circles (RWoC) and spheres (RWoS)
  - We simulated Brownian motion in various regions and studied the probability density functions (PDFs) of the point of first encounter in these regions

- Pick point  $(x_0, y_0)$  in the region

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- Continue until process until you hit boundary of area

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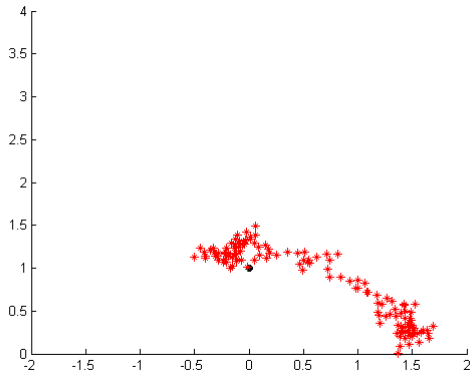
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  - Upper Half-Space
  - Sphere

## *Simulation: Half Plane*

- Beginning at  $(x_0, y_0)$ , with  $y_0 > 0$  we simulate Brownian motion on the upper half plane

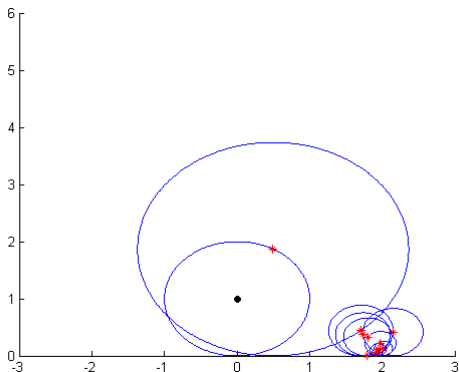
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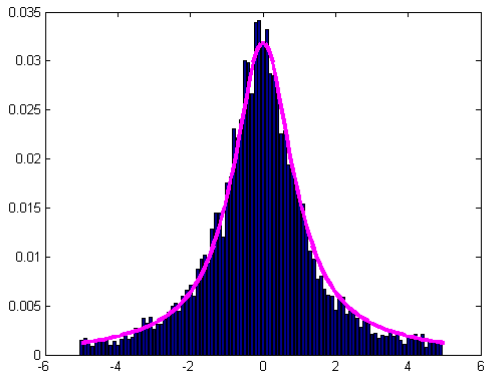


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- How did our simulation perform?



## *More General Regions*

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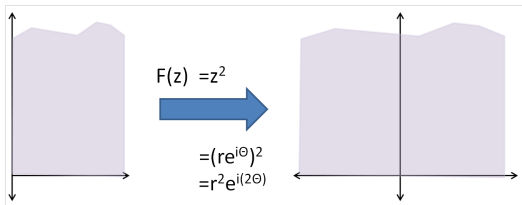
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# Using Conformal Mappings



- PDF

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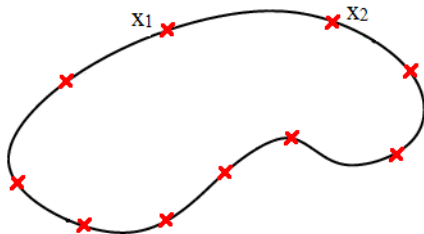
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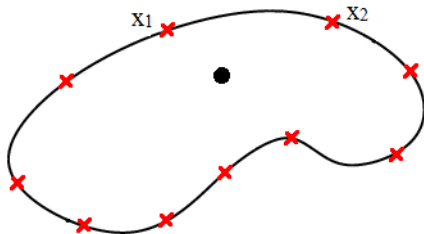
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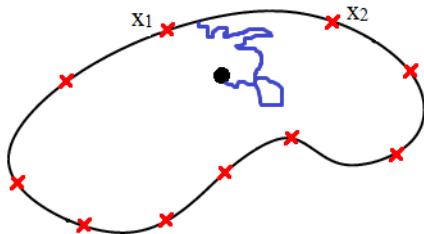


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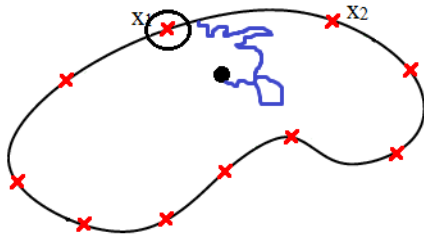
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- But we can do better,  $u_0$  can be up to a  $2(10) - 1$  degree polynomial

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  - Pick  $x_i$  as the roots of  $p_n(x)$

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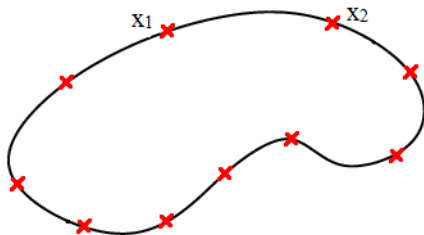
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# What Have We Done?



- So if  $u_0$  is a “nice” (smooth) function, then

$$\begin{aligned}u(x, y) &= \int_{-\infty}^{\infty} D(\tau) u_0(\tau) d\tau \\ &\approx \sum_{i=1}^{10} u_0(x_i) D_i\end{aligned}$$

- This will be a *good* approximation

- Brownian Motion and Laplace's Equation
- Walk on Circles and Spheres
- Simulating Walk on Circles and Spheres in different regions
- Probability Density Functions and Conformal Mapping Techniques
- Less “Expensive” Real World Applications

# Acknowledgements

- NSA
  - Grant: H98230-11-10222.
- Dr. Igor Nazarov
- Dr. Nick Boros



*Thanks for Listening!*

Questions?