

A BIDIRECTIONAL FORMULATION FOR WALK ON SPHERES

Yang Qi, Dario Seyb, Benedikt Bitterli, Wojciech Jarosz

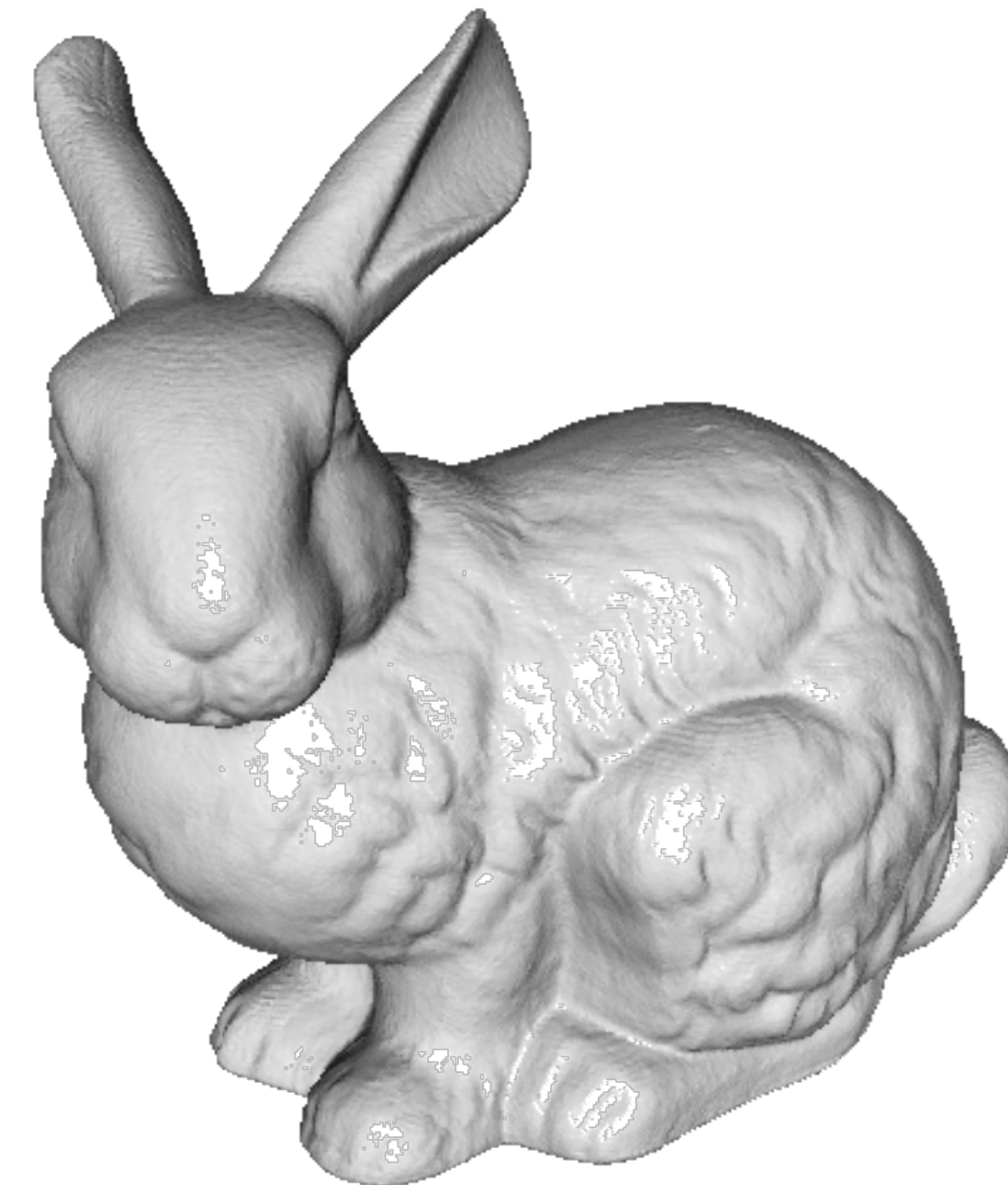
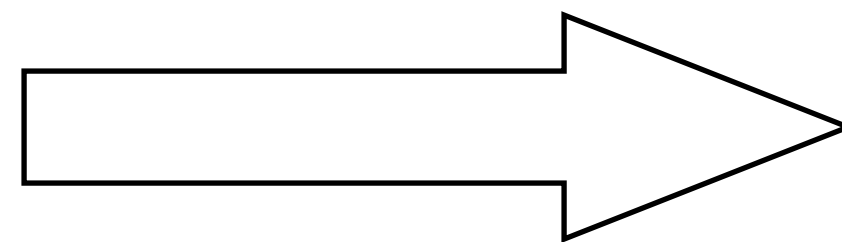
Diffusion curves

- ❑ Orzan, A. Bousseau, A. Winnemöller, H. Barla, P. Thollot, J. and Salesin, D.
“*Diffusion curves: a vector representation for smooth-shaded images*” (SIGGRAPH 2008)
- ❑ Bowers, J. Leahey, J. and Wang, R.
“*A Raytracing Approach to Diffusion Curve*” (EGSR 2009)
- ❑ Prévost, R. Jarosz, W. Sorkine-Hornung, O.
“*A Vectorial Framework for Ray Traced Diffusion Curve*” (Computer Graphics Form 2015)
- ❑ ...



Surface reconstruction

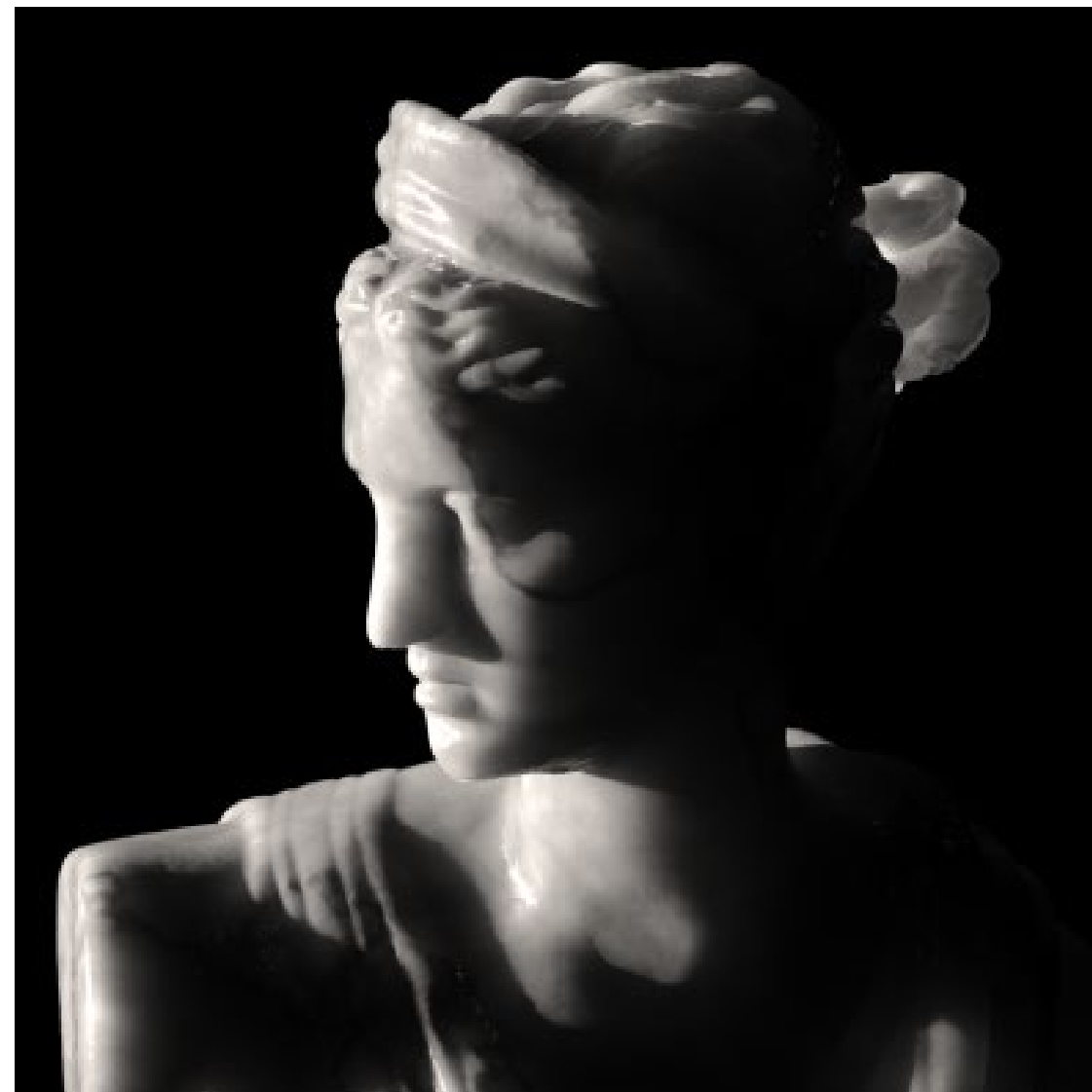
- Kazhdan, M. Bolith, M. and Hoppe, H.
“*Poisson Surface Reconstruction*” (SGP 2006)
- Kazhdan, M. and Hoppe, H.
“*Screened Poisson surface reconstruction*” (SIGGRAPH 2013)
- ...



<https://hhoppe.com/proj/screenedpoisson/>

Subsurface Scattering

- ❑ Joe Stam “*Multiple scattering as a diffusion process*” (EGSR 1995)
- ❑ Christensen, N. and Jensen, H.
“*A Practical Guide to Global Illumination using Photon maps*” (SIGGRAPH 2000)
- ❑ d’Eon, E. Irving, G.
“*A Quantized-Diffusion Model for Rendering Translucent Materials*” (SIGGRAPH 2011)
- ❑ ...



<https://graphics.stanford.edu/courses/cs348b-00/course8.pdf>



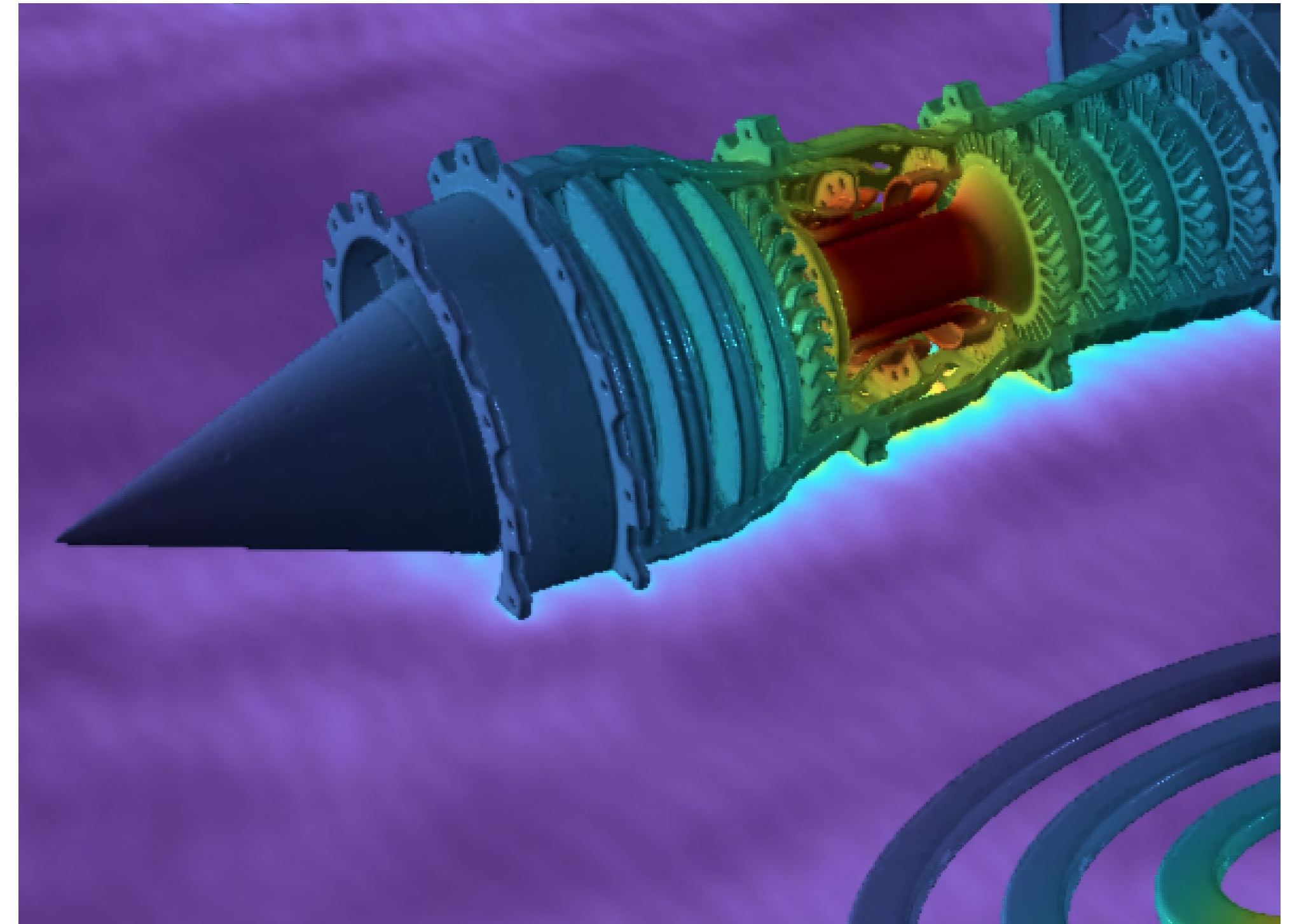
https://naml.us/paper/deon2011_subsurface.pdf

PDE Solver

Traditional PDE solver

- Finite element method
- Finite difference method

Meshing is costly for complex shape



cs.dartmouth.edu/wjarosz/publications/sawhneyseyb22gridfree.html

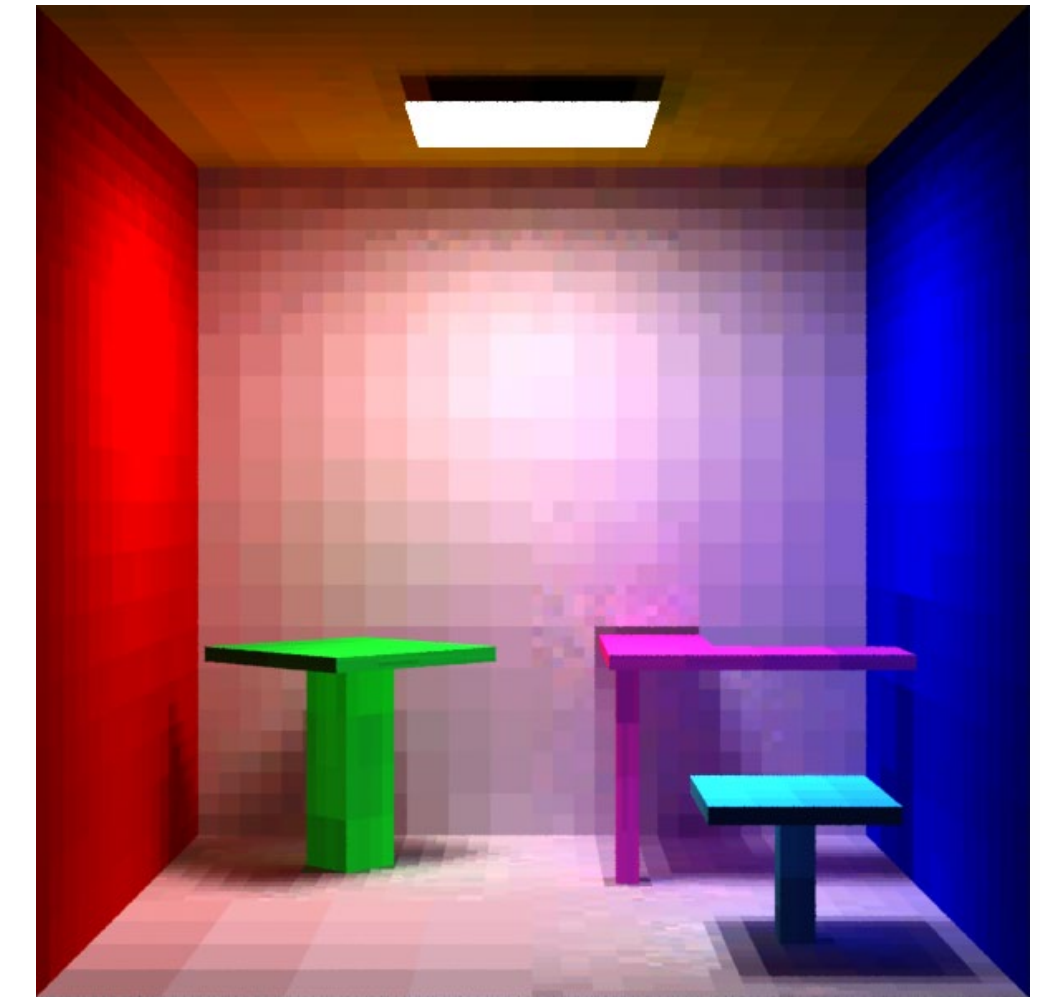
Methods in Rendering

Mesh based method

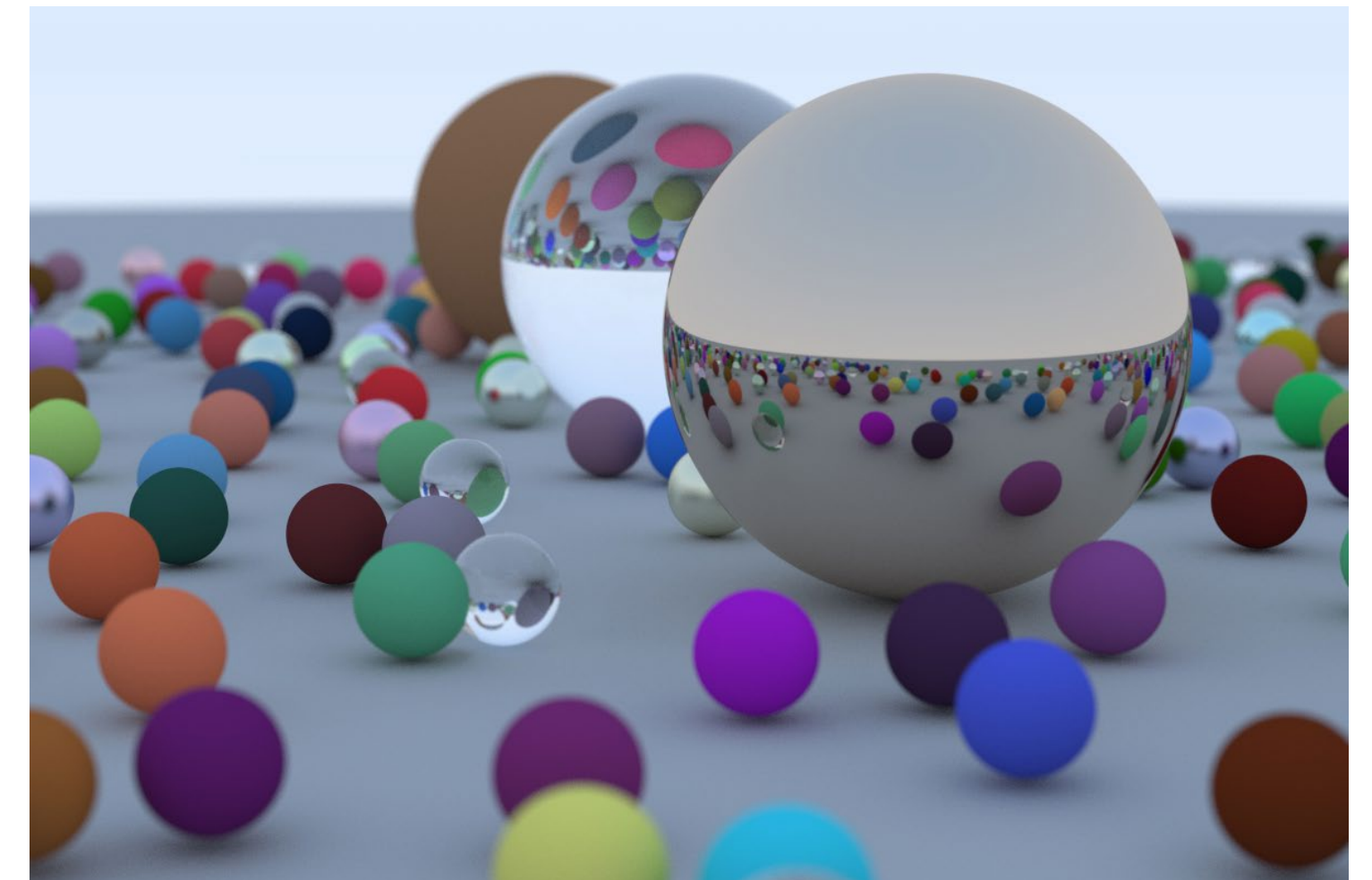
- ❑ *“Radiosity and Realistic Image Synthesis”*
Wallace, J. and Cohen, M. (1993)
- ❑ ...

Monte-Carlo method

- ❑ *“The Rendering Equation”*
James T. Kajiya. (1986)
- ❑ ...



www.cg.tuwien.ac.at/research/rendering/rays-radio/



raytracing.github.io/books/RayTracingInOneWeekend.html

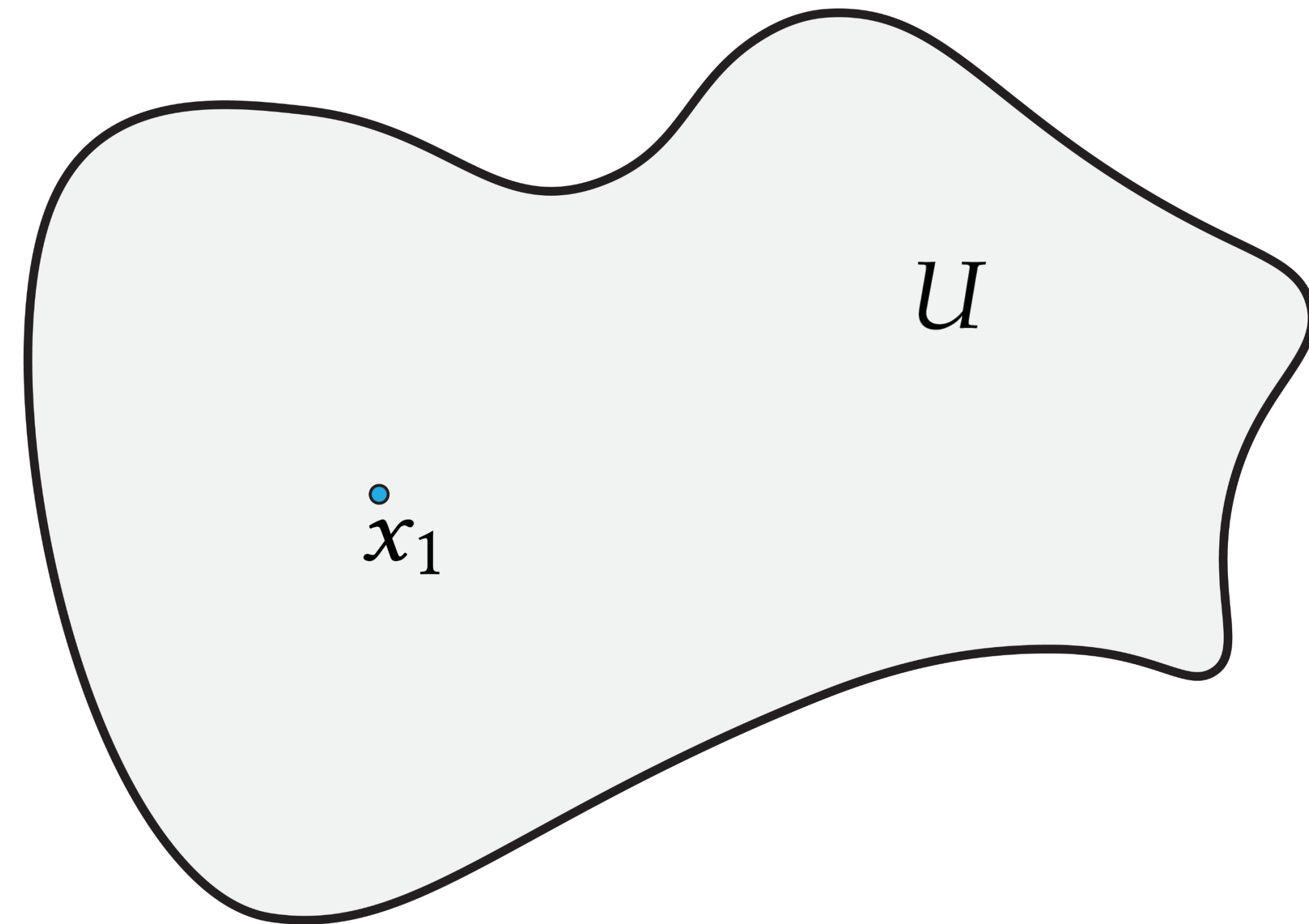
Walk on Spheres (WoS)

- Mervin E. Muller (*Ann. Math. Statist. 1956*)
“Some continuous monte carlo methods for the Dirichlet problem”
- Sawhney, R. and Crane, K. (*SIGGRAPH 2020*)
“Monte Carlo geometry processing”
- Sawhney, R. Seyb, D. Jarosz, W. and Crane, K (*SIGGRAPH 2020*)
“Grid free Monte Carlo for PDEs with spatially varying coefficients”

Walk on Spheres (WoS)

$$u(x) = g(x) \quad \text{if } x \in \partial U$$

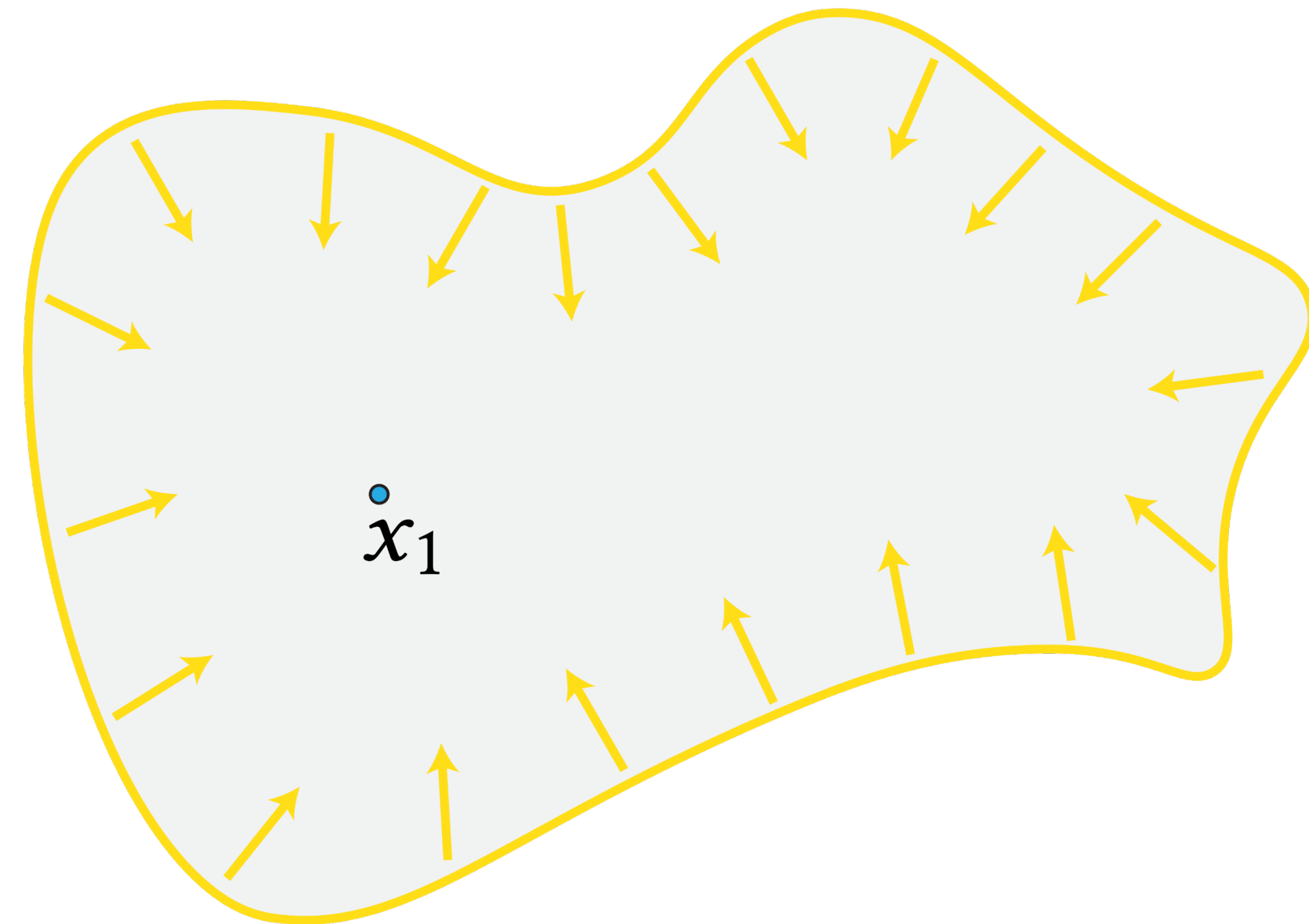
How to solve $u(x)$ inside U ?



Walk on Spheres (WoS)

$$u(x) = g(x) \quad \text{if } x \in \partial U$$

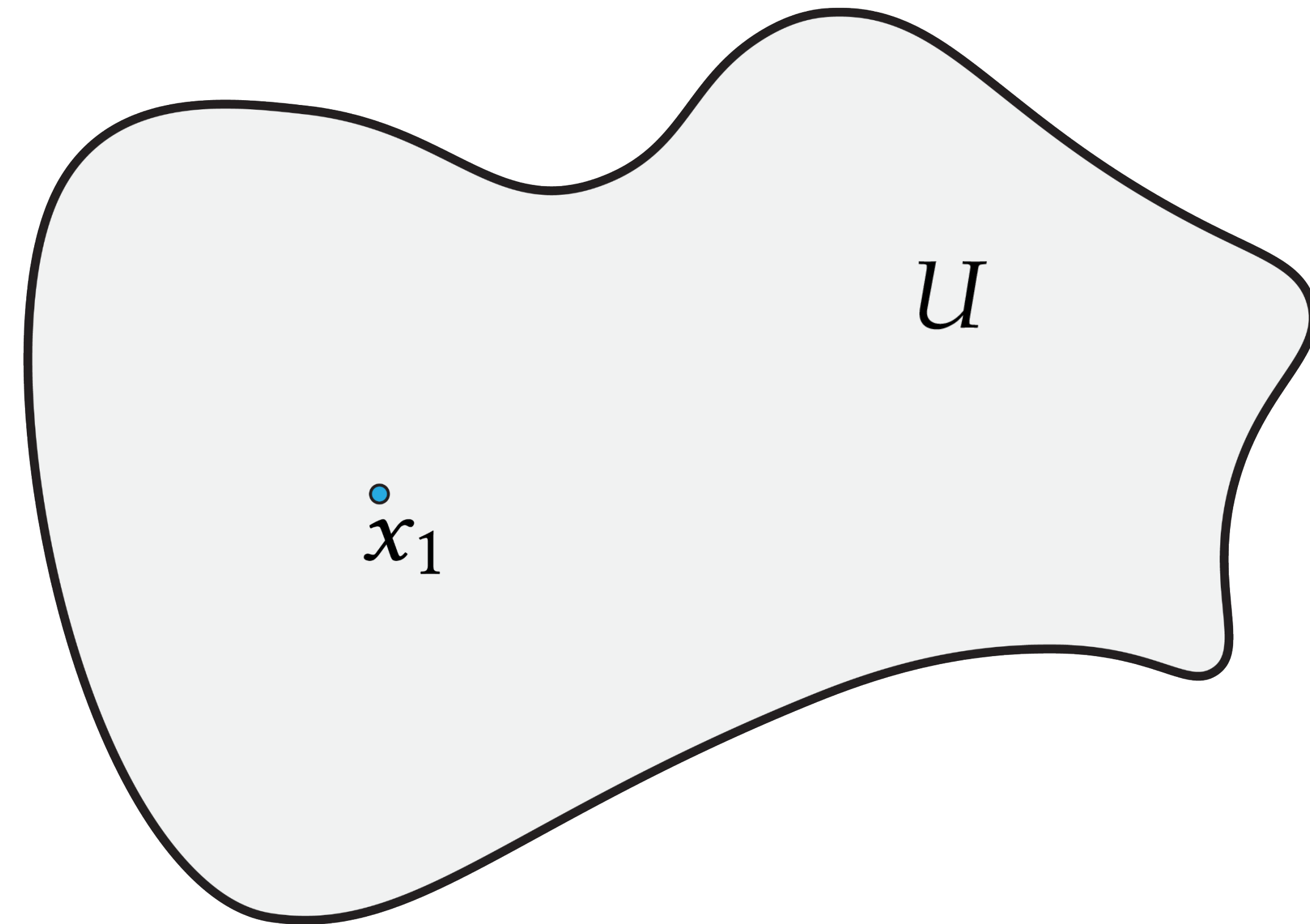
How to solve $u(x)$ inside U ?



Walk on Spheres (WoS)

Mean value theorem:

$$u(x) = \frac{1}{|\partial B_x|} \int_{\partial B_x} u(x') dx'$$



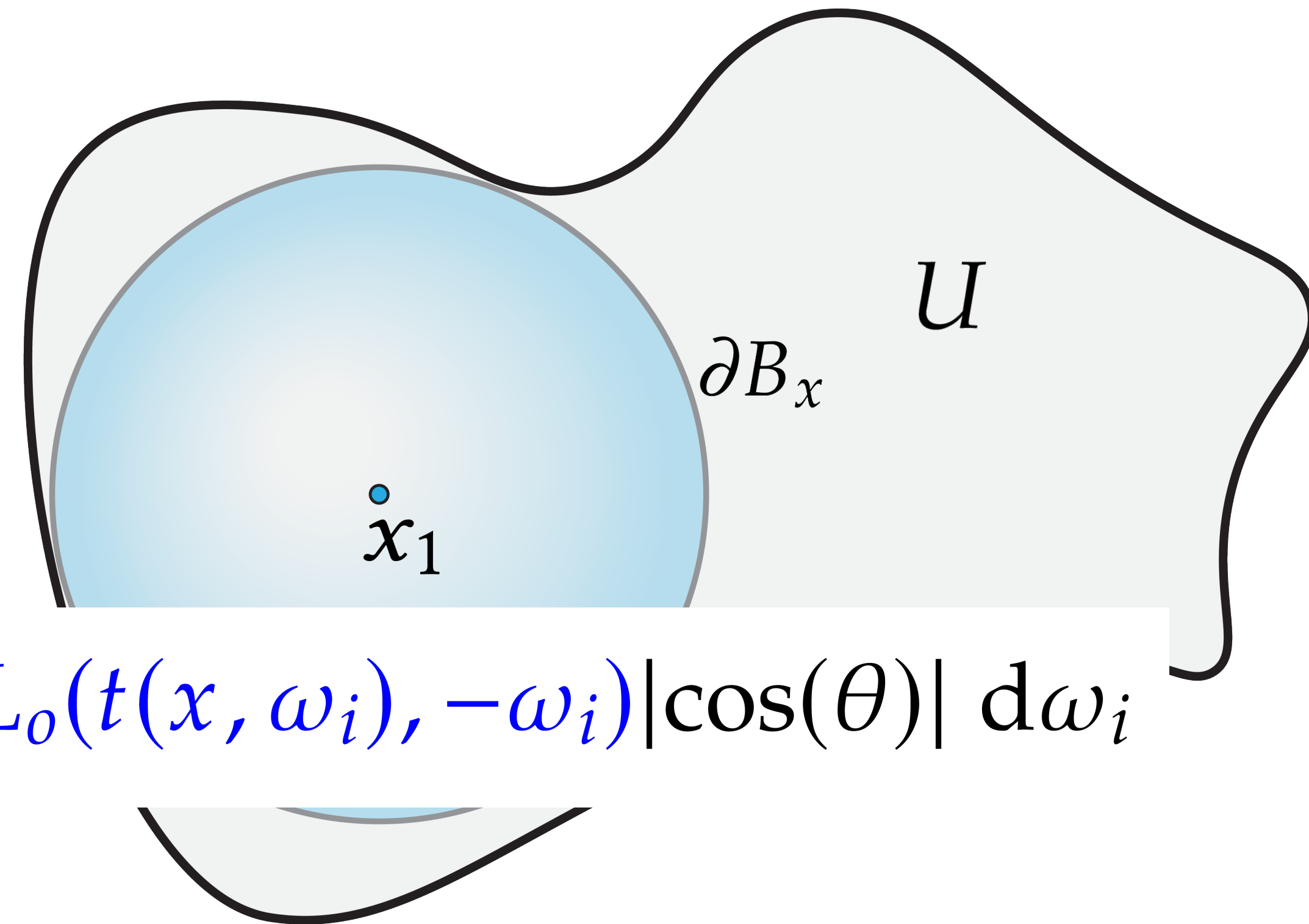
Walk on Spheres (WoS)

Mean value theorem:

$$u(x) = \frac{1}{|\partial B_x|} \int_{\partial B_x} u(x') dx'$$

Rendering Equation[Kaj86]:

$$L_o(x, \omega_o) = L_e(x, \omega_o) + \int_{S^2} f(x, \omega_o, \omega_i) L_o(t(x, \omega_i), -\omega_i) |\cos(\theta)| d\omega_i$$



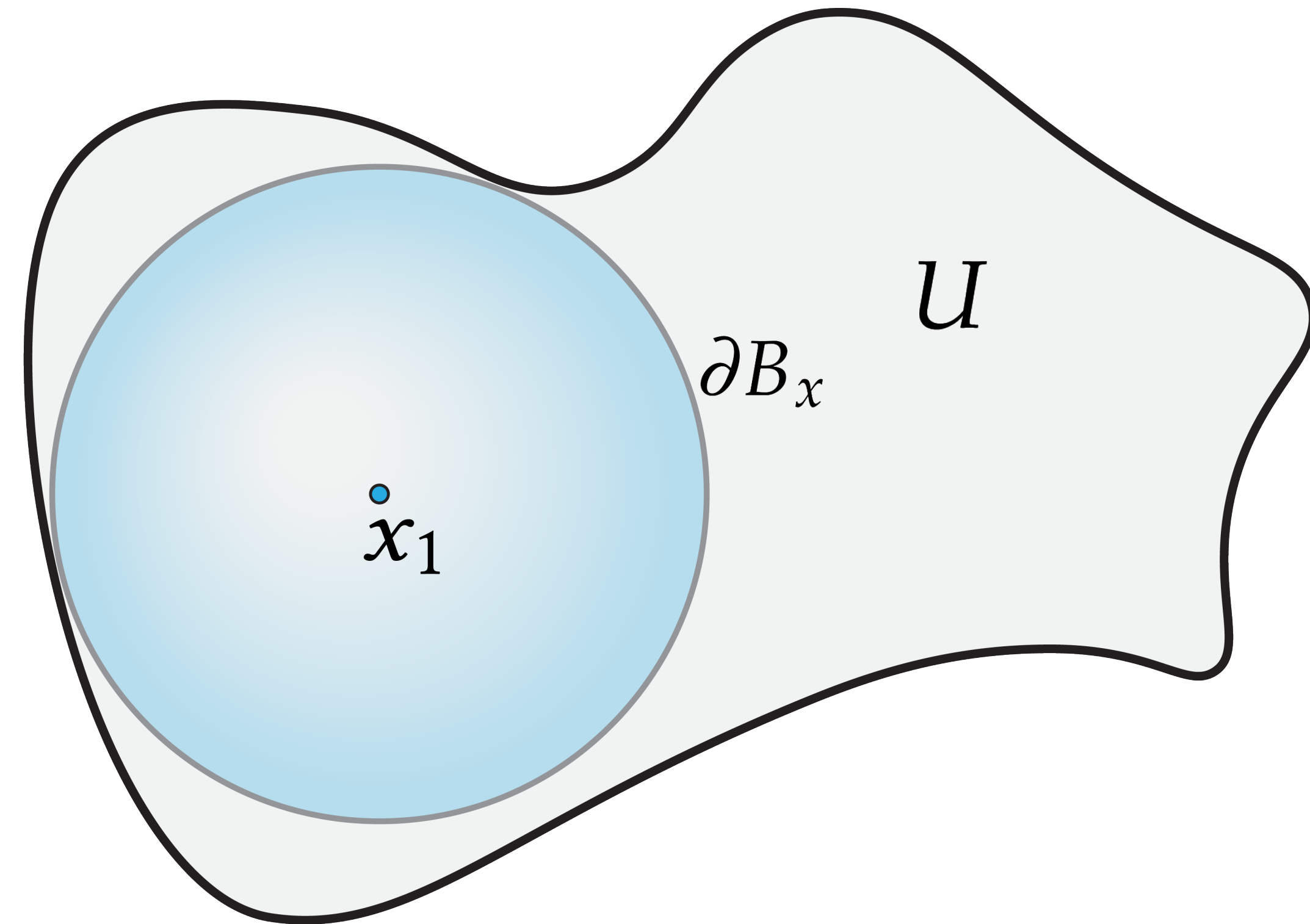
Walk on Spheres (WoS)

Mean value theorem:

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Estimator:

$$\langle u(x_i) \rangle = u(x_{i+1})$$



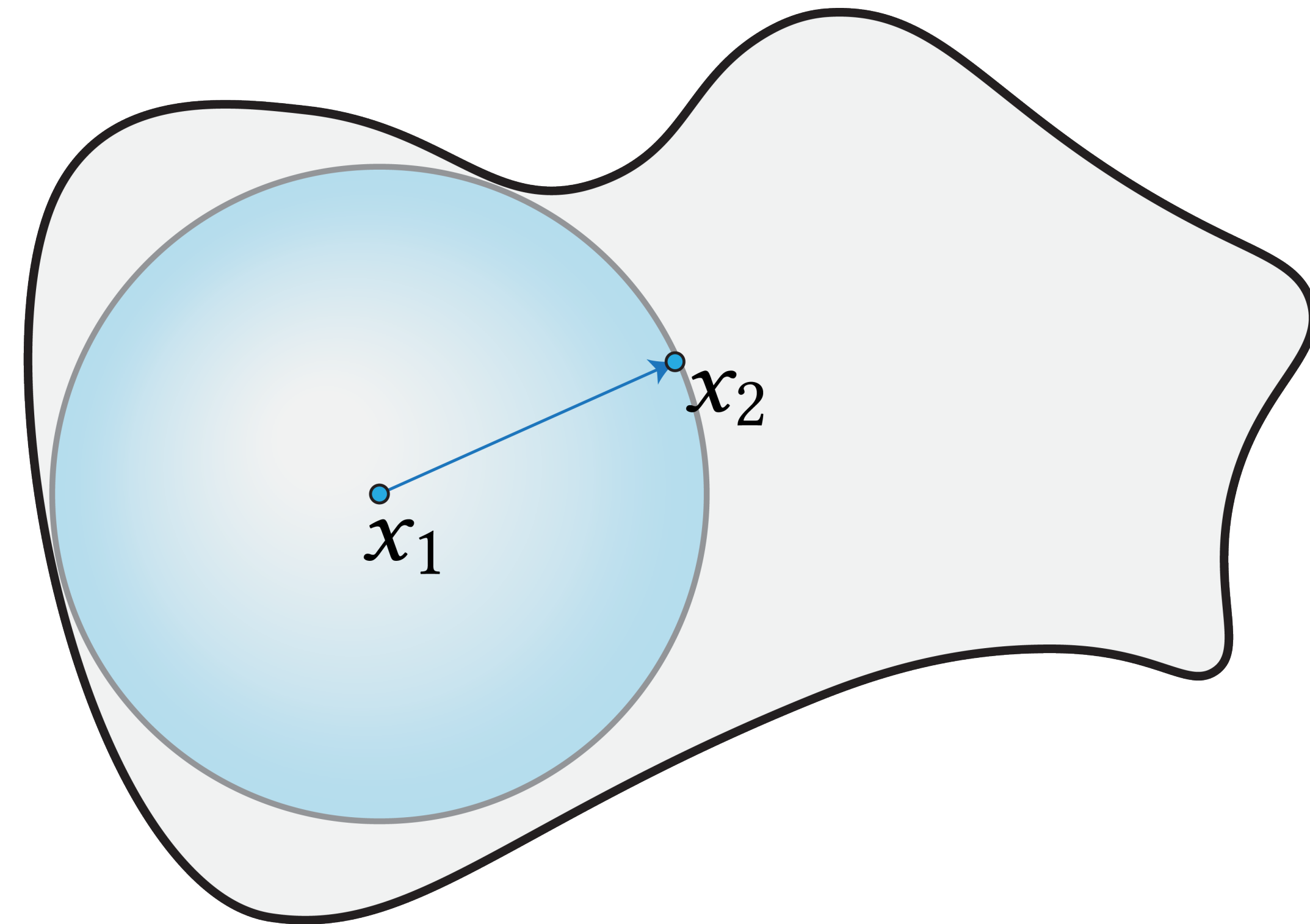
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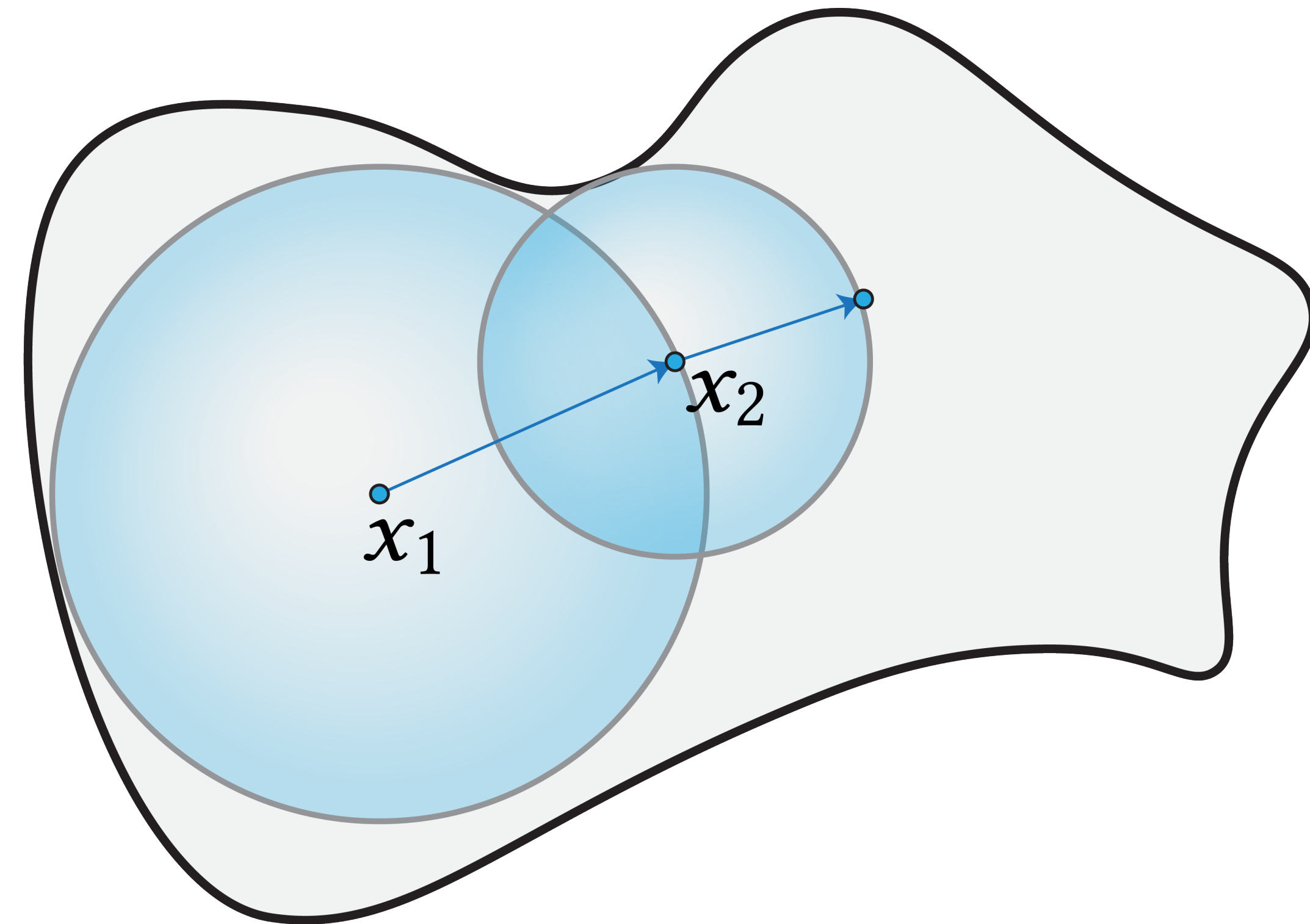
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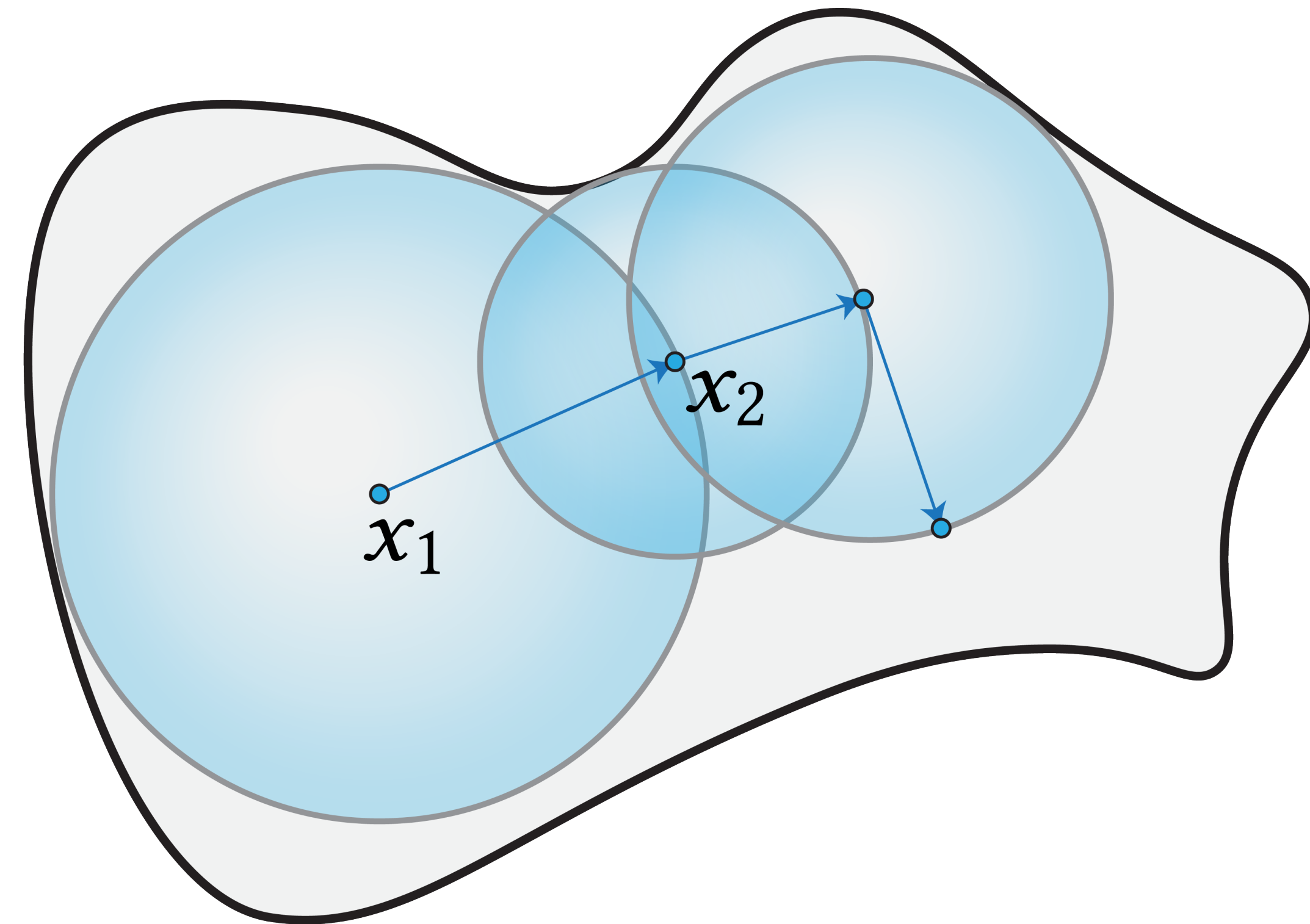
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Walk on Spheres (WoS)

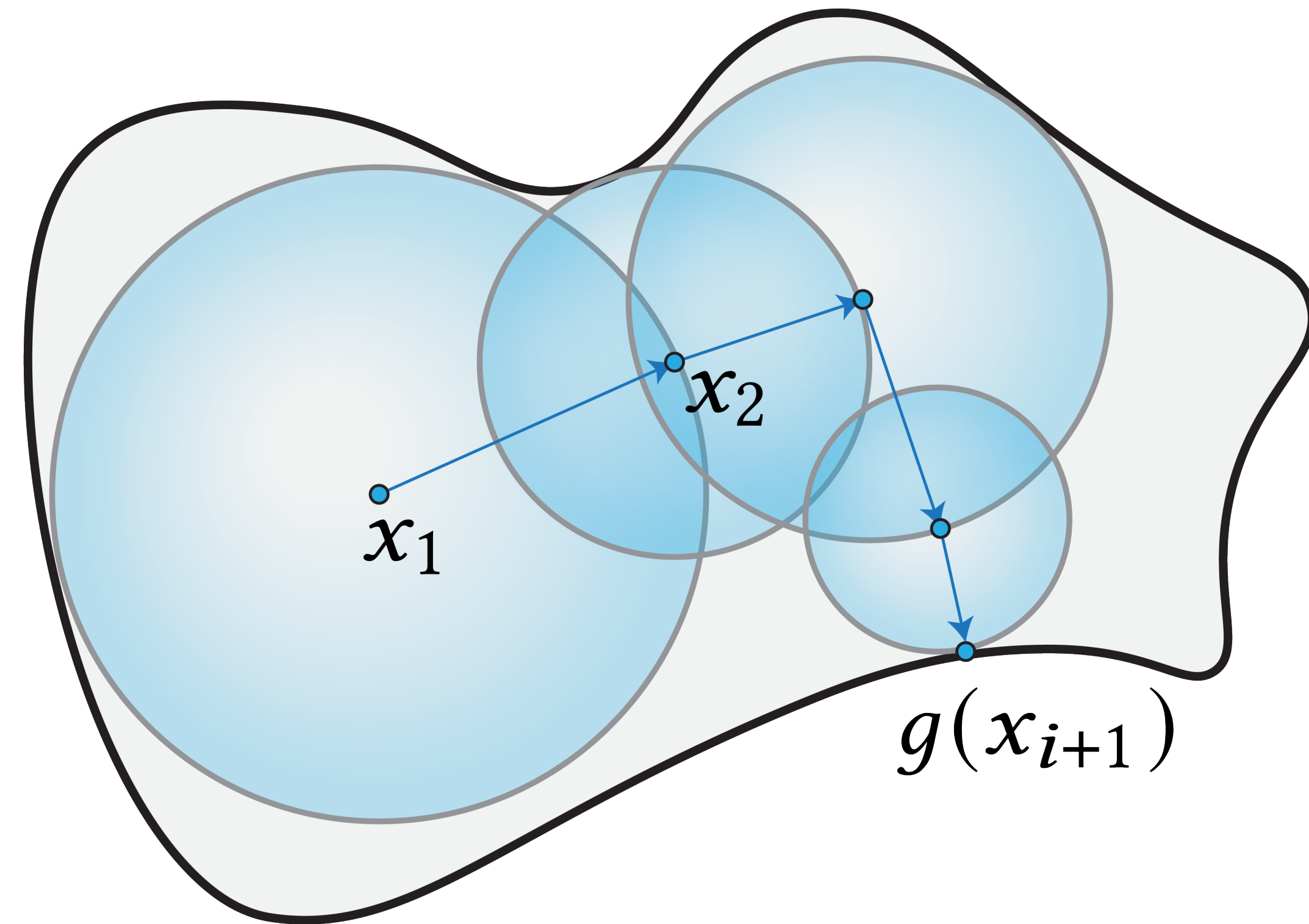
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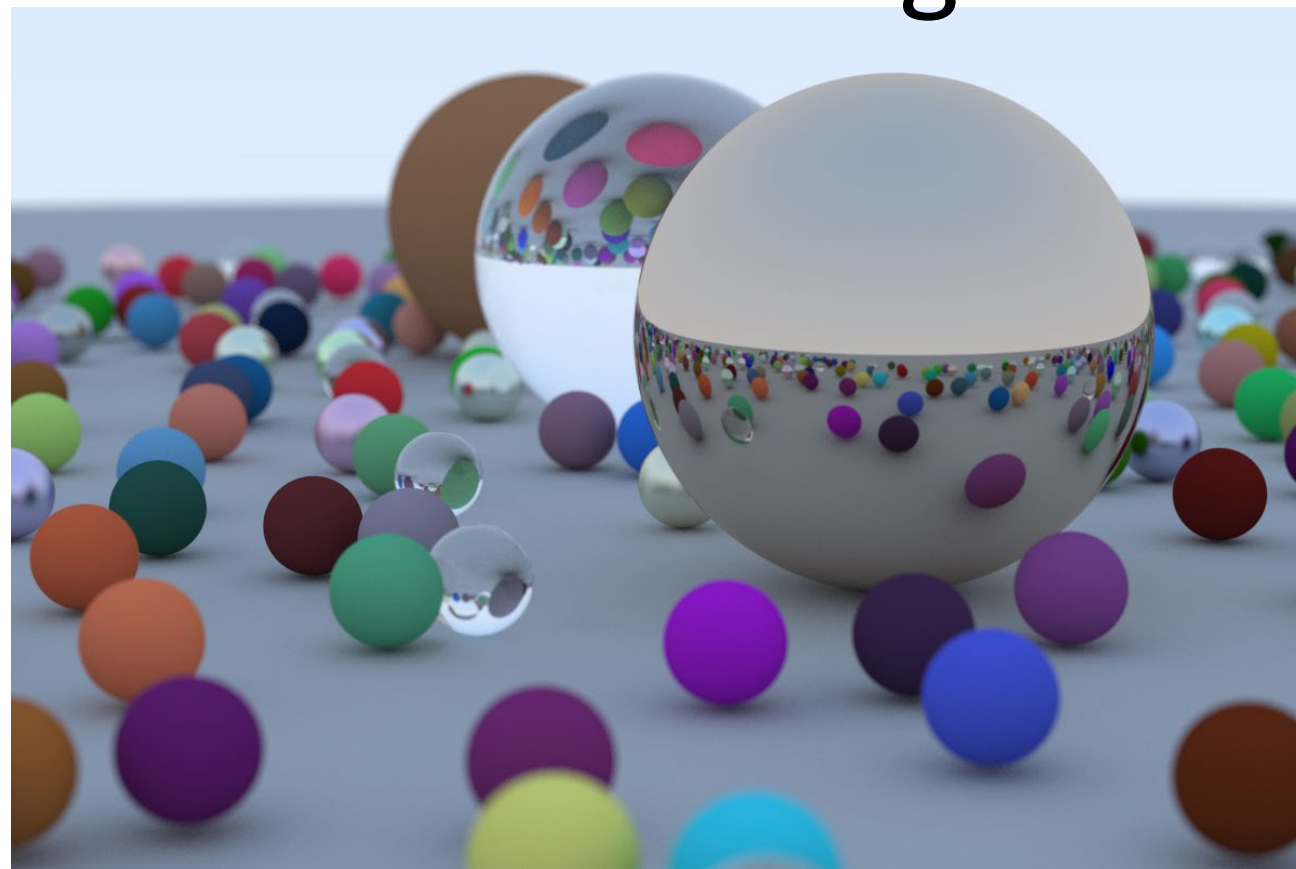
$$\langle u(x_i) \rangle = g(x_i) \quad \text{if } x \in \partial U$$



Rendering & PDE

“Forward” Method Path Tracing

Rendering:



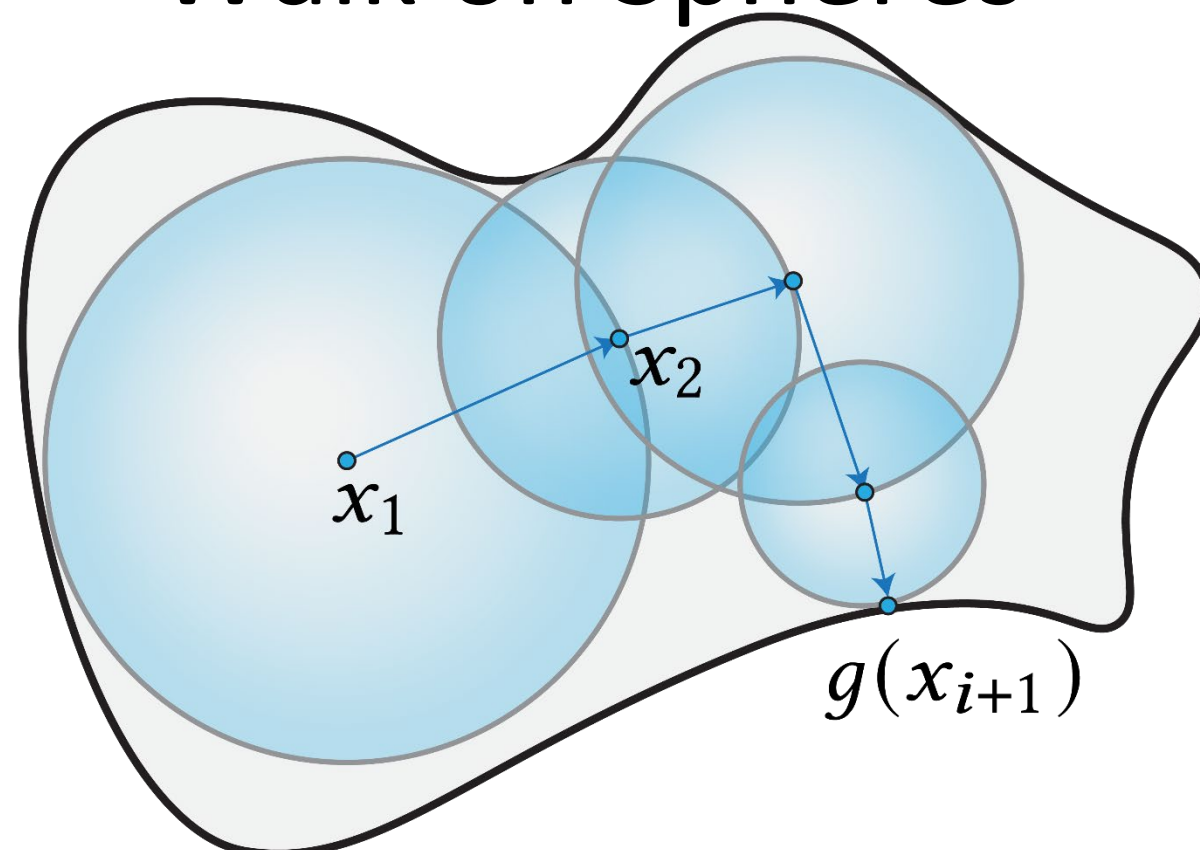
raytracing.github.io/books/RayTracingInOneWeekend.html

“Reverse” Method

1. No reuse of paths
2. No global importance sampling

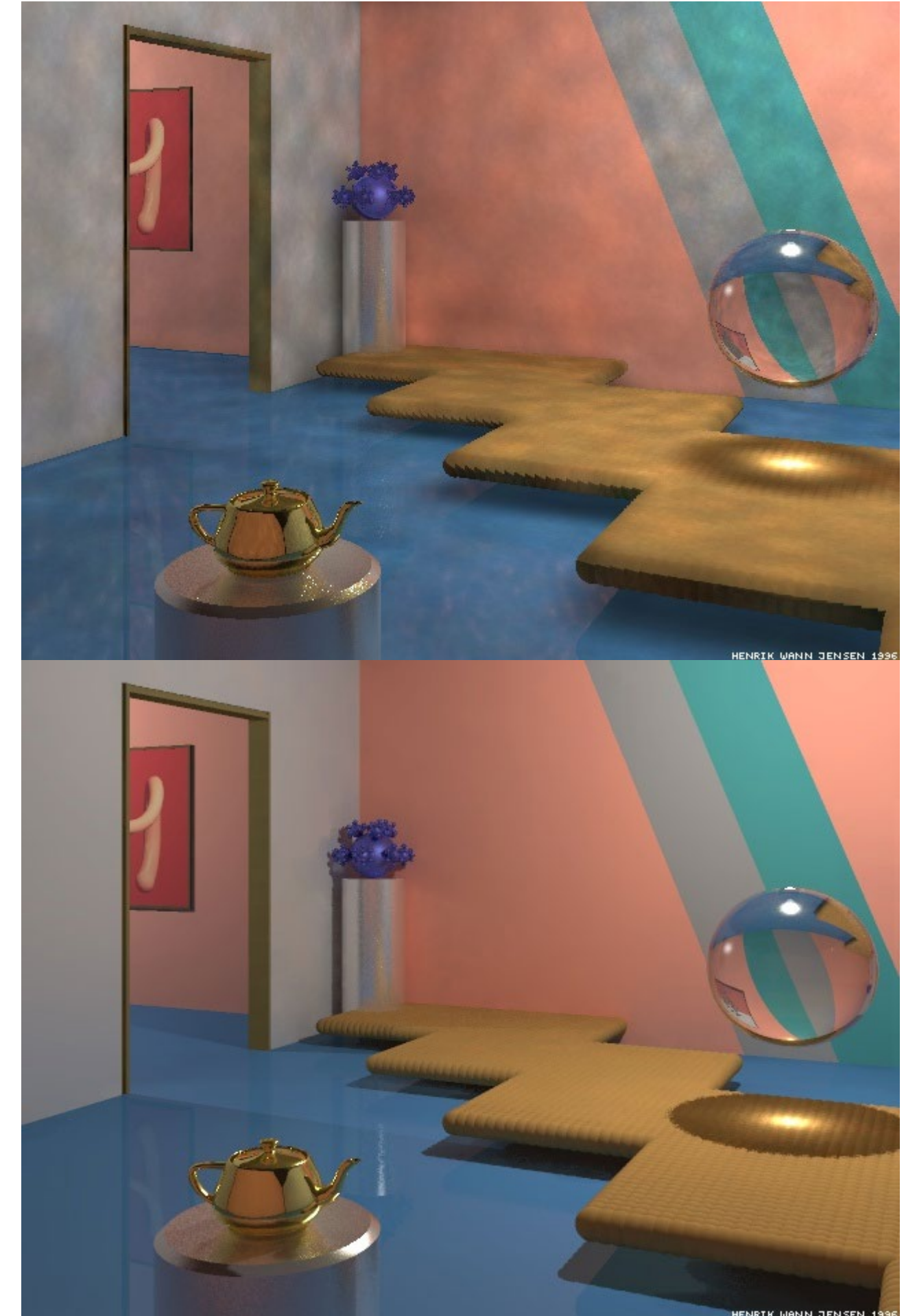
Walk on Spheres

PDE:



Photon mapping

- Henrik Wann Jensen
“*Global illumination using photon maps*”
- Henrik Wann Jensen
“*Realistic Image Synthesis Using Photon Mapping*”
- ...



http://graphics.ucsd.edu/~henrik/papers/photon_map/

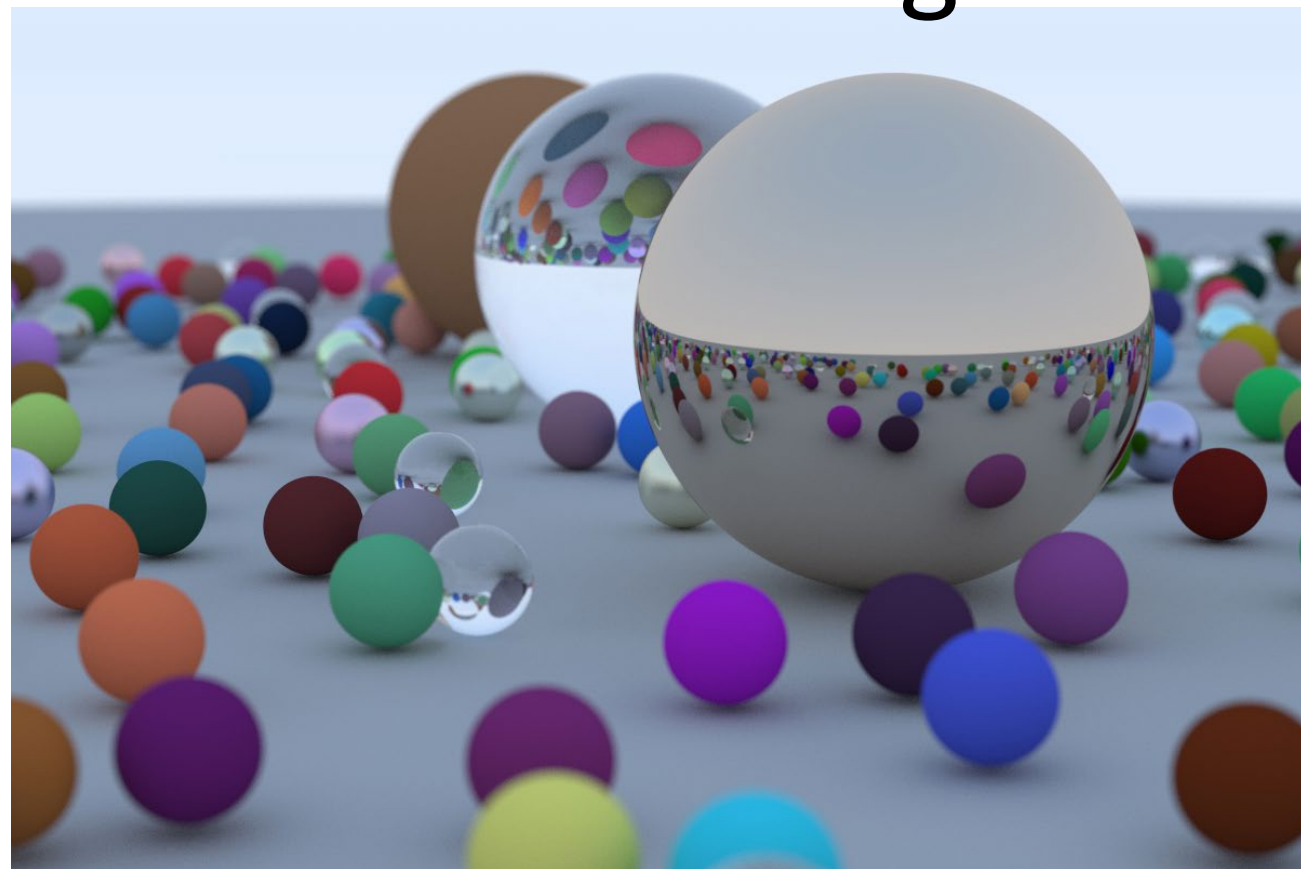
VPL / Many-light rendering

- Keller, A.
“Instant Radiosity” (SIGGRAPH 1997)
- Walter, B. Fernadandez, S. Arbree, A. Bala, K. Donikian, M. and Greenberg, D. P.
“Lightcuts: a scalable approach to illumination” (SIGGRAPH 2005)
- Walter, B., Arbree, A., Bala, K., and GREENBERG, D. P.
“Multidimensional lightcuts” (SIGGRAPH 2006)
- Hašan, M., Pellacini, F., and Bala, K.
“Matrix row-column sampling for the many-light problem” (SIGGRAPH 2007)
- ...

Rendering & PDE

“Forward” Method Path Tracing

Rendering:



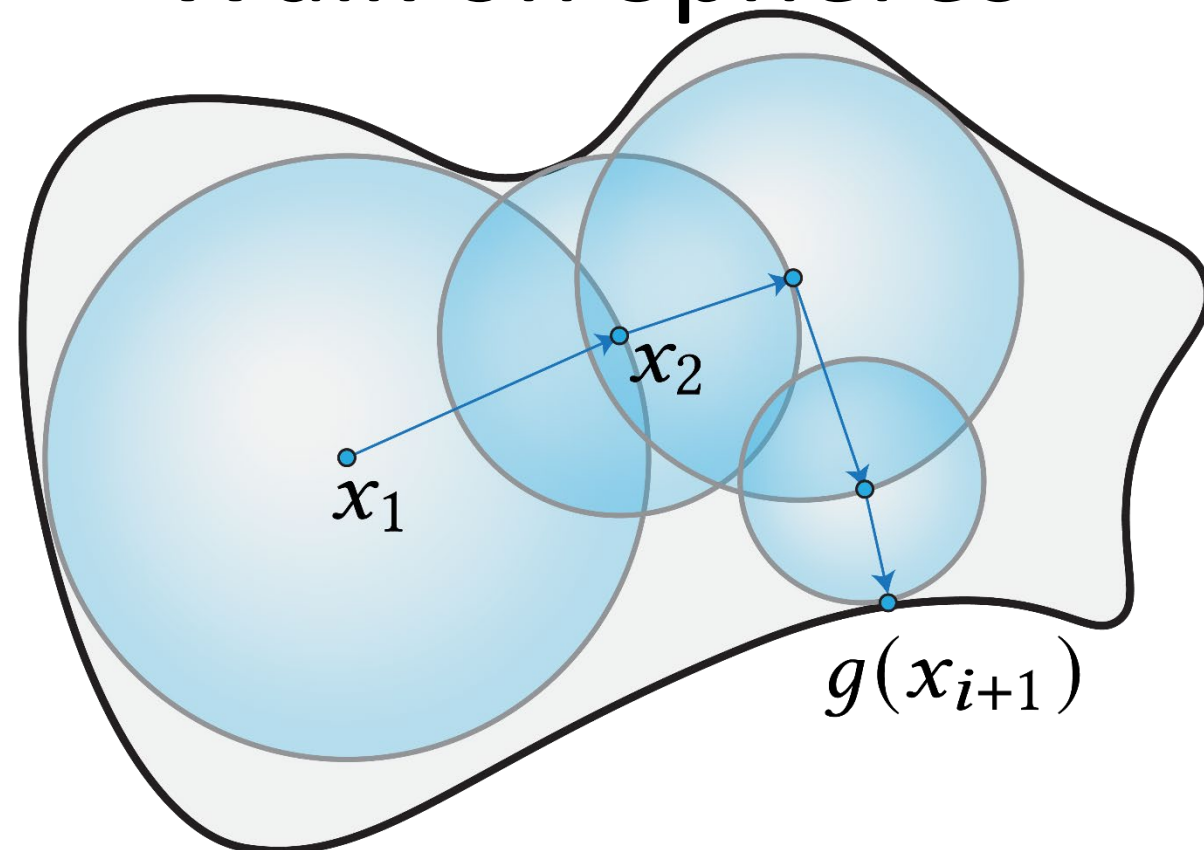
raytracing.github.io/books/RayTracingInOneWeekend.html

“Reverse” Method Photon Mapping / VPLs



graphics.ucsd.edu/~henrik/papers/photon_map/

Walk on Spheres



“Reverse WoS”?



PDE:

PDE

Laplace operator:

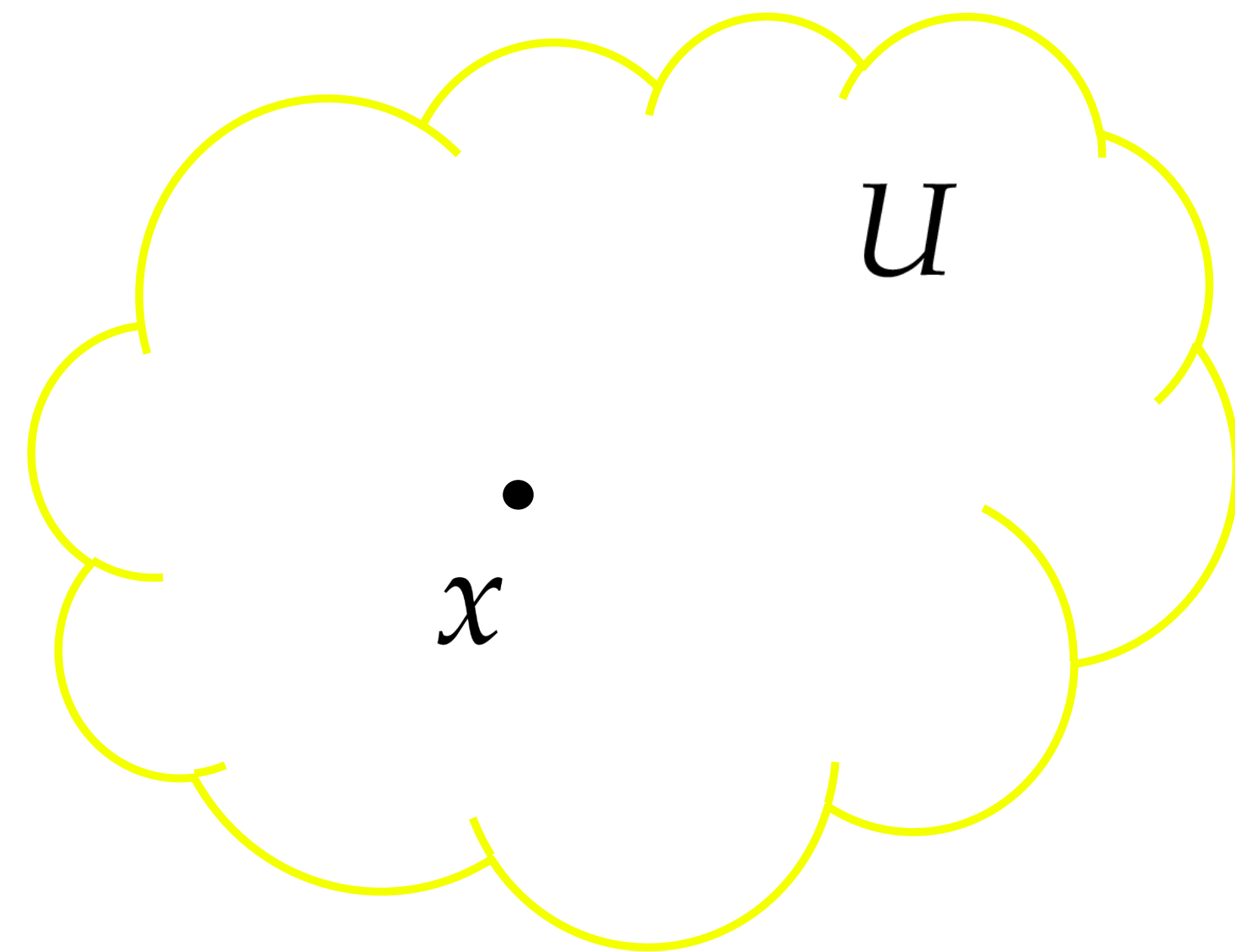
Sum of second derivatives

$$\Delta u(x, y) = \frac{\partial^2 u}{\partial x^2}(x, y) + \frac{\partial^2 u}{\partial y^2}(x, y)$$

PDE

Poisson's equation:

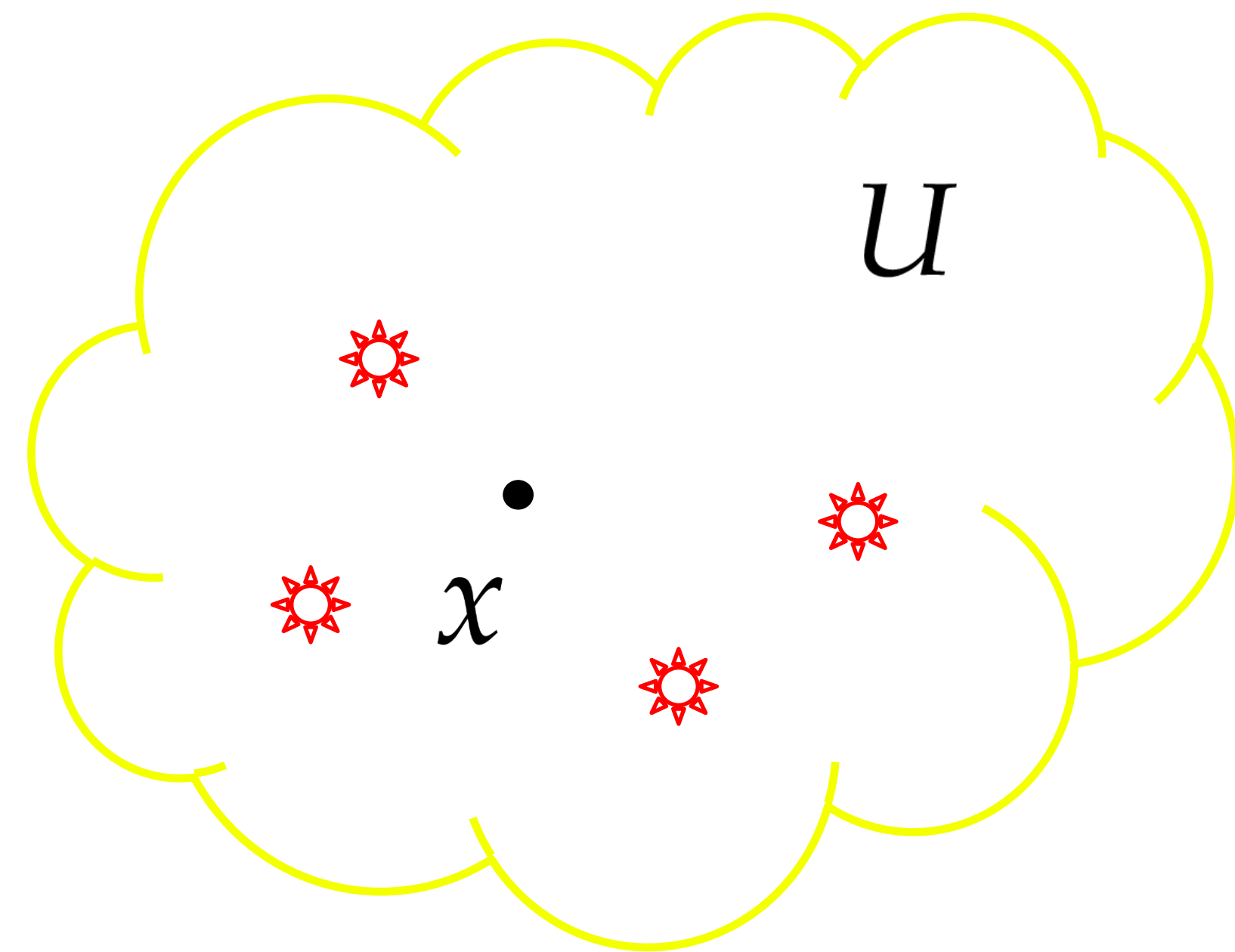
$$\begin{aligned}\Delta u(x) &= 0 && \text{if } x \in U, \\ u(x) &= g(x) && \text{if } x \in \partial U.\end{aligned}$$



PDE

Poisson's equation (with sources):

$$\begin{aligned}\Delta u(x) &= f(x) && \text{if } x \in U, \\ u(x) &= g(x) && \text{if } x \in \partial U.\end{aligned}$$



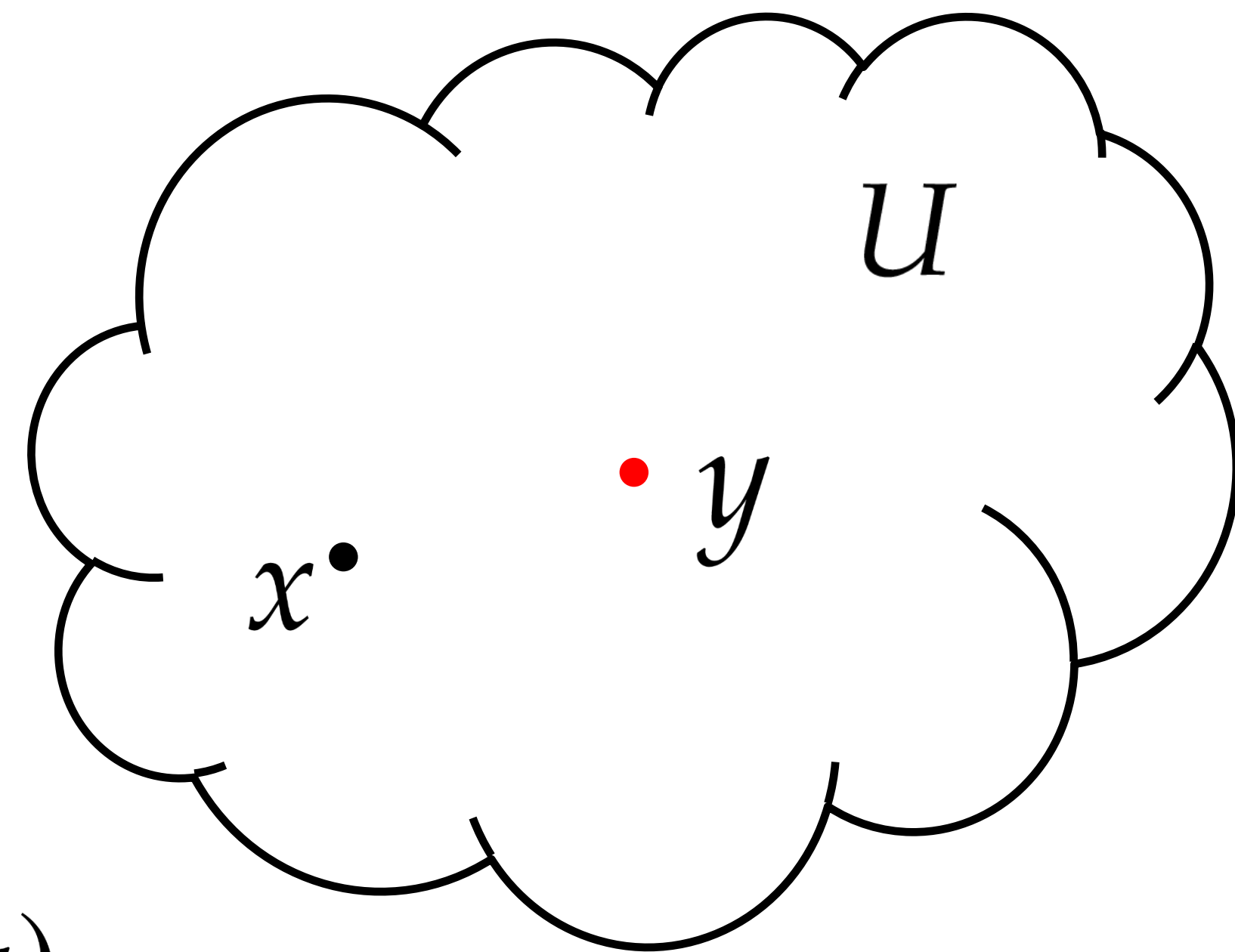
Green's Function

Delta-Point source:

$$\begin{aligned}\Delta u(x) &= \delta_y(x) & \text{if } x \in U \\ u(x) &= 0 & \text{if } x \in \partial U\end{aligned}$$

Green's function:

Solution to the delta point source PDE $\mathcal{G}(x, y)$



Green's Function

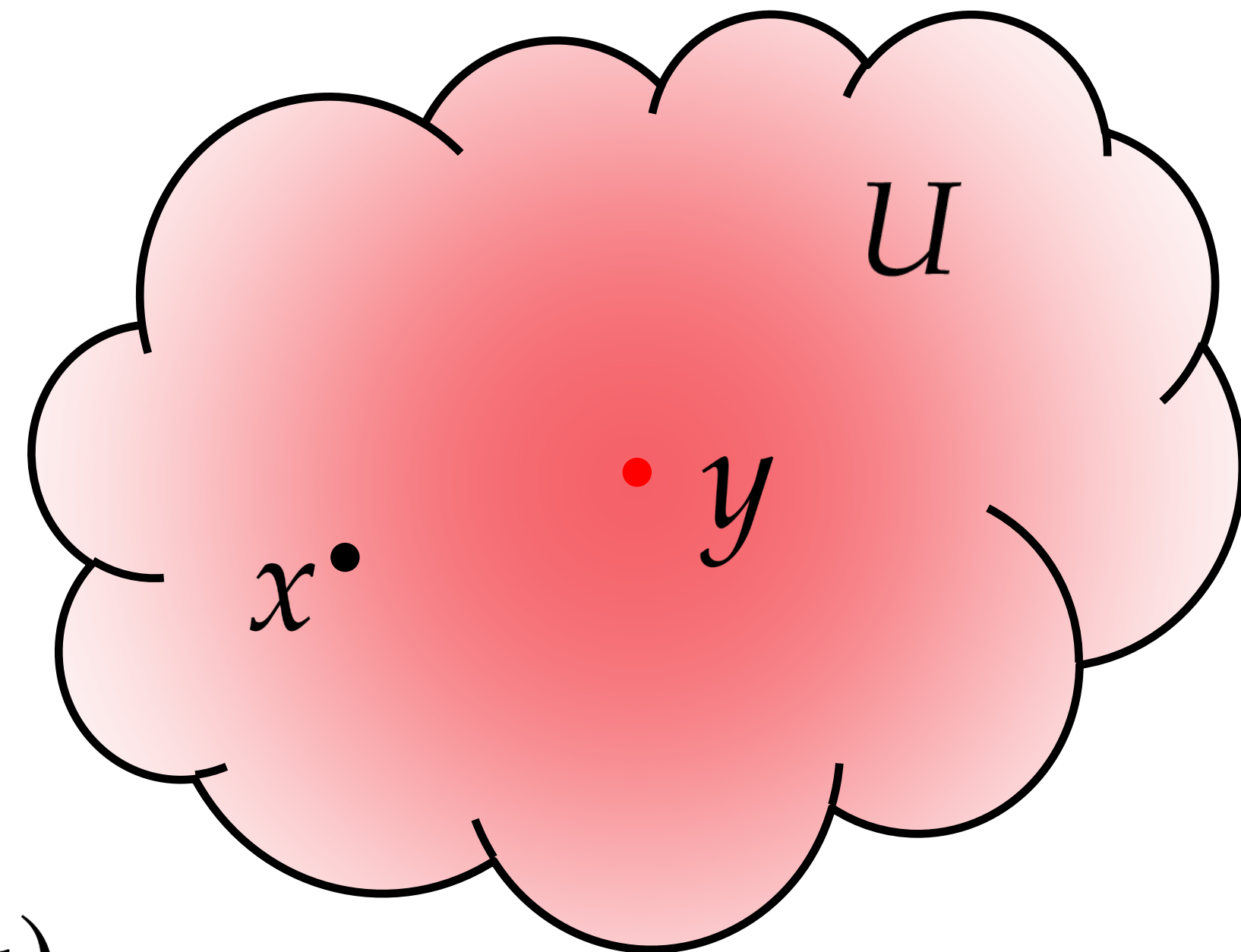
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Green's function:

Solution to the delta point source PDE $\mathcal{G}(x, y)$

Define $\mathcal{G}(x, y) = 0$ if x or y is not inside U



Green's Function

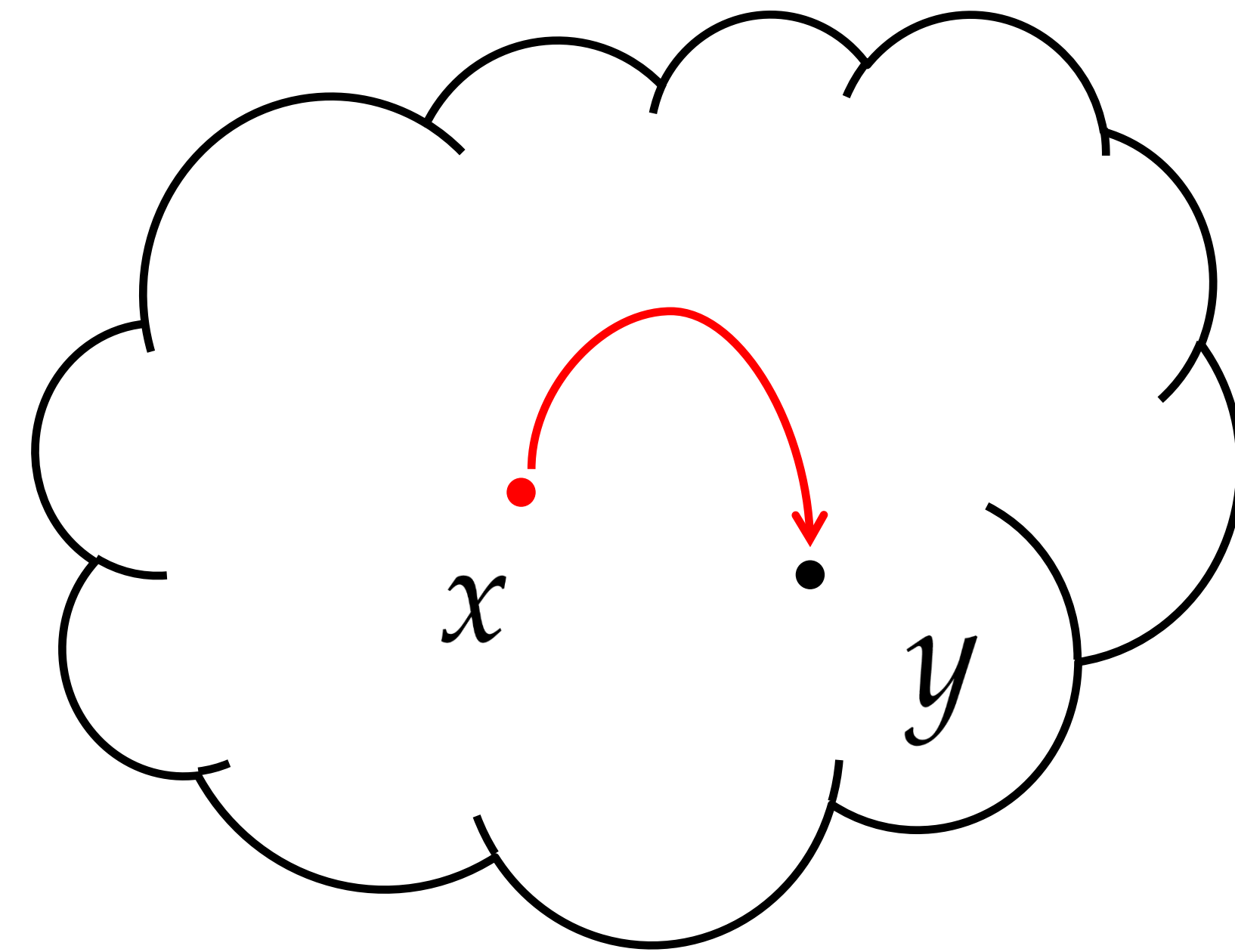
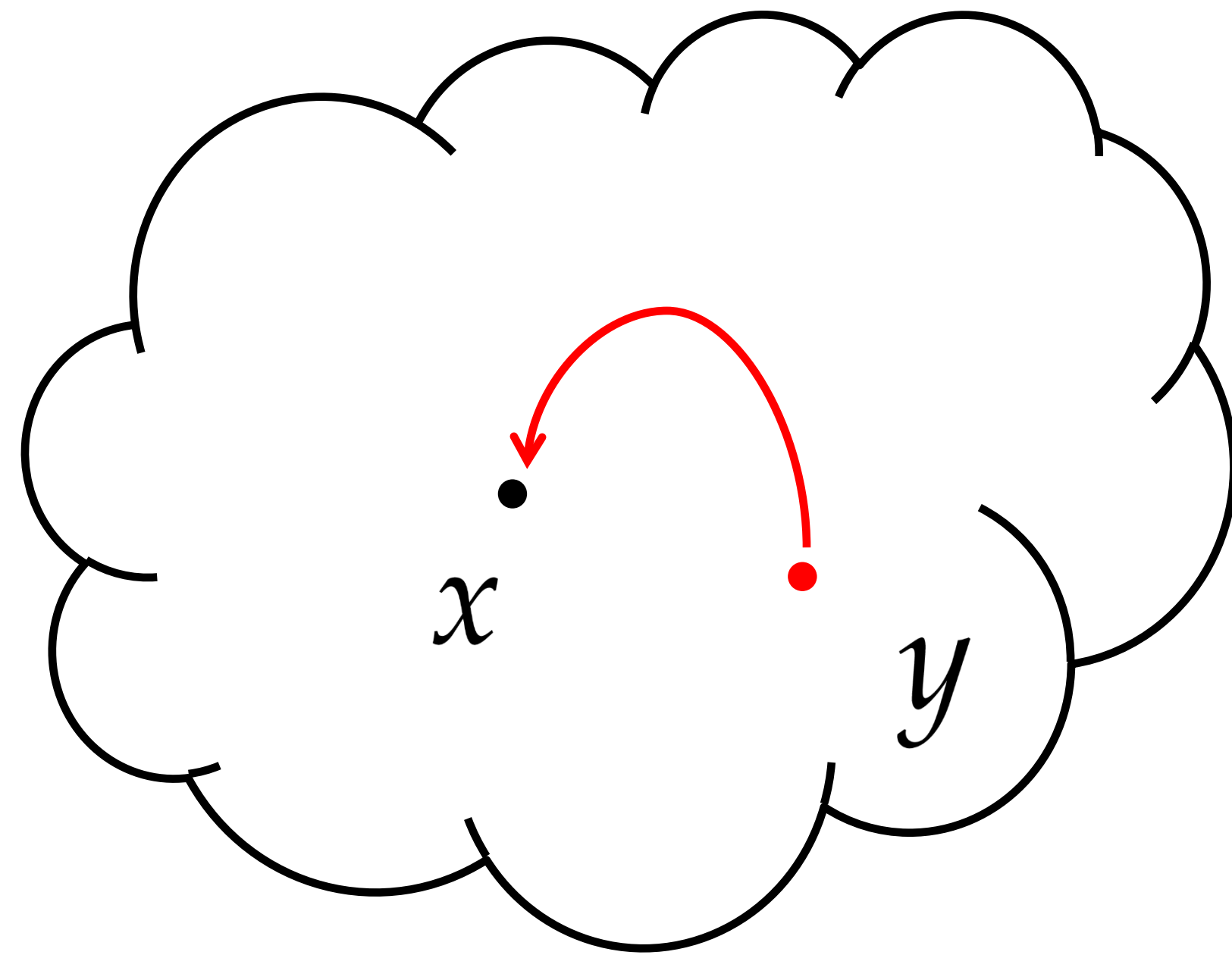
$$\mathcal{G}(x, y)$$

Green's function for the domain
(Usually hard to calculate)

$$\mathcal{G}^{B_x}(x, y)$$

Green's function for the ball
(Has a simple analytical form)

Symmetry property of Green's Function



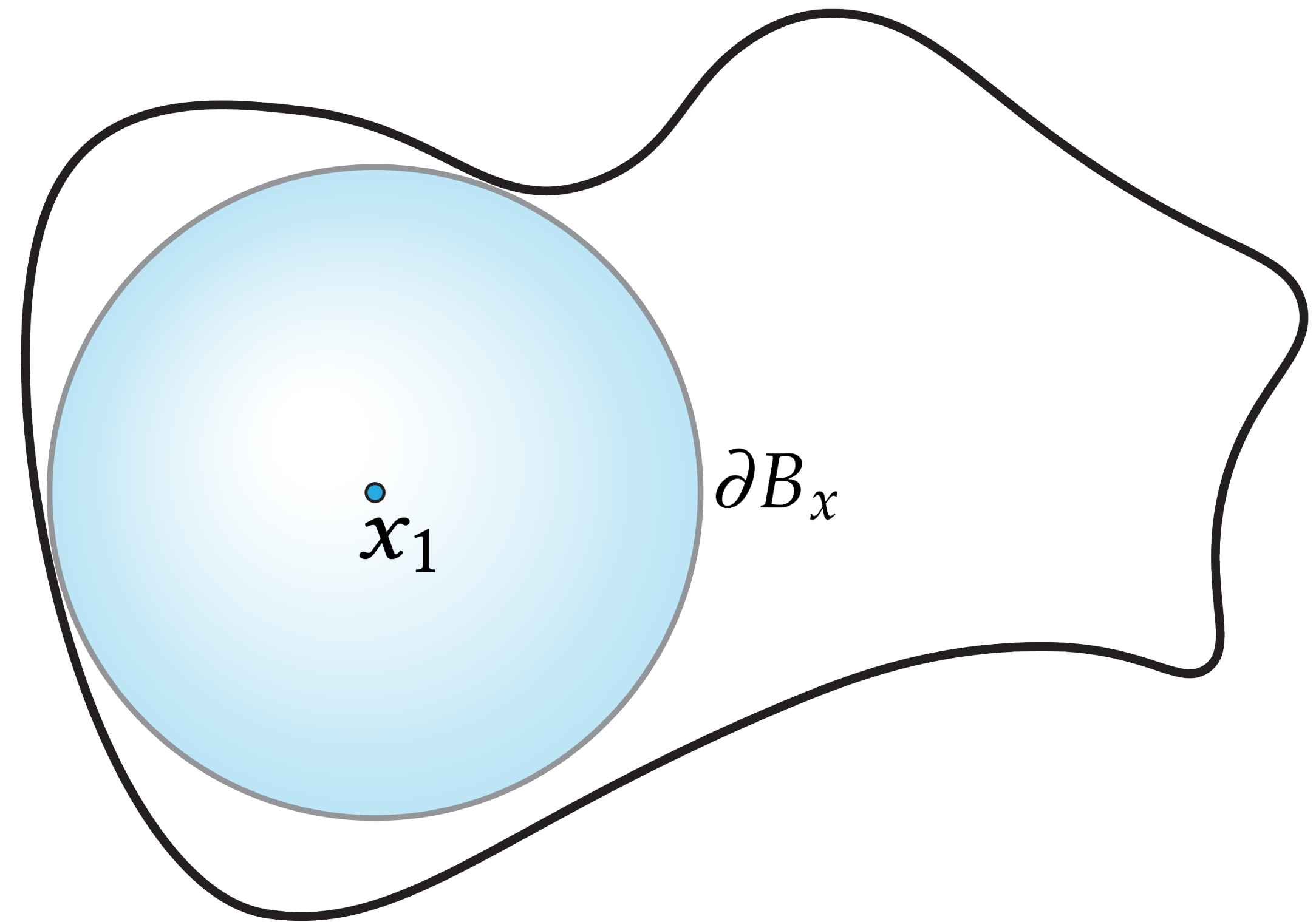
Use double side arrow to denote the symmetric property

$$\mathcal{G}(x \leftrightarrow y) := \mathcal{G}(x, y) = \mathcal{G}(y, x)$$

Walk on Spheres with Sources

Mean Value Theorem:

$$u(x) = \frac{1}{|\partial B_x|} \int_{\partial B_x} u(x') \, dx' + \int_{B_x} f(y) \mathcal{G}^{B_x}(x \leftrightarrow y) \, dy$$



Walk on Spheres with Sources

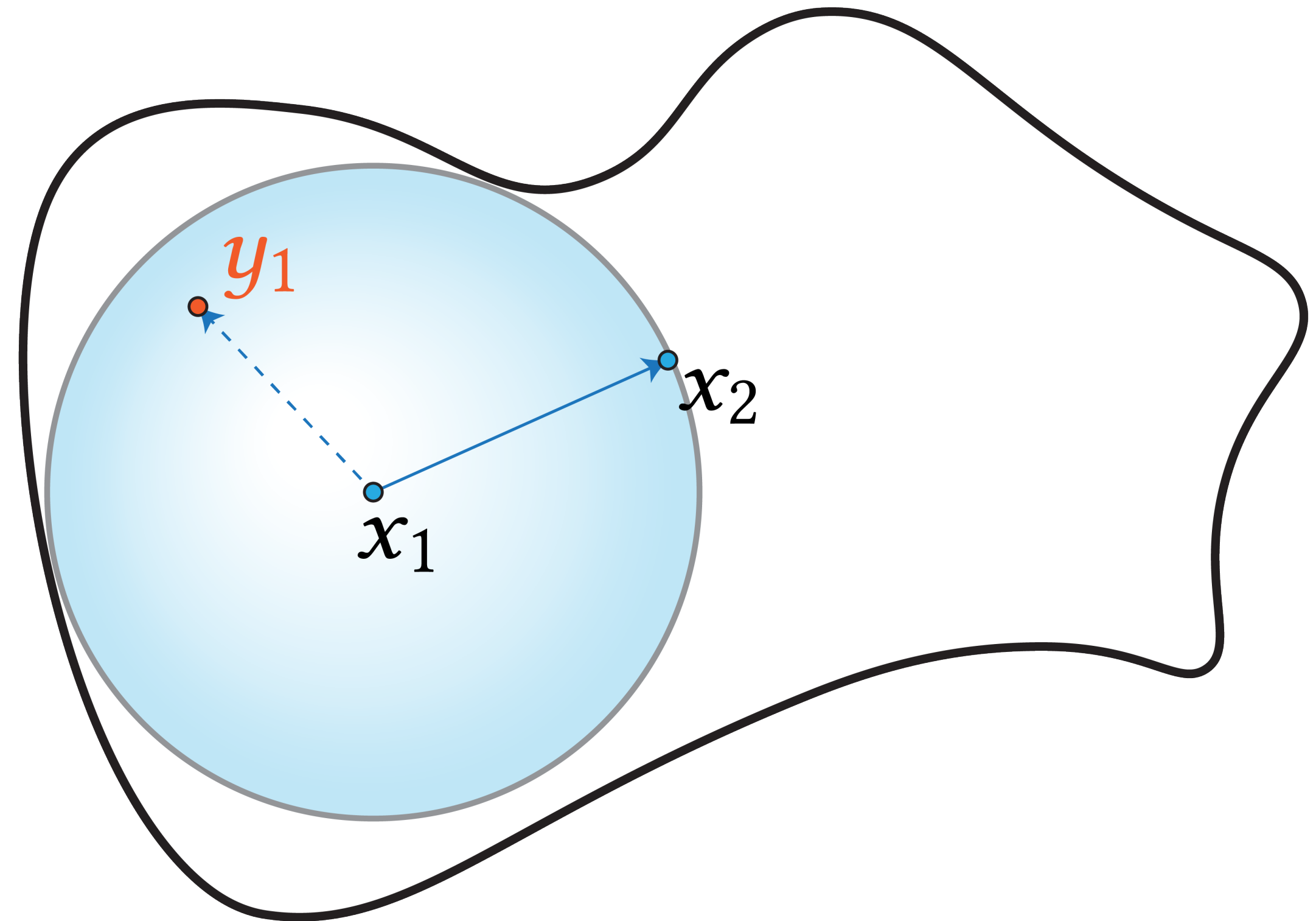
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Estimator:

$$\langle u(x_i) \rangle = u(x_{i+1}) + \underbrace{\frac{f(y_i) \mathcal{G}^{B_{x_i}}(x_i \leftrightarrow y_i)}{p(y_i)}}_{\text{Estimate contribution from sources}}$$

Estimate contribution from sources



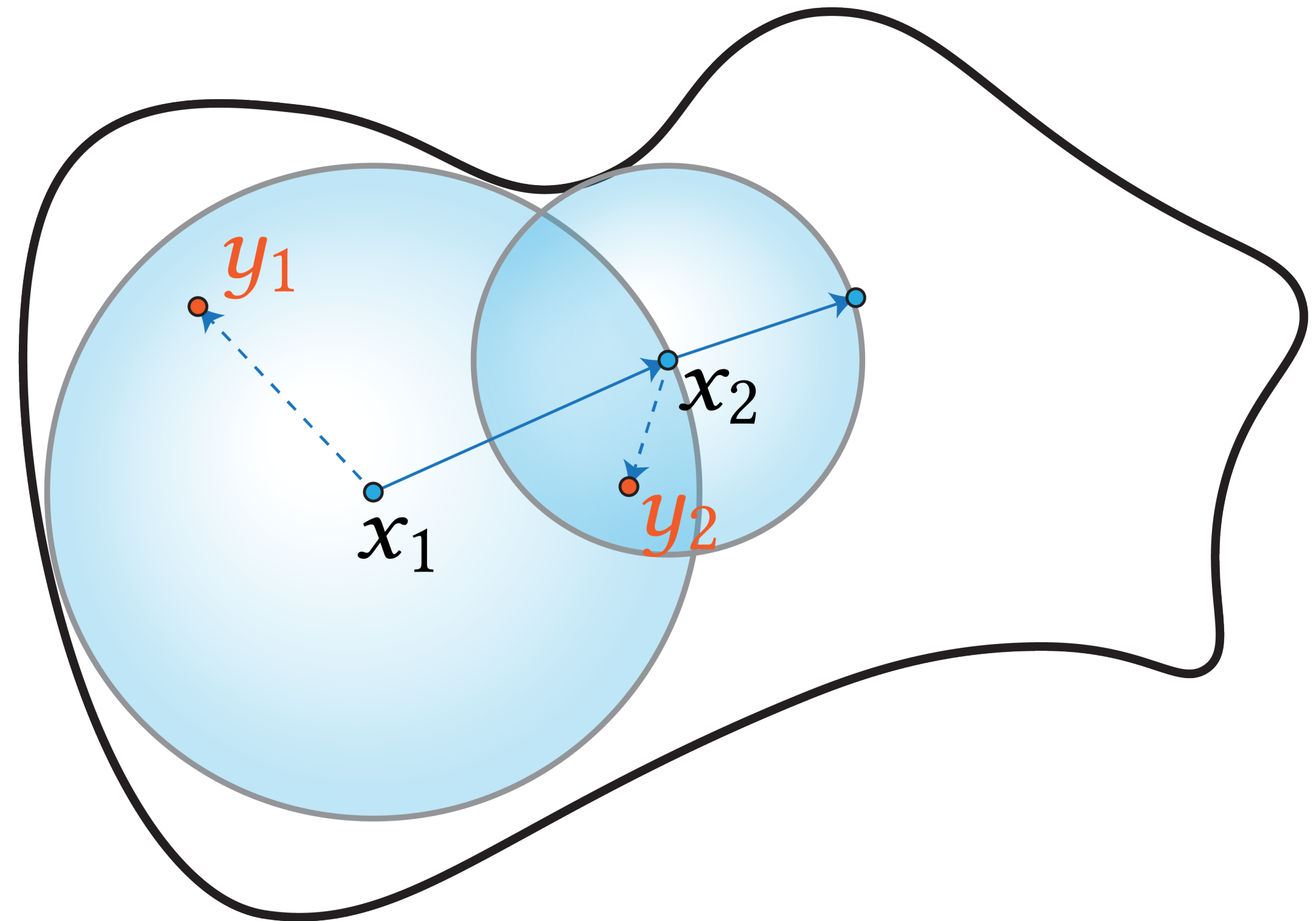
Walk on Spheres with Sources

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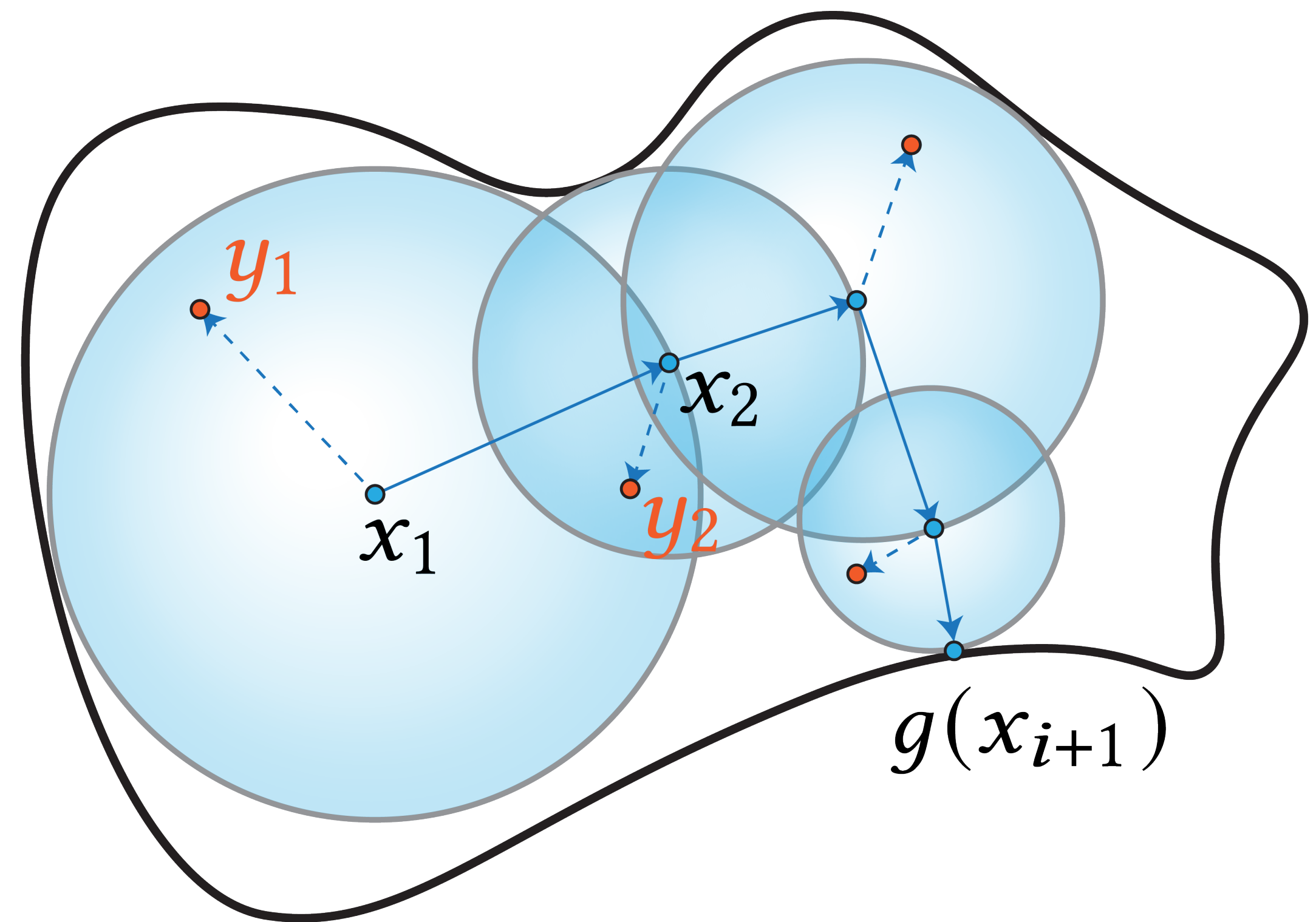
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$$\langle u(x_i) \rangle = g(x_i) \quad \text{if } x \in \partial U$$



Our Approach

Mean Value Theorem:

$$u(x) = \frac{1}{|\partial B_x|} \int_{\partial B_x} u(z) \, dz + \int_{B_x} f(y) \mathcal{G}^{B_x}(x \leftrightarrow y) \, dy$$

Representation formula using Green's function:

$$u(x) = \int_{\partial U} g(z) \frac{\partial \mathcal{G}(x \leftrightarrow z)}{\partial z} \, dz + \int_U f(y) \mathcal{G}(x \leftrightarrow y) \, dy$$

Need an estimate for the Green's function!

MVT for Green's function

Delta point source PDE + Mean value theorem

= MVT for Green's function:

$$\mathcal{G}(x \leftrightarrow y) = \mathcal{G}^{B_x}(x \leftrightarrow y) + \frac{1}{|\partial B_x|} \int_{\partial B_x} \mathcal{G}(x' \leftrightarrow y) \, dx' .$$

Swap x and y :

$$\mathcal{G}(x \leftrightarrow y) = \mathcal{G}^{B_y}(x \leftrightarrow y) + \frac{1}{|\partial B_y|} \int_{\partial B_y} \mathcal{G}(x \leftrightarrow y') \, dy' .$$

MVT for Green's function

MVT for x :

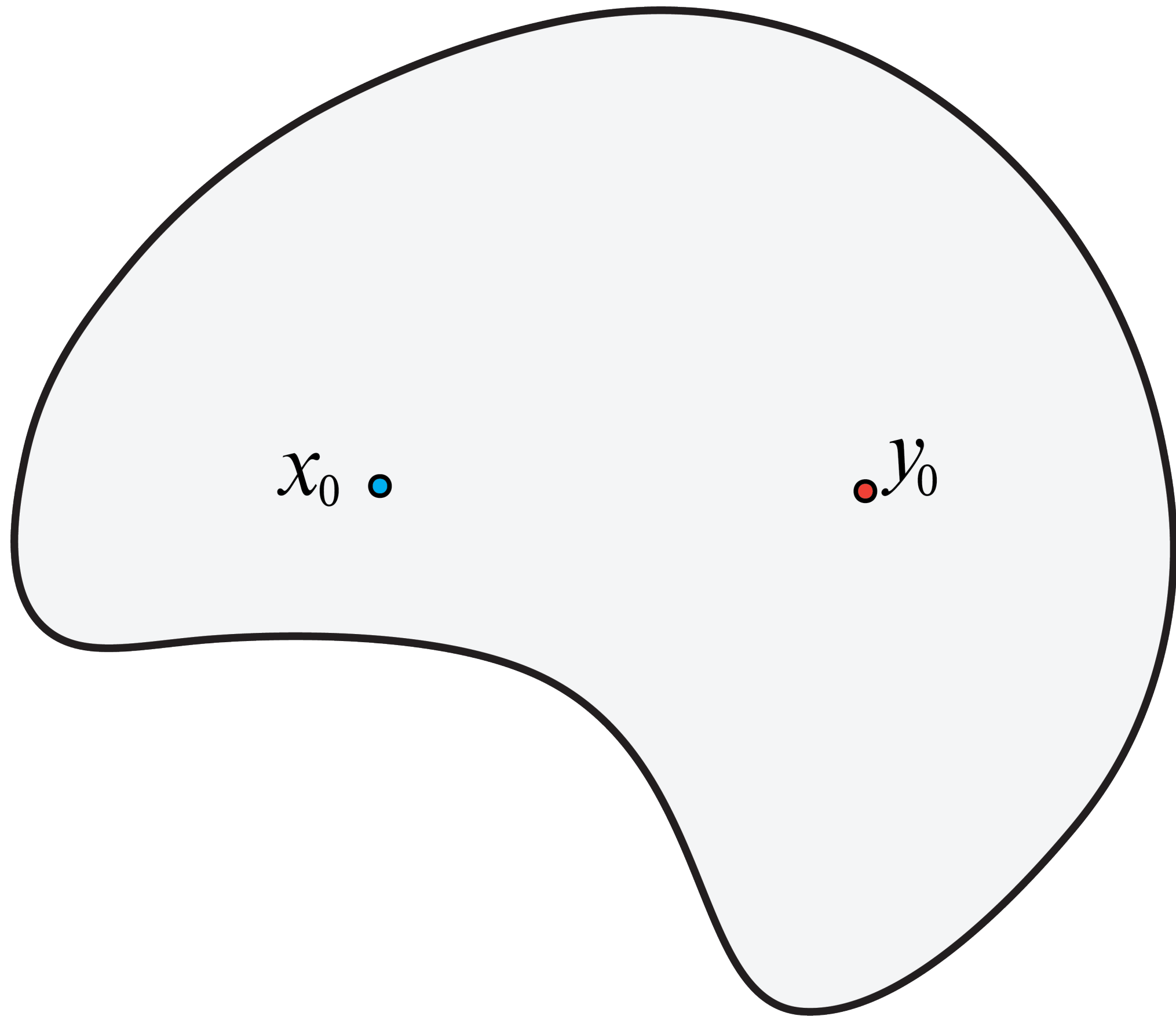
$$\mathcal{G}(x \leftrightarrow y) = \mathcal{G}^{B_x}(x \leftrightarrow y) + \frac{1}{|\partial B_x|} \int_{\partial B_x} \mathcal{G}(x' \leftrightarrow y) dx'$$

MVT for y :

$$\mathcal{G}(x \leftrightarrow y) = \mathcal{G}^{B_y}(x \leftrightarrow y) + \frac{1}{|\partial B_y|} \int_{\partial B_y} \mathcal{G}(x \leftrightarrow y') dy'$$

Forward Estimator

Forward: $\mathcal{G}(x \leftrightarrow y) = \mathcal{G}^{B_x}(x \leftrightarrow y) + \frac{1}{|\partial B_x|} \int_{\partial B_x} \mathcal{G}(x' \leftrightarrow y) dx' .$

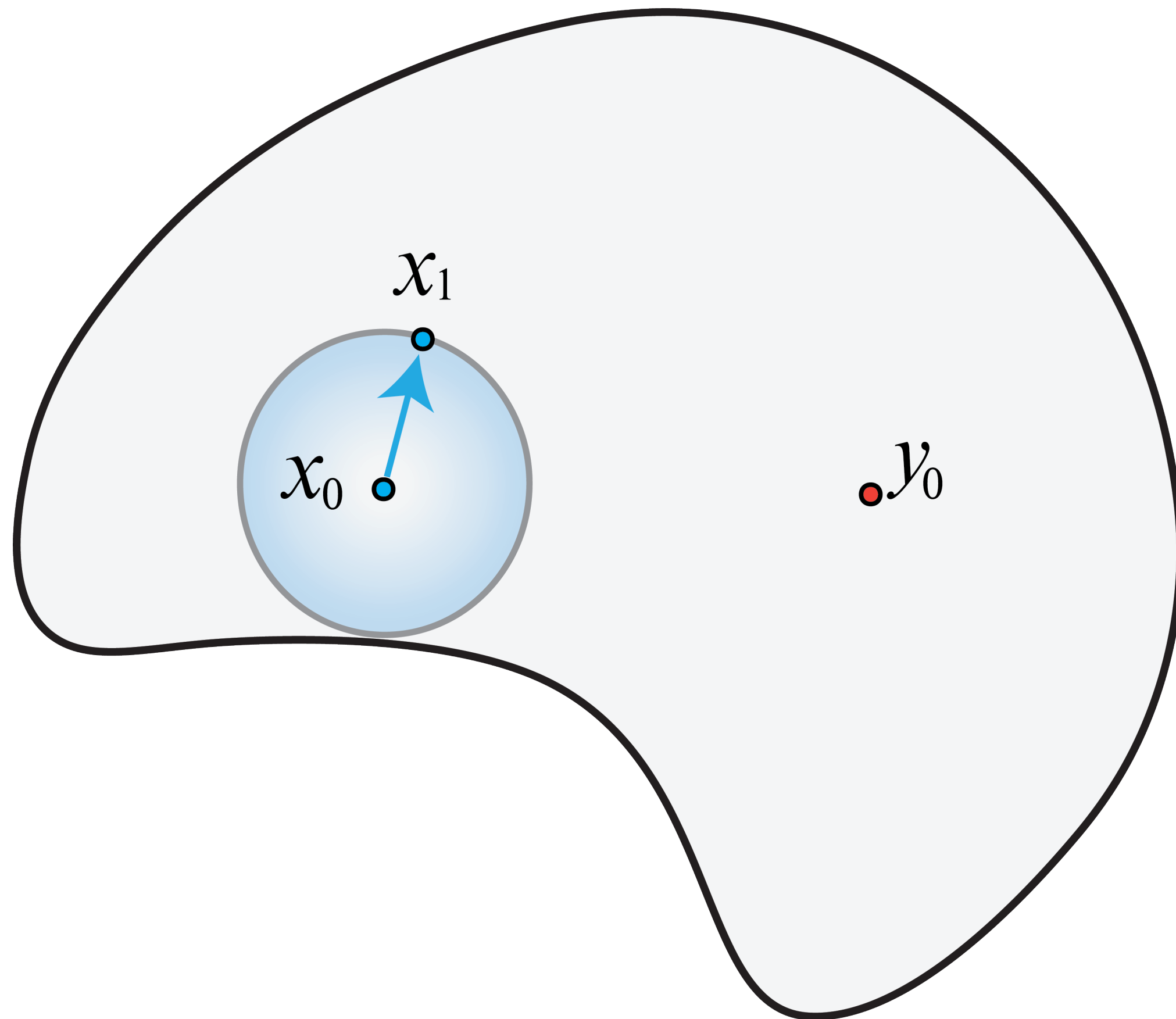


Forward Estimator:

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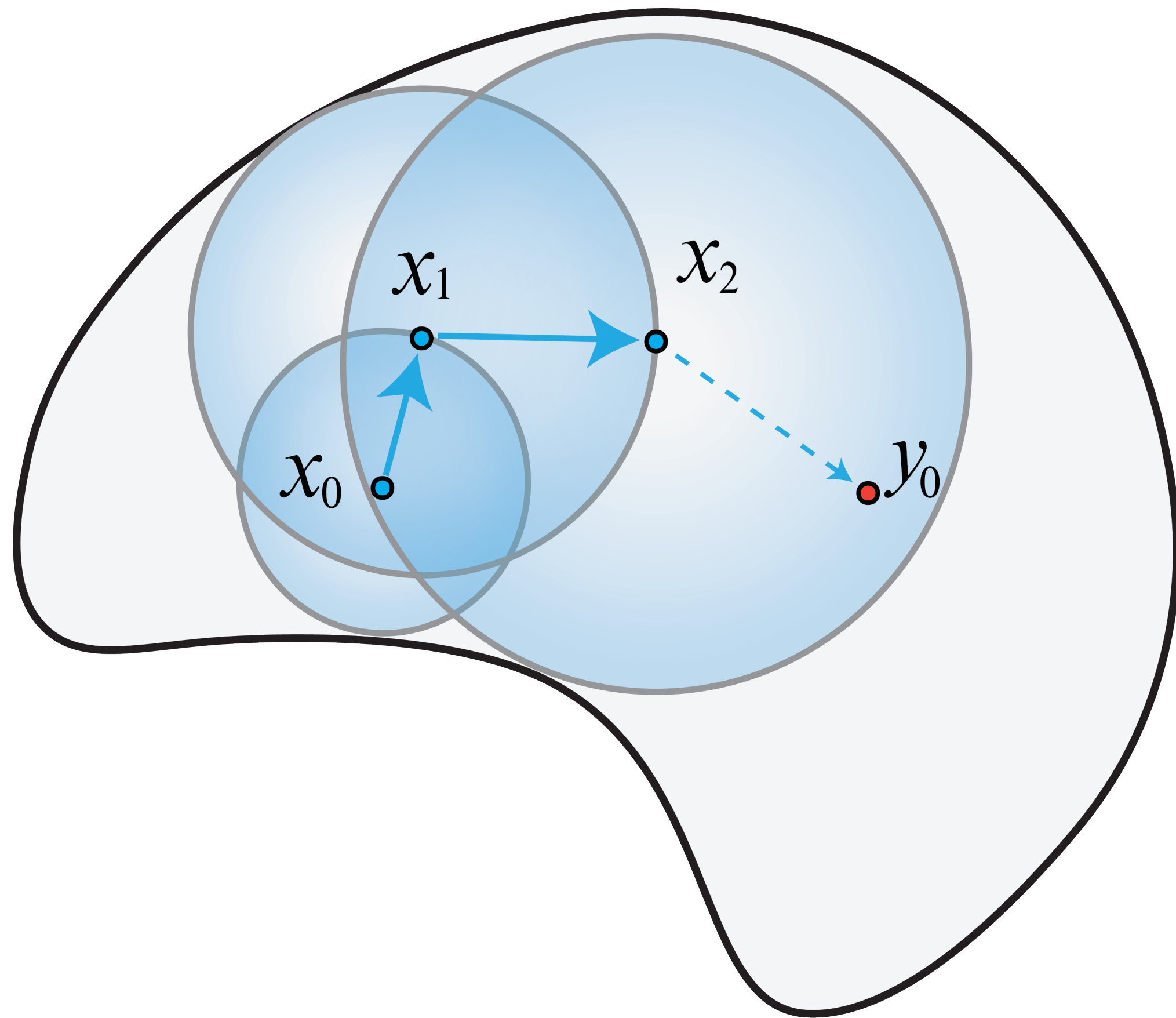


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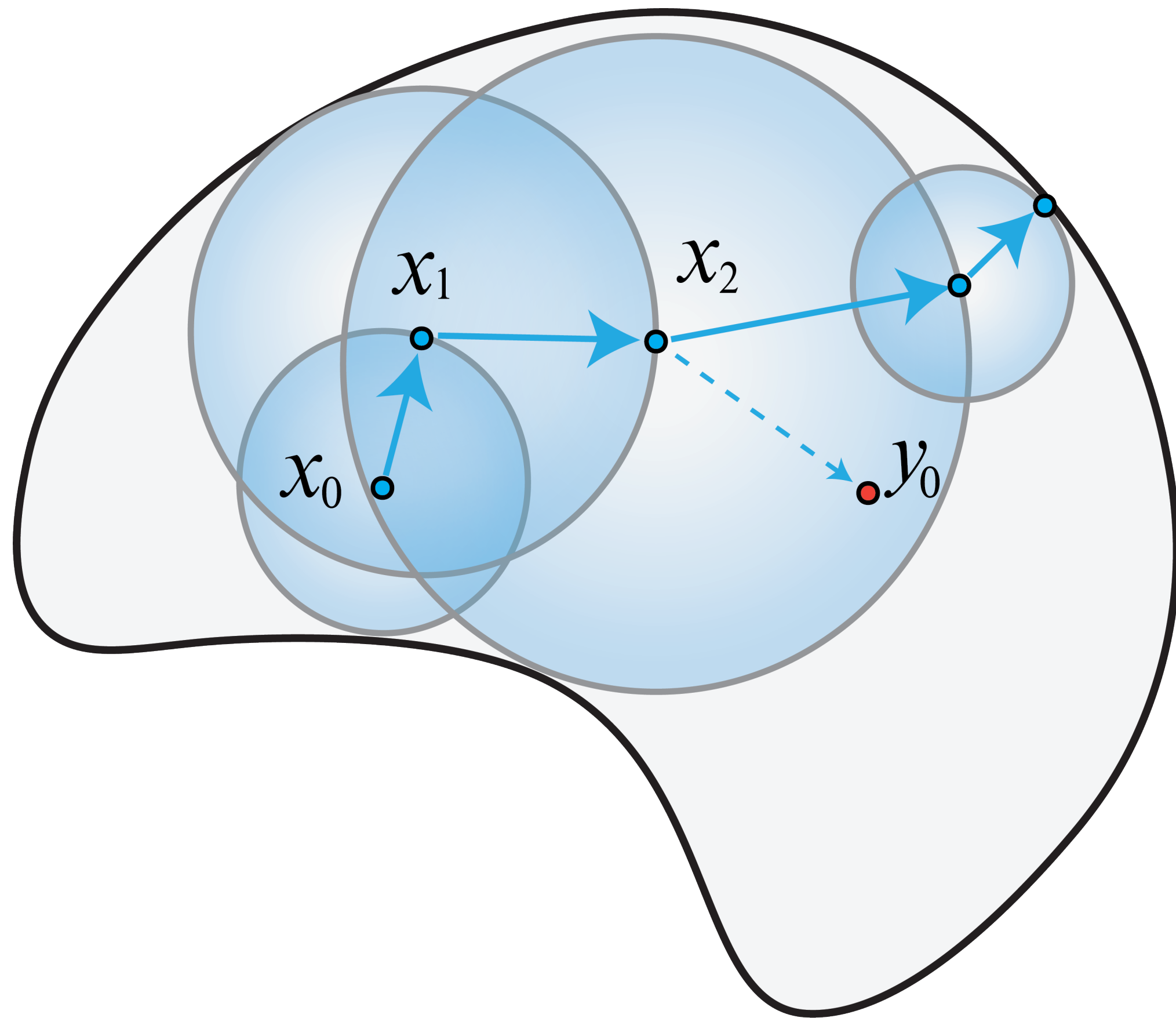


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$$\langle \mathcal{G}(x \leftrightarrow y) \rangle = \mathcal{G}^{B_x}(x \leftrightarrow y) + \langle \mathcal{G}(x' \leftrightarrow y) \rangle$$

$$\langle \mathcal{G}(x \leftrightarrow y) \rangle = 0 \quad \text{if } x \in \partial U$$

Reverse Estimator

Reverse: $\mathcal{G}(x \leftrightarrow y) = \mathcal{G}^{B_y}(x \leftrightarrow y) + \frac{1}{|\partial B_y|} \int_{\partial B_y} \mathcal{G}(x \leftrightarrow y') dy' .$

Reverse Estimator:

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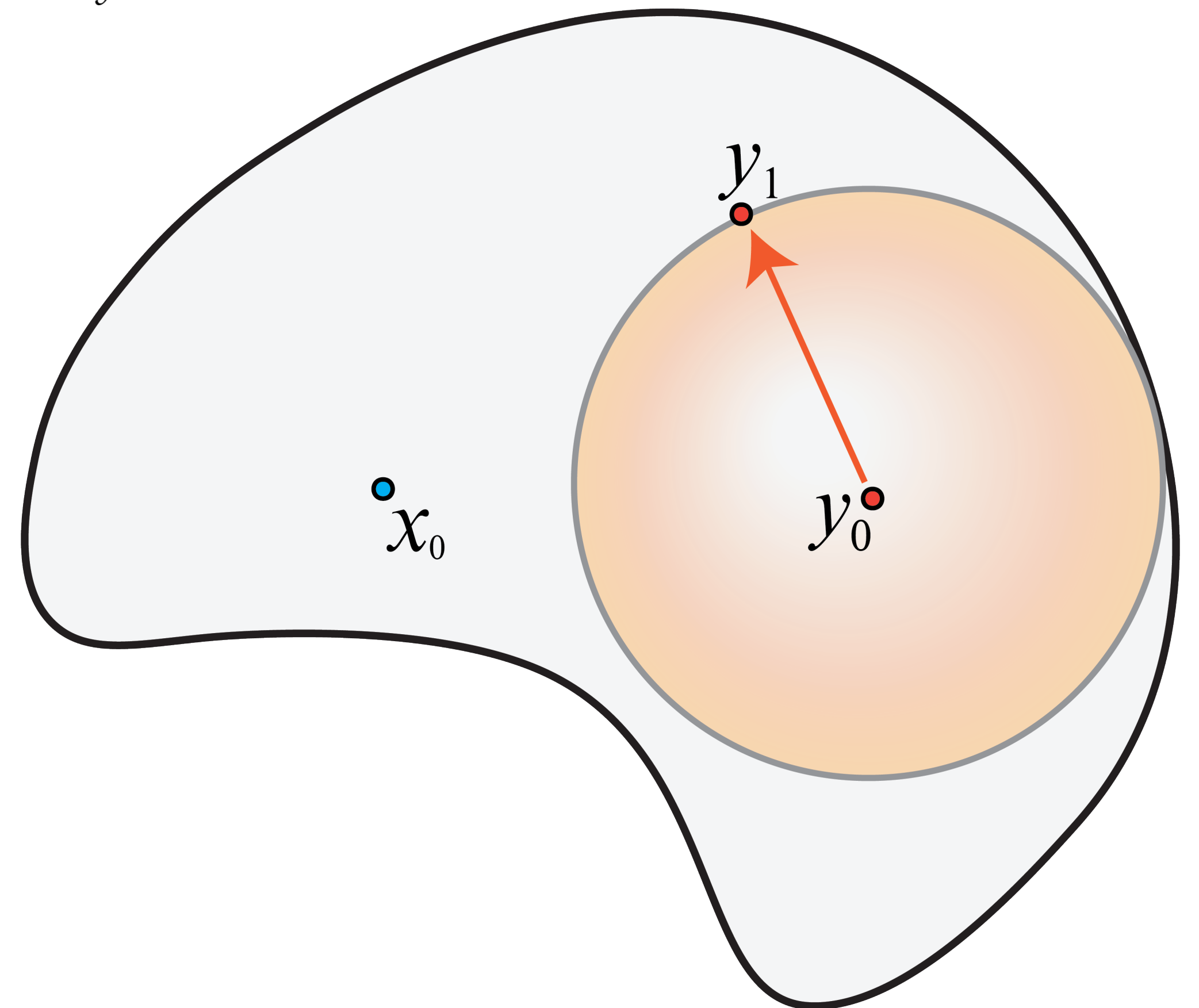


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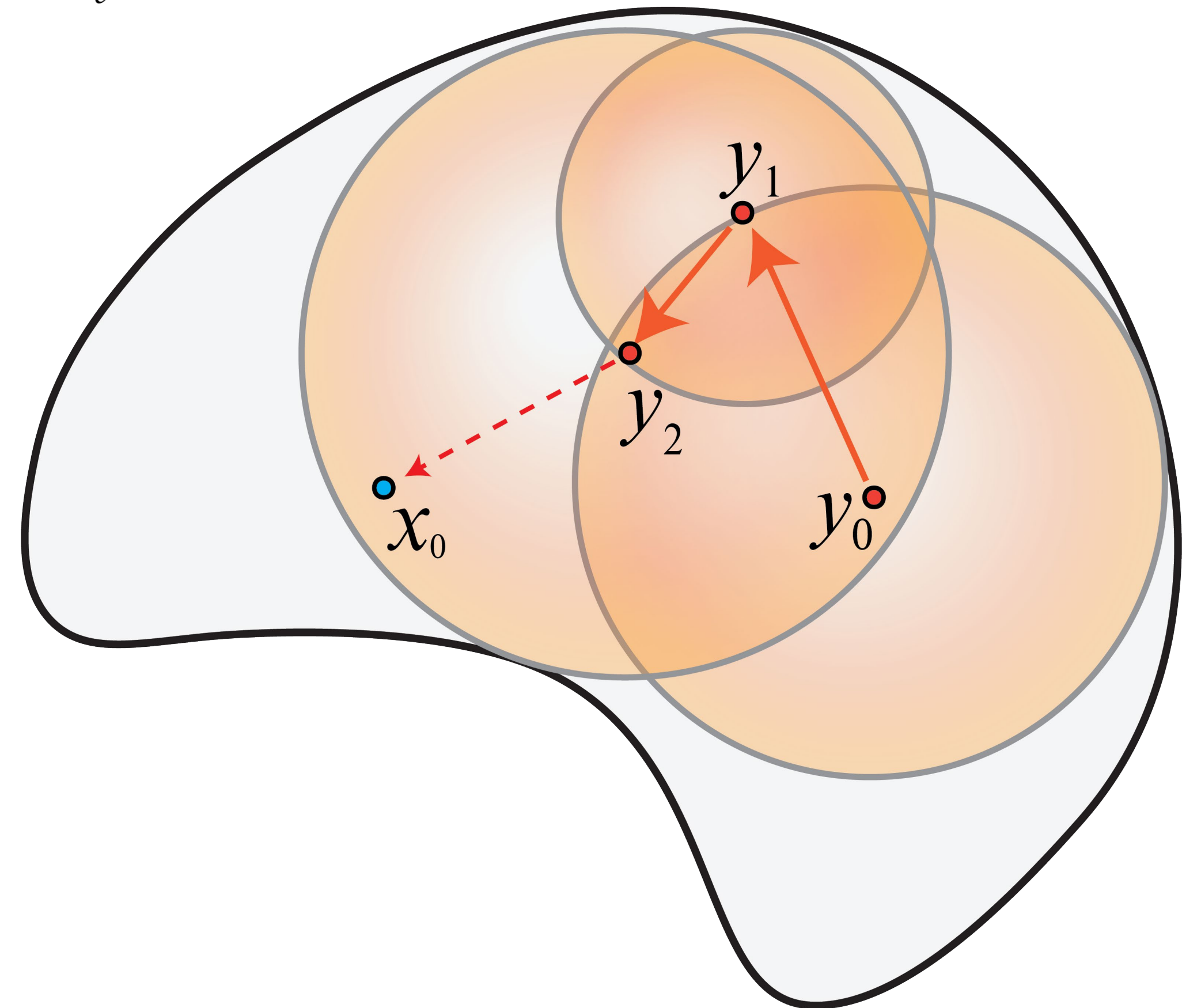


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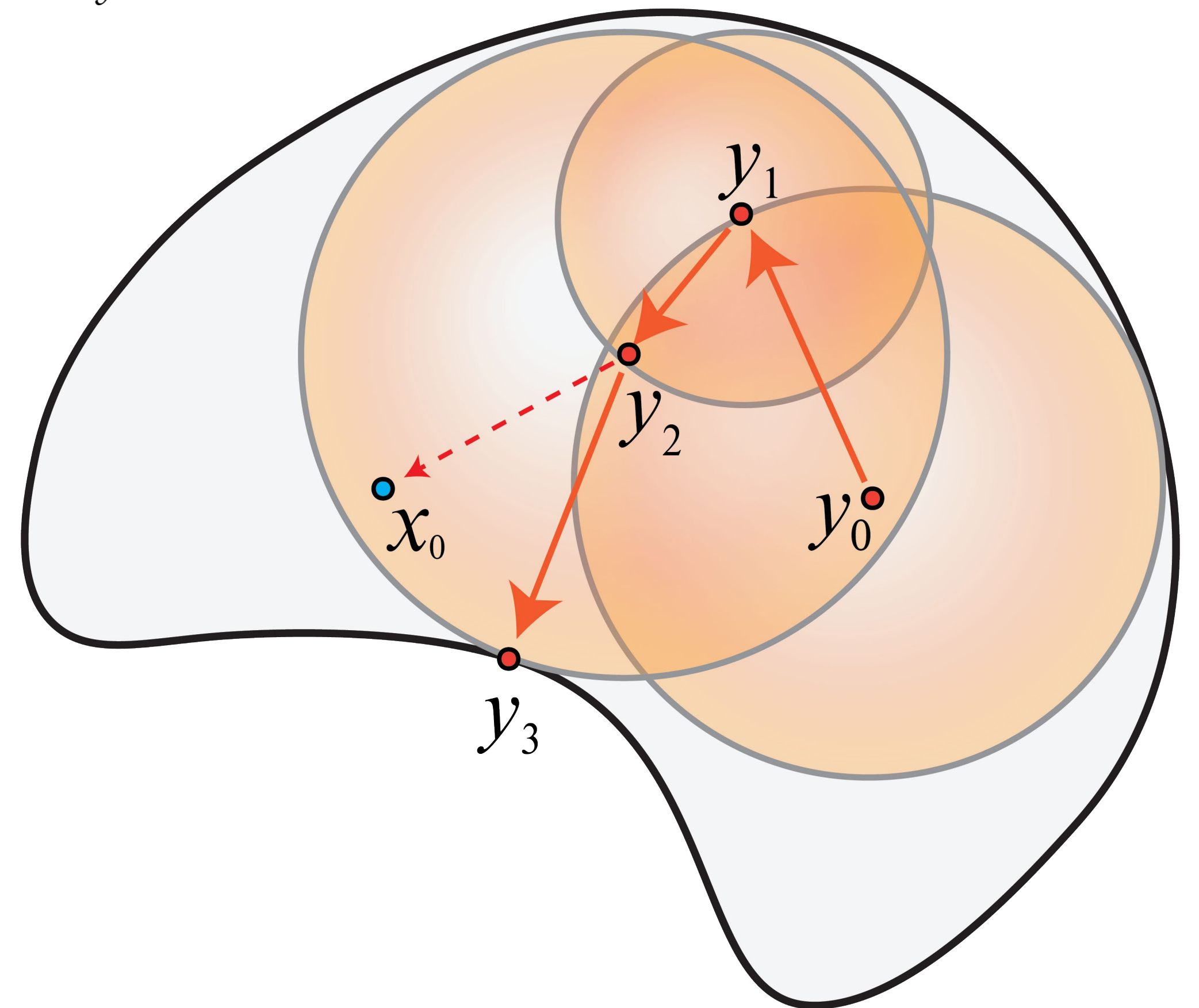
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$$\langle \mathcal{G}(x \leftrightarrow y) \rangle = 0 \quad \text{if } y \in \partial U$$

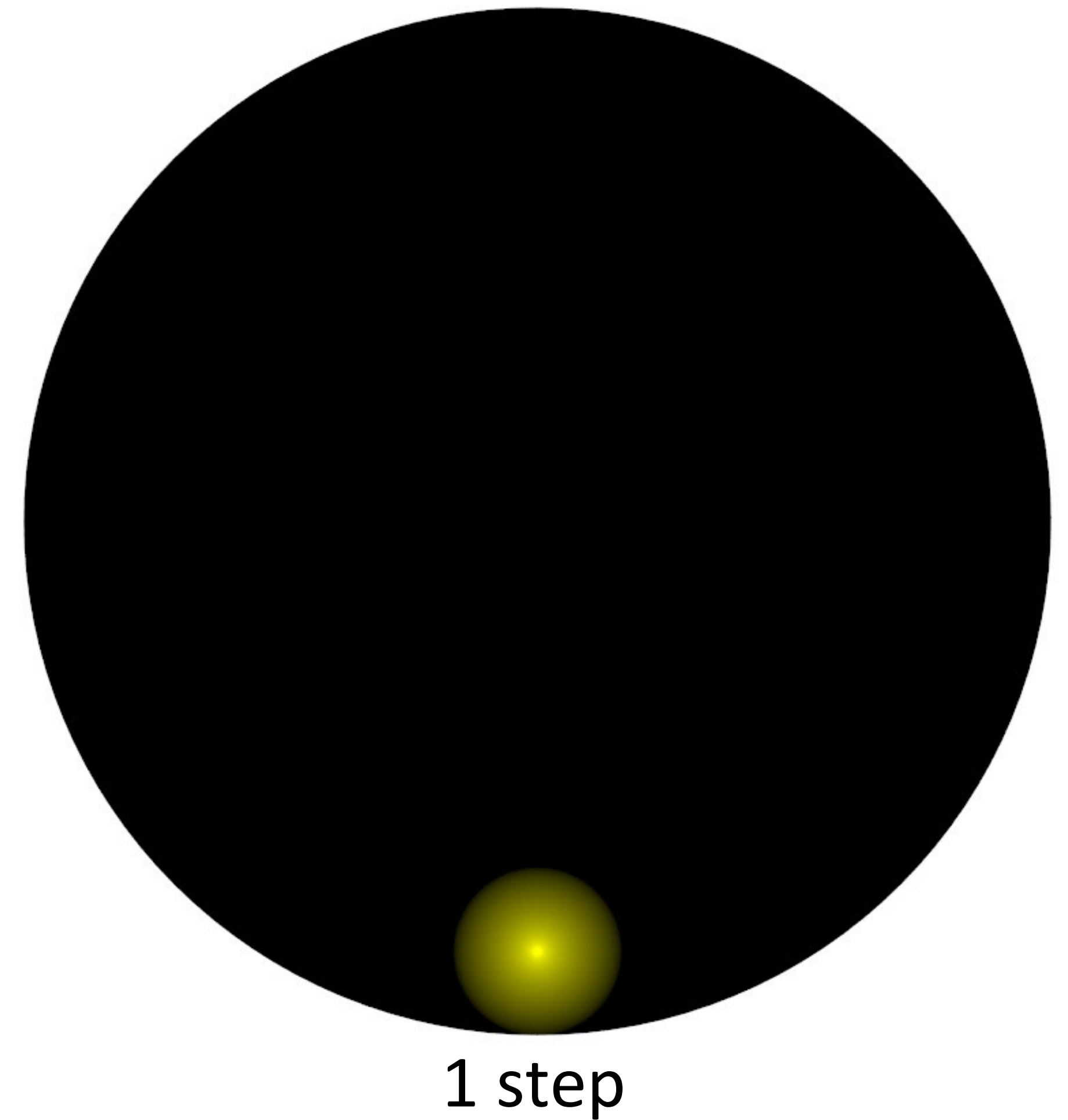


Reverse Estimator

- Start walks from sources
- Globally importance sampling
- Reuse walks

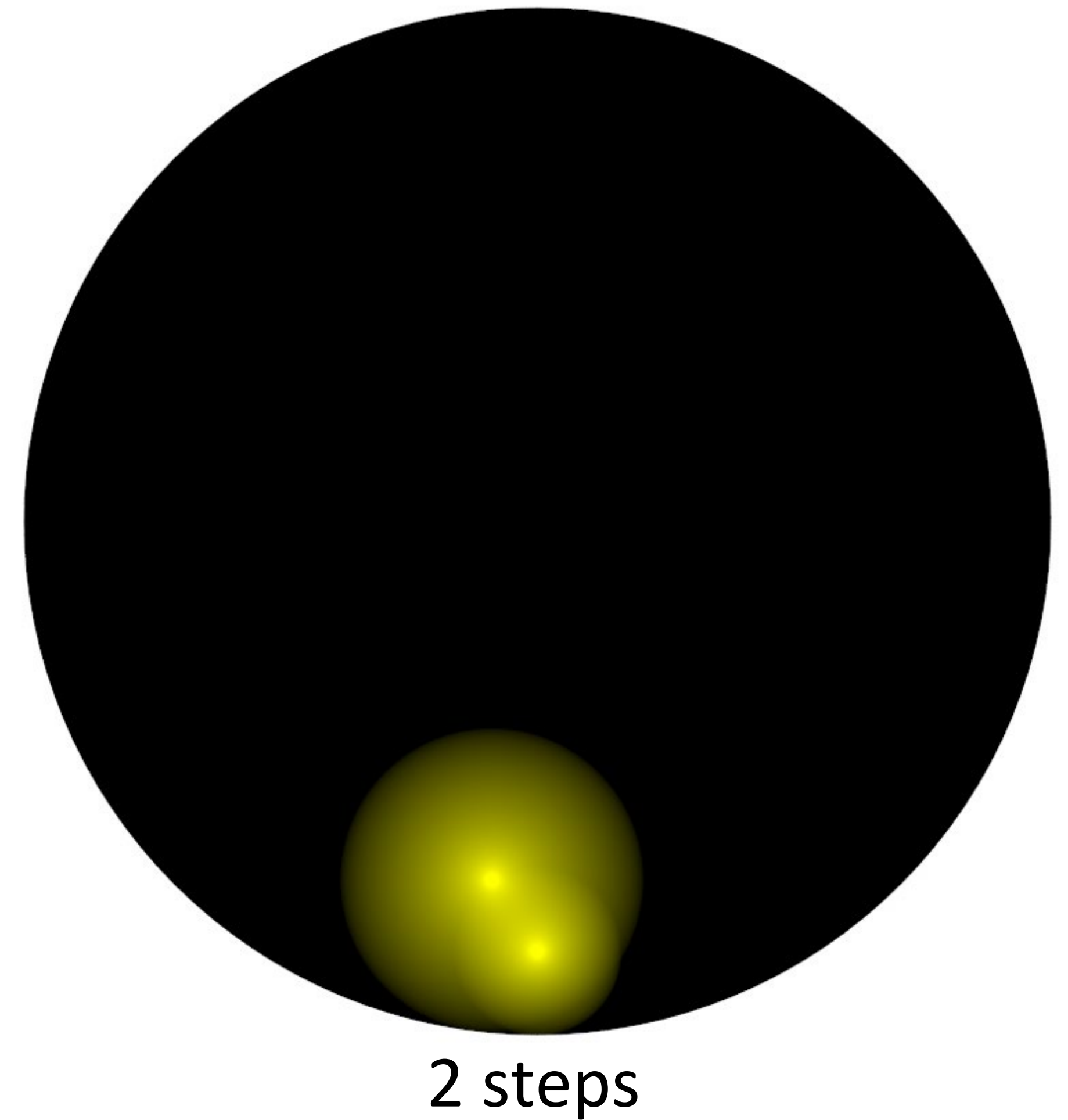
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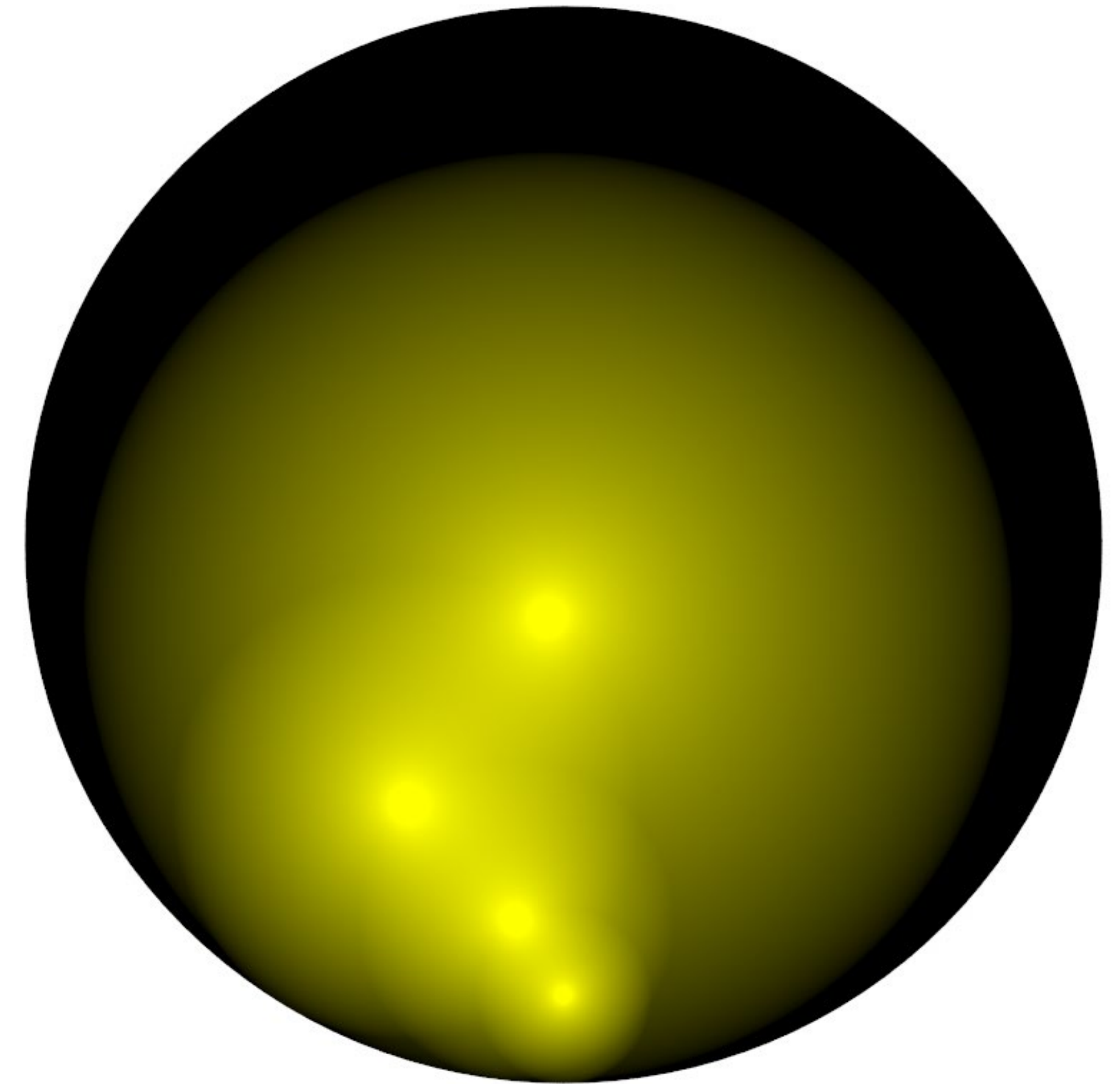
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Reverse Estimator

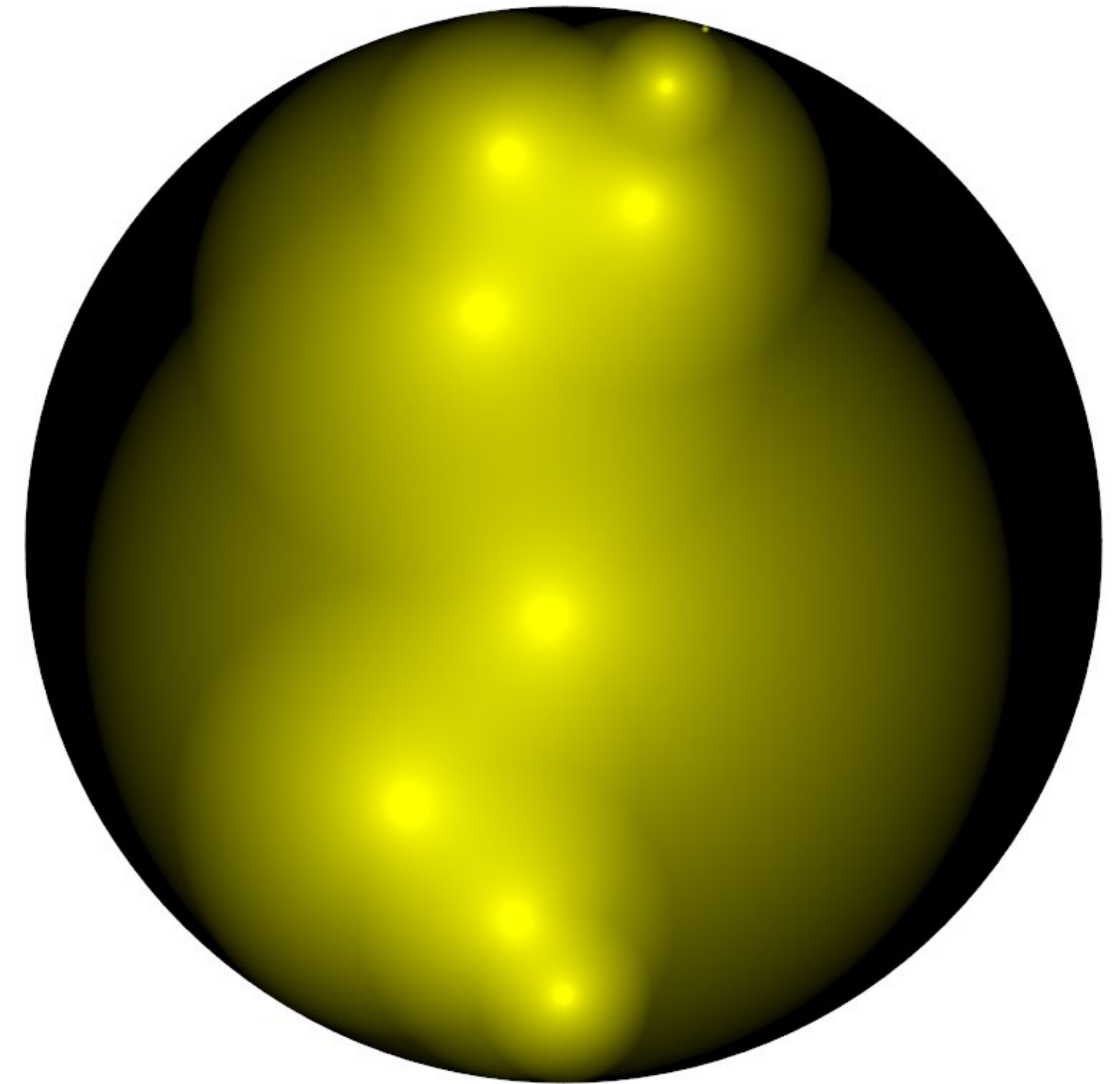
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4 steps

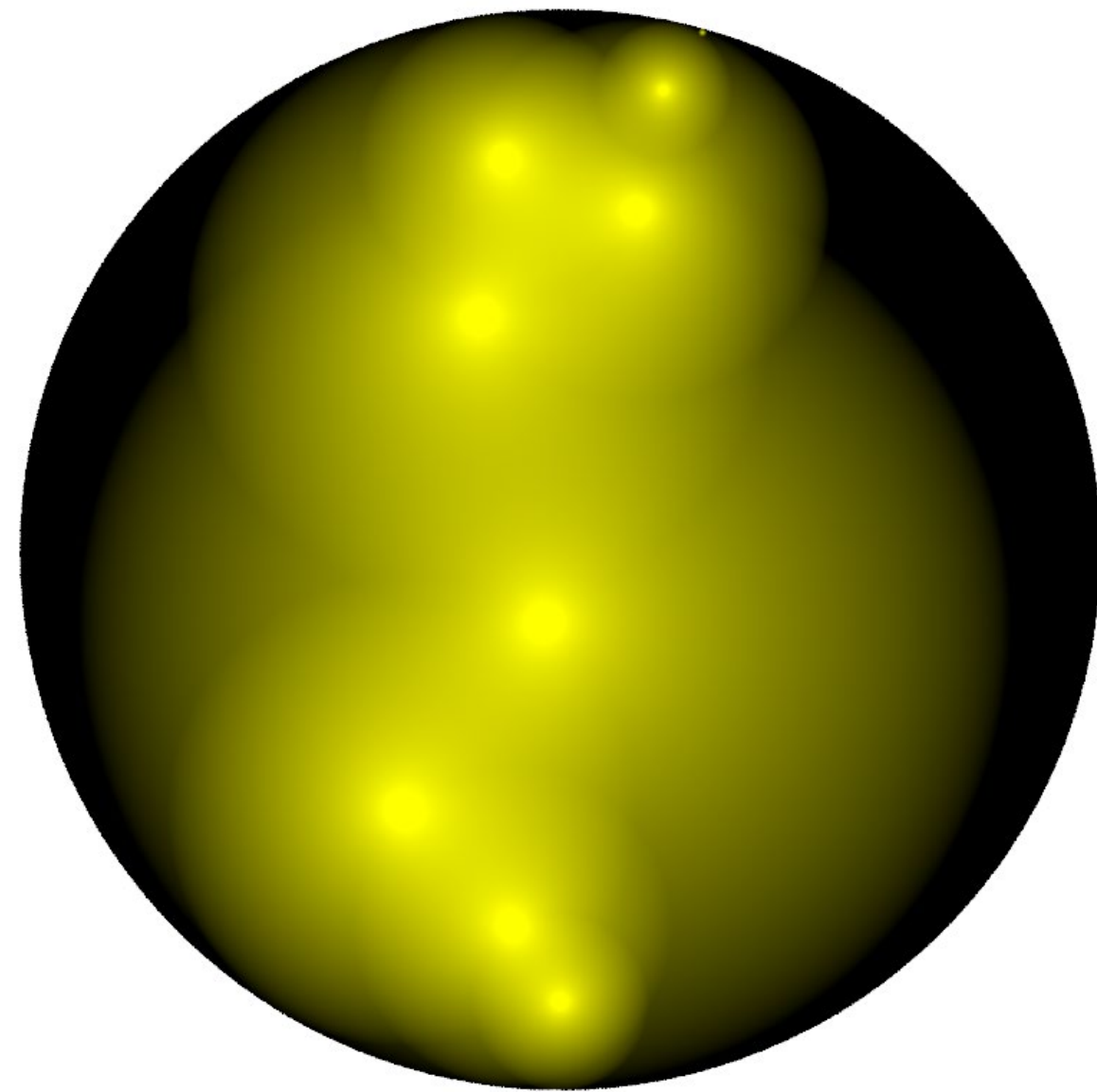
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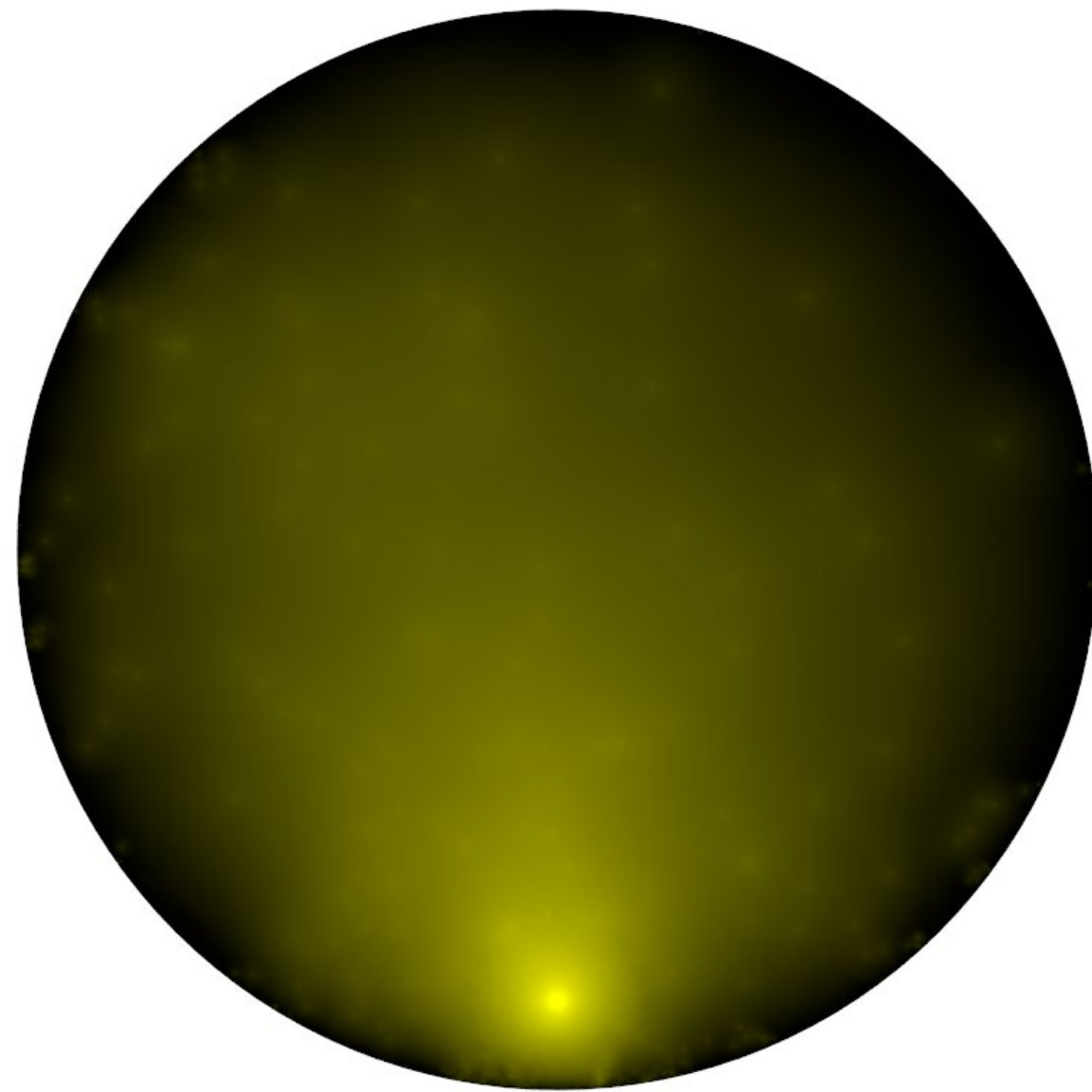


Hits boundary

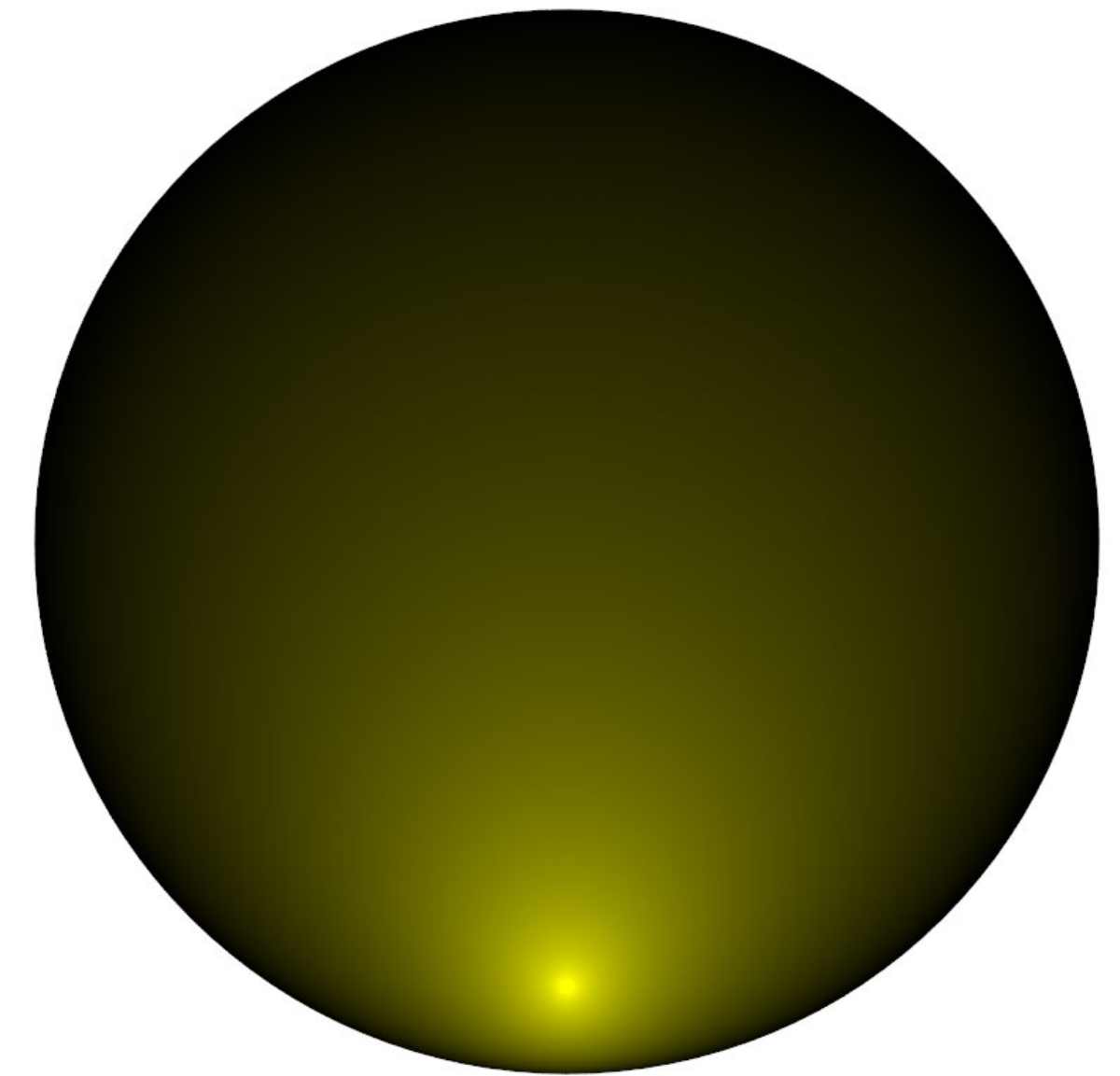
Reverse Estimator



1 sample



64 samples



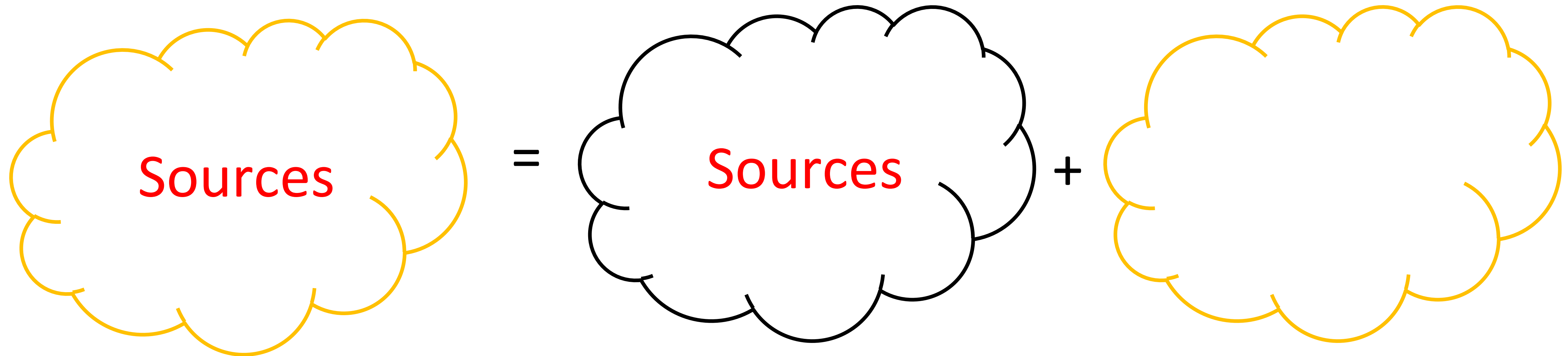
Converged

Solve Poisson's equation

Linear Property:

Source only part

Boundary only part



$$u(x) = \int_U f(y) \mathcal{G}(x, y) dy + \int_{\partial U} g(z) \frac{\partial \mathcal{G}(x, z)}{\partial n(z)} dz$$

Source Part

$$u(x) = \int_U f(y) \mathcal{G}(x \leftrightarrow y) dy$$

1. Sample $y \sim p^U(y)$

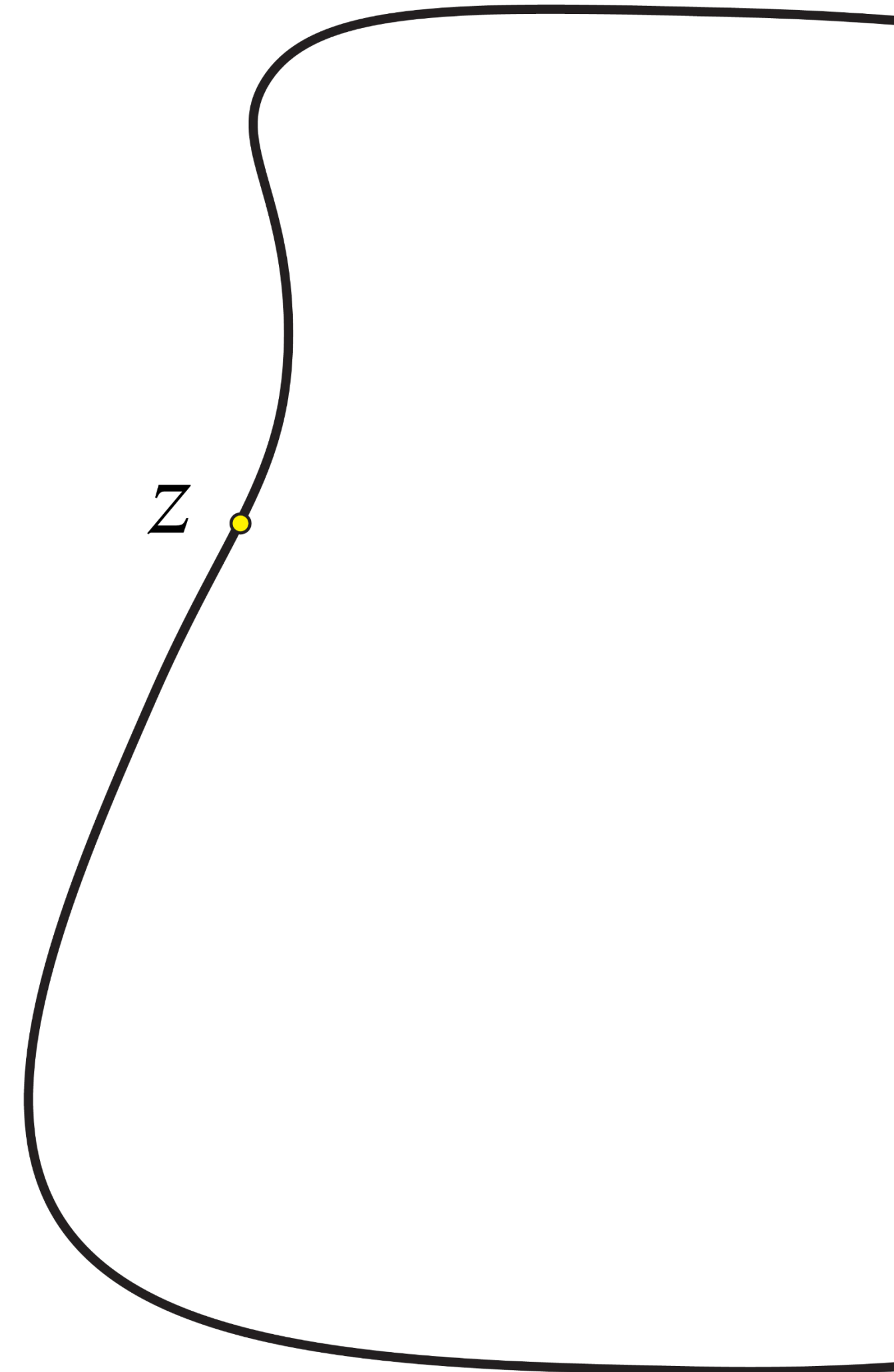
2. Estimate the solution using Green's function estimator $\langle u(x) \rangle = \frac{f(y) \langle \mathcal{G}(x \leftrightarrow y) \rangle}{p^U(y)}$

Boundary Part

$$u(x) = \int_{\partial U} g(z) \frac{\partial \mathcal{G}(x \leftrightarrow z)}{\partial z} dz$$

Use finite difference method to estimate the derivative.

1. Sample $z \sim p^{\partial U}(z)$

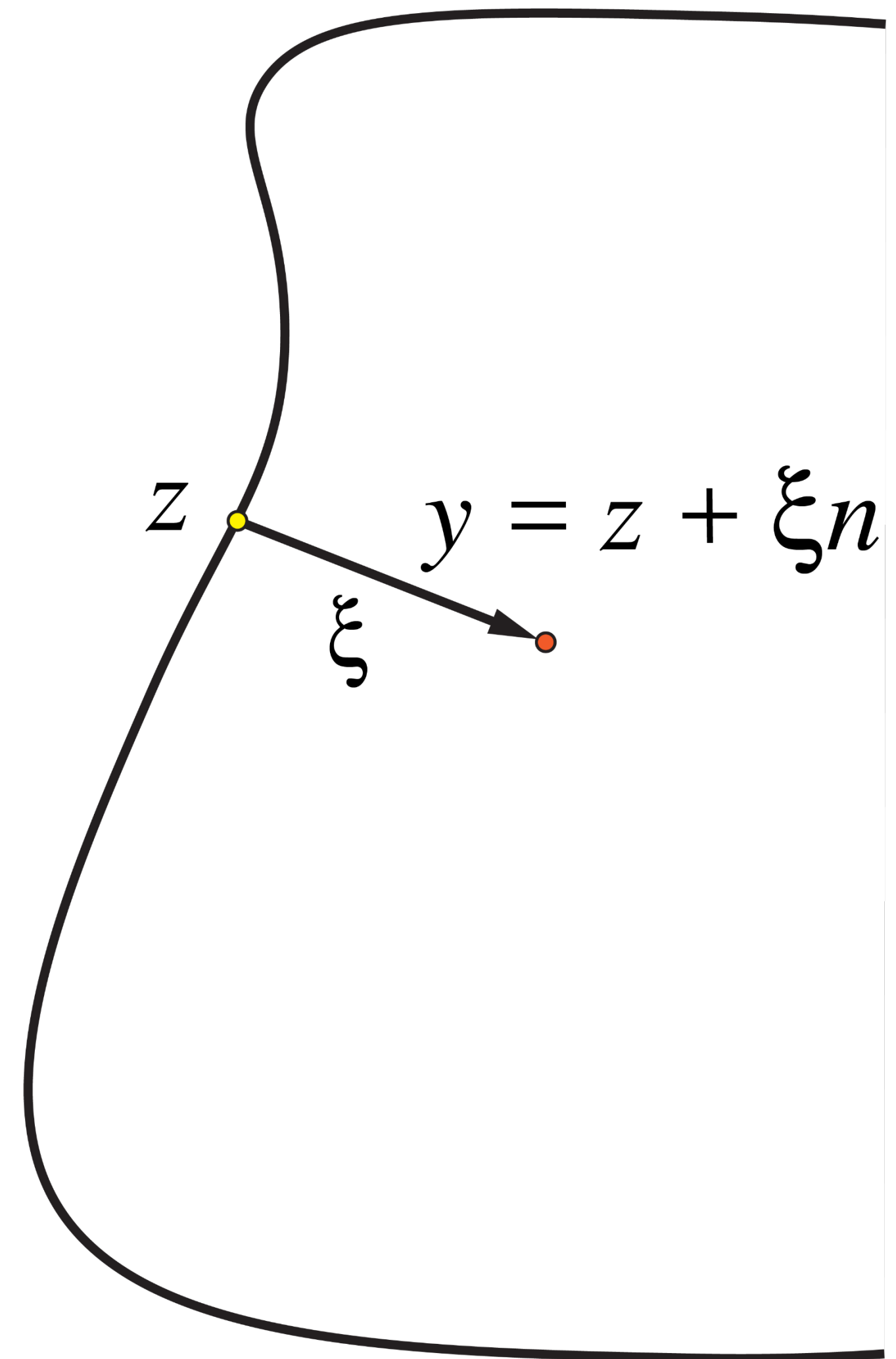


Boundary Part

$$u(x) = \int_{\partial U} g(z) \frac{\partial \mathcal{G}(x \leftrightarrow z)}{\partial z} dz$$

Use finite difference method to estimate the derivative

1. Sample $z \sim p^{\partial U}(z)$
2. Push the point out by ξ



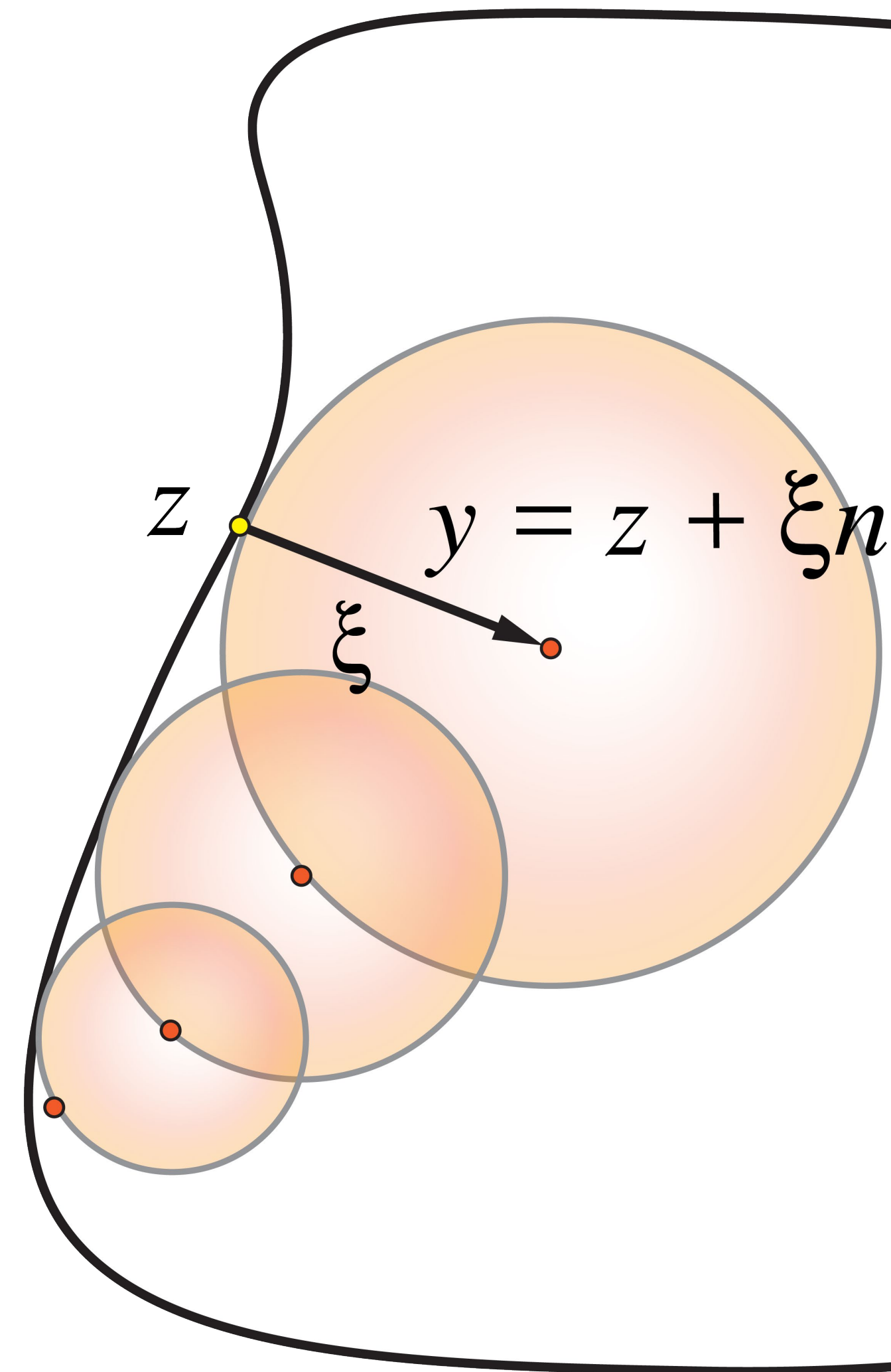
Boundary Part

$$u(x) = \int_{\partial U} g(z) \frac{\partial \mathcal{G}(x \leftrightarrow z)}{\partial z} dz$$

Use finite difference method to estimate the derivative

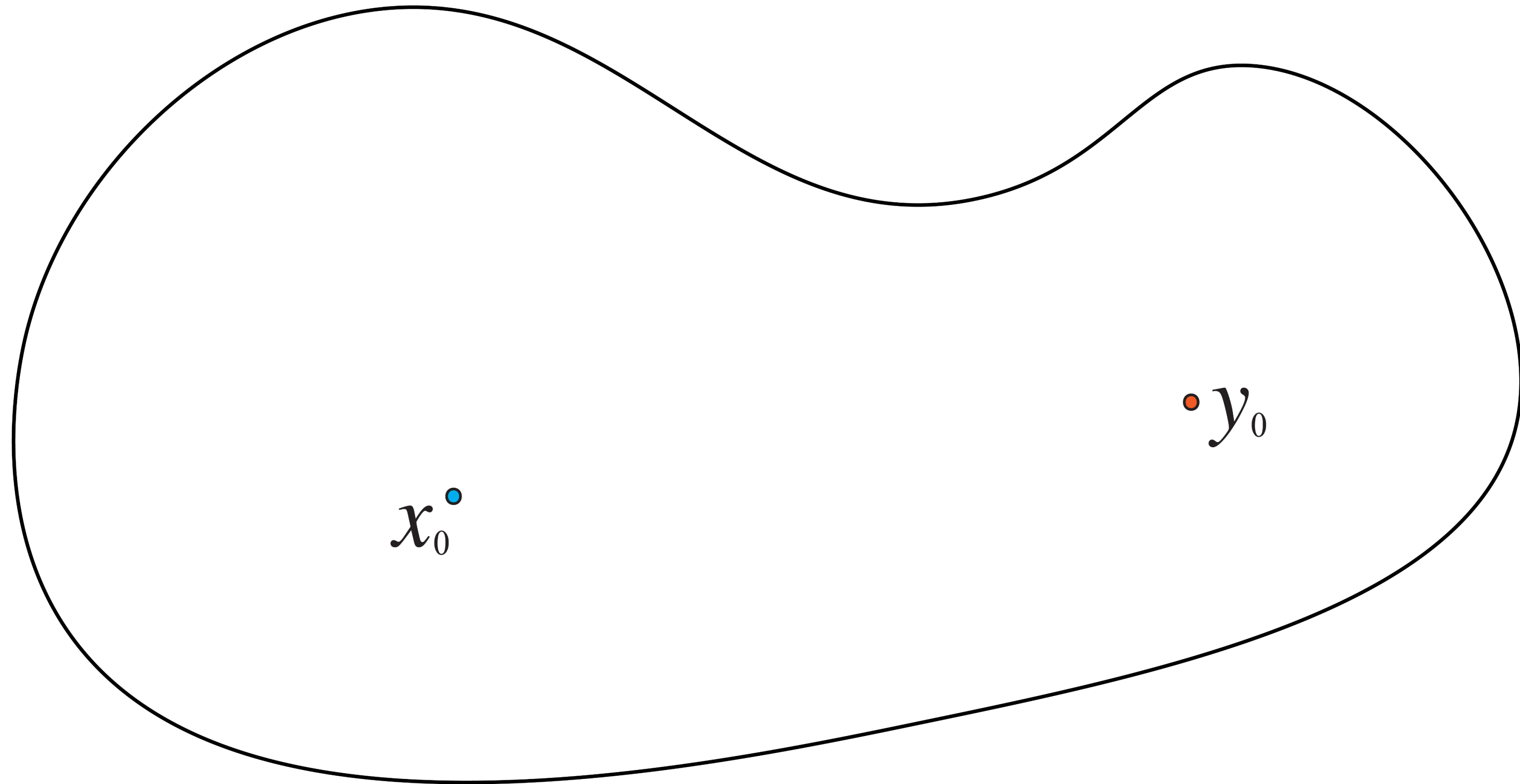
1. Sample $z \sim p^{\partial U}(z)$
2. Push the point out by ξ
3. Estimate the Green's function
4. Estimate the solution through finite difference

$$\langle u(x) \rangle = \frac{g(z) \langle \mathcal{G}(x \leftrightarrow z) \rangle}{p^{\partial U}(z) \xi}$$



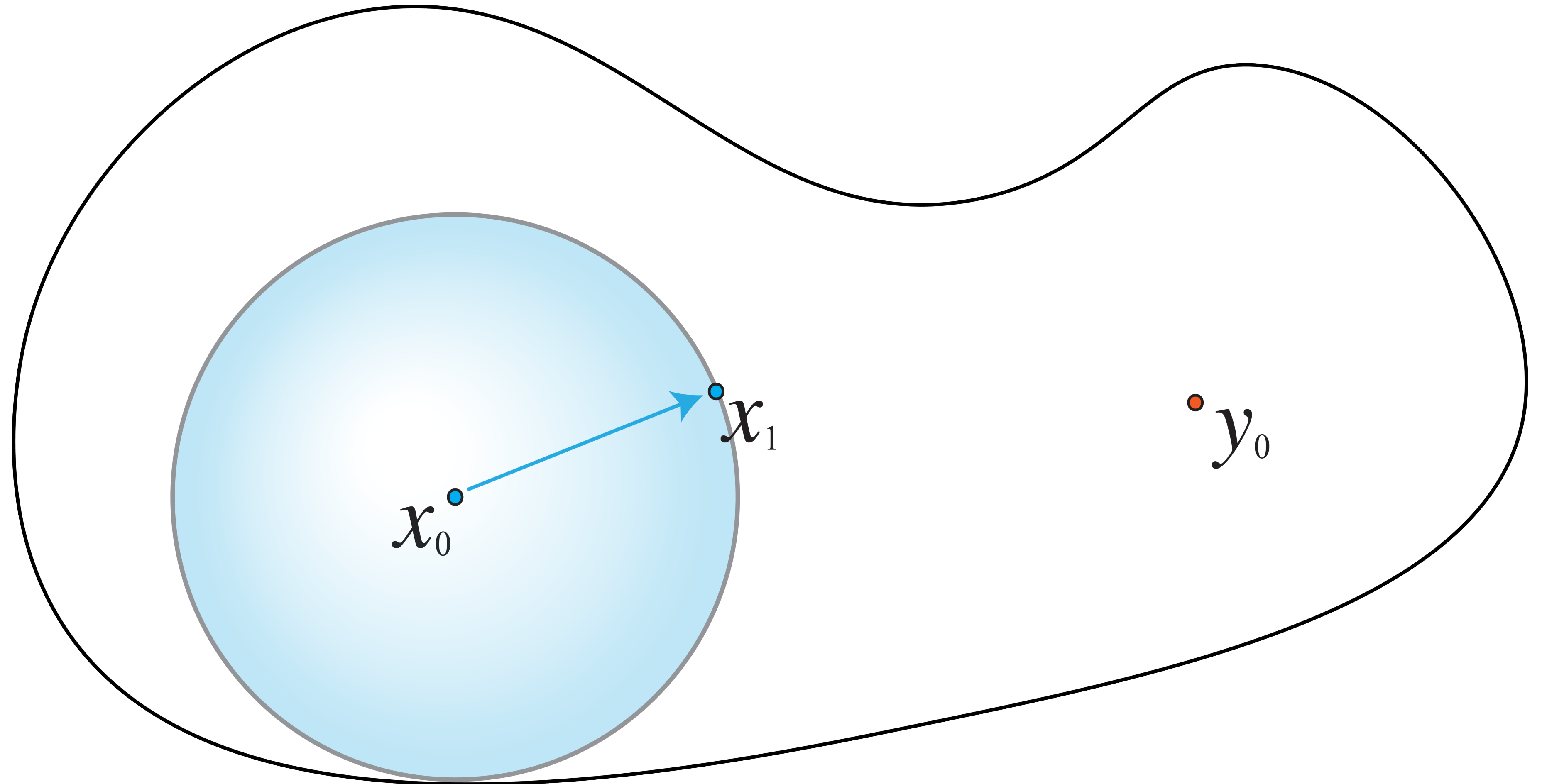
Combine Forward & Reverse

$$\langle \mathcal{G}(x_0 \leftrightarrow y_0) \rangle =$$



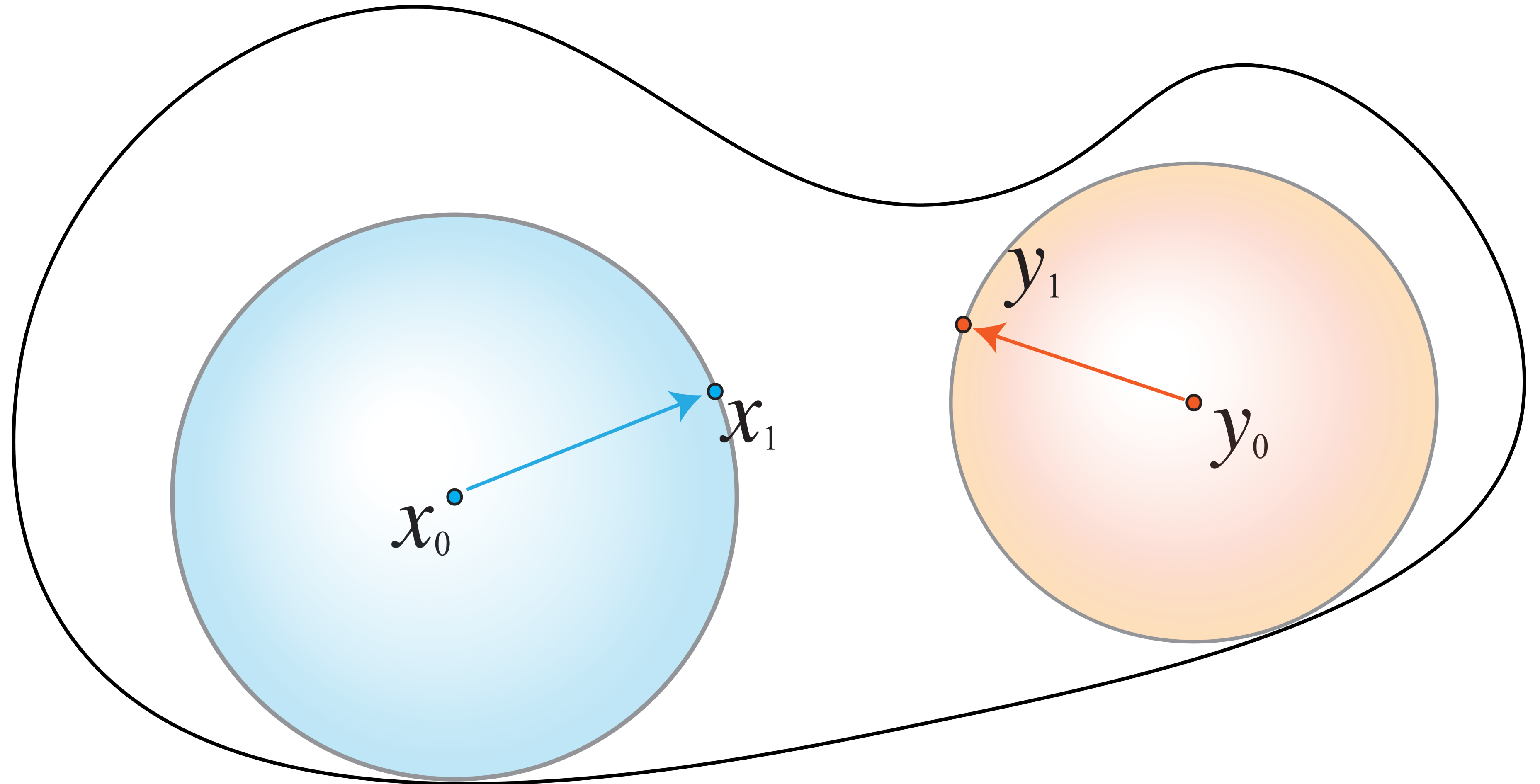
Combine Forward & Reverse

$$\langle \mathcal{G}(x_0 \leftrightarrow y_0) \rangle = 0 + \langle \mathcal{G}(x_1 \leftrightarrow y_0) \rangle$$



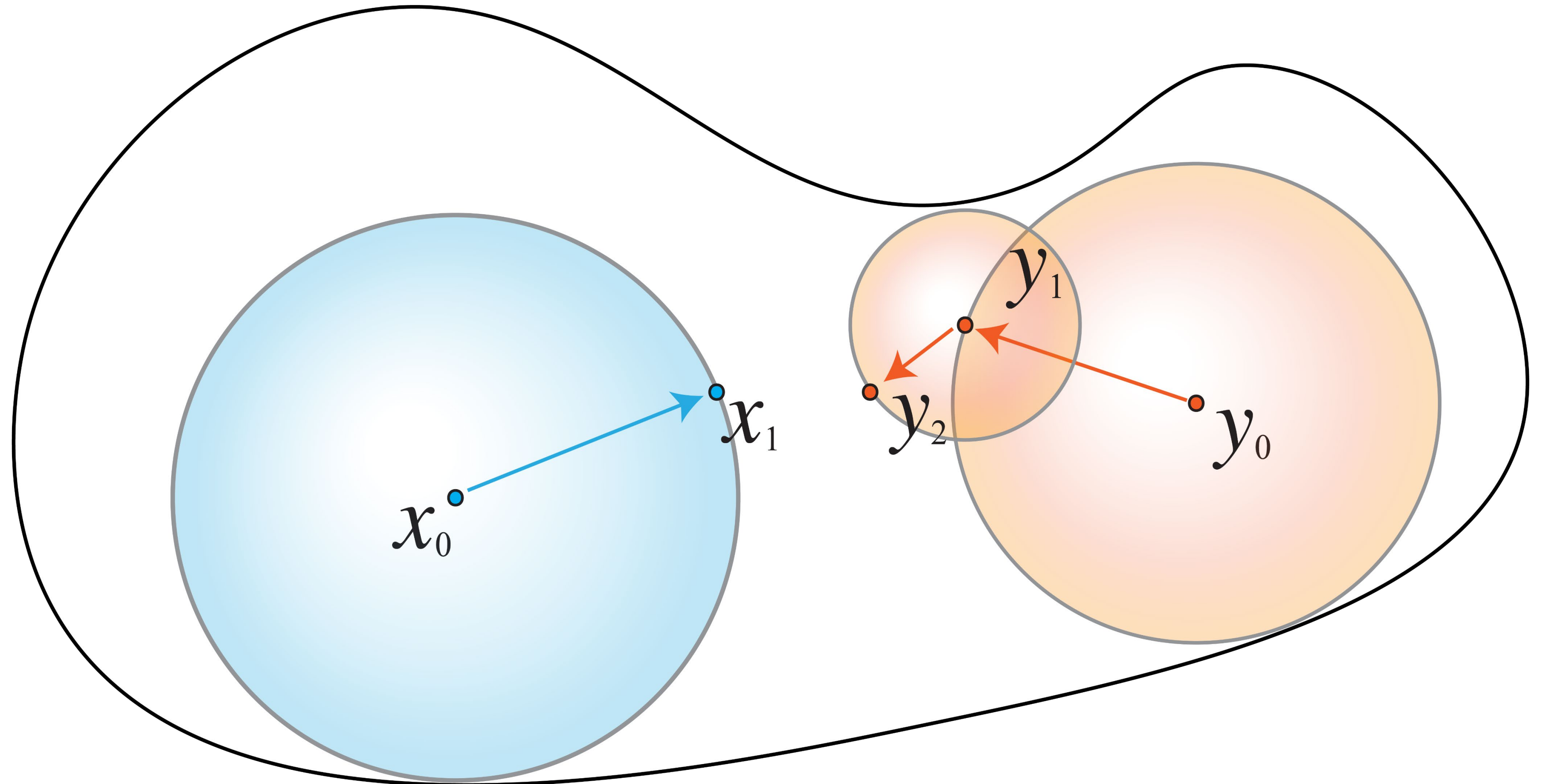
Combine Forward & Reverse

$$\langle \mathcal{G}(x_0 \leftrightarrow y_0) \rangle = 0 + \langle \mathcal{G}(x_1 \leftrightarrow y_1) \rangle$$



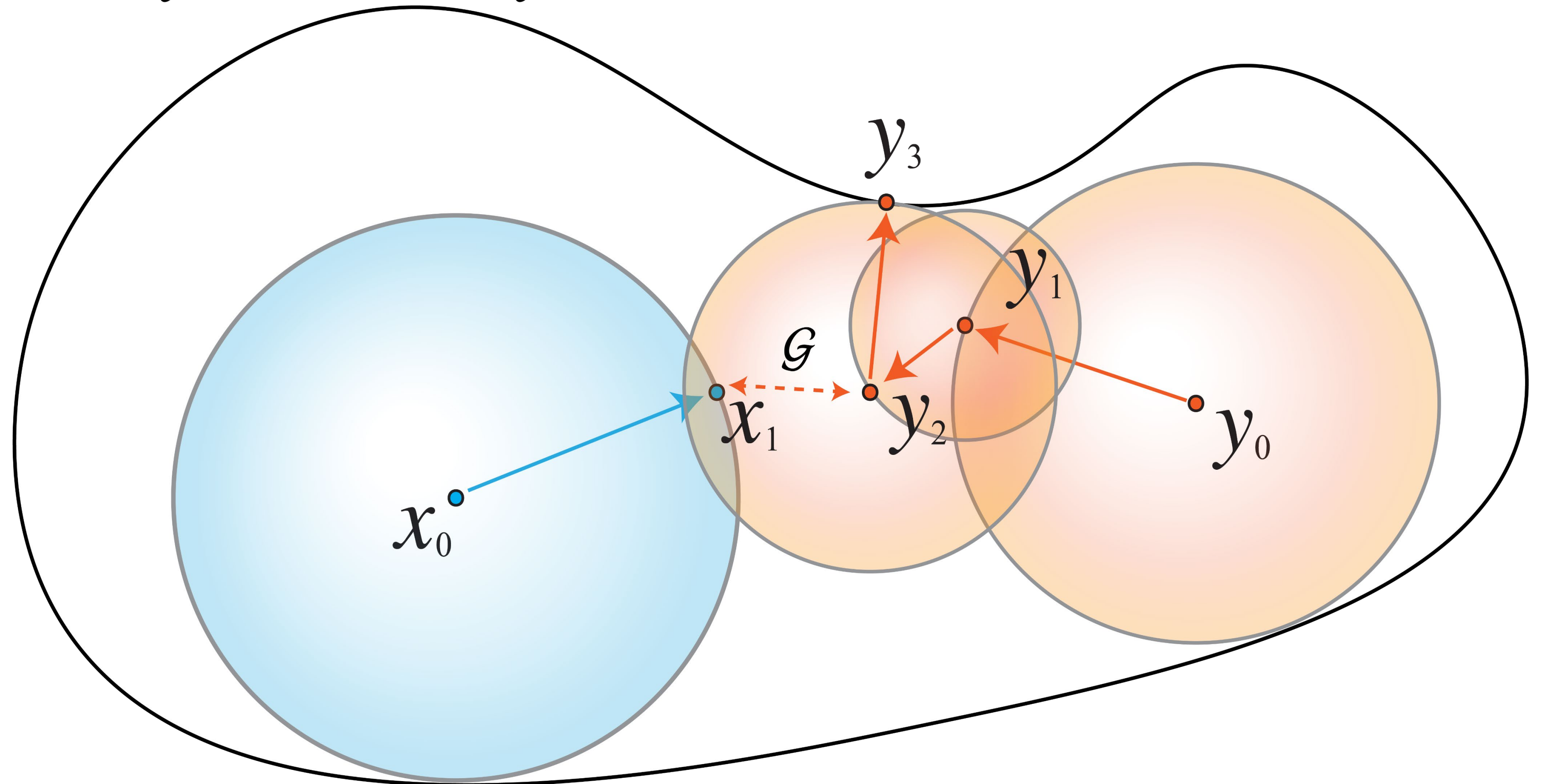
Combine Forward & Reverse

$$\langle \mathcal{G}(x_0 \leftrightarrow y_0) \rangle = 0 + \langle \mathcal{G}(x_1 \leftrightarrow y_2) \rangle$$



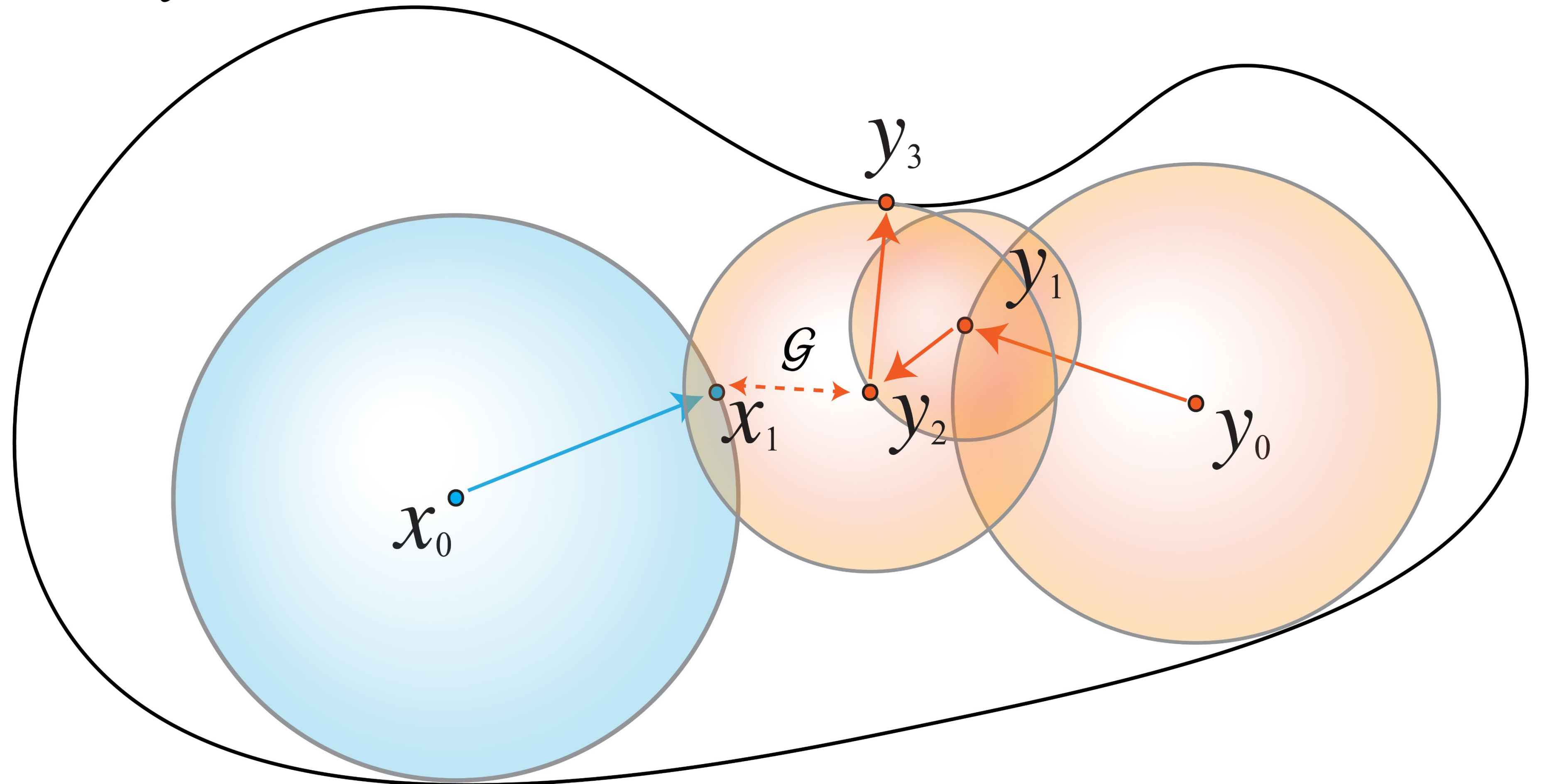
Combine Forward & Reverse

$$\langle \mathcal{G}(x_0 \leftrightarrow y_0) \rangle = \mathcal{G}^{B_{y_2}}(x_1 \leftrightarrow y_2) + \langle \mathcal{G}(x_1 \leftrightarrow y_3) \rangle$$

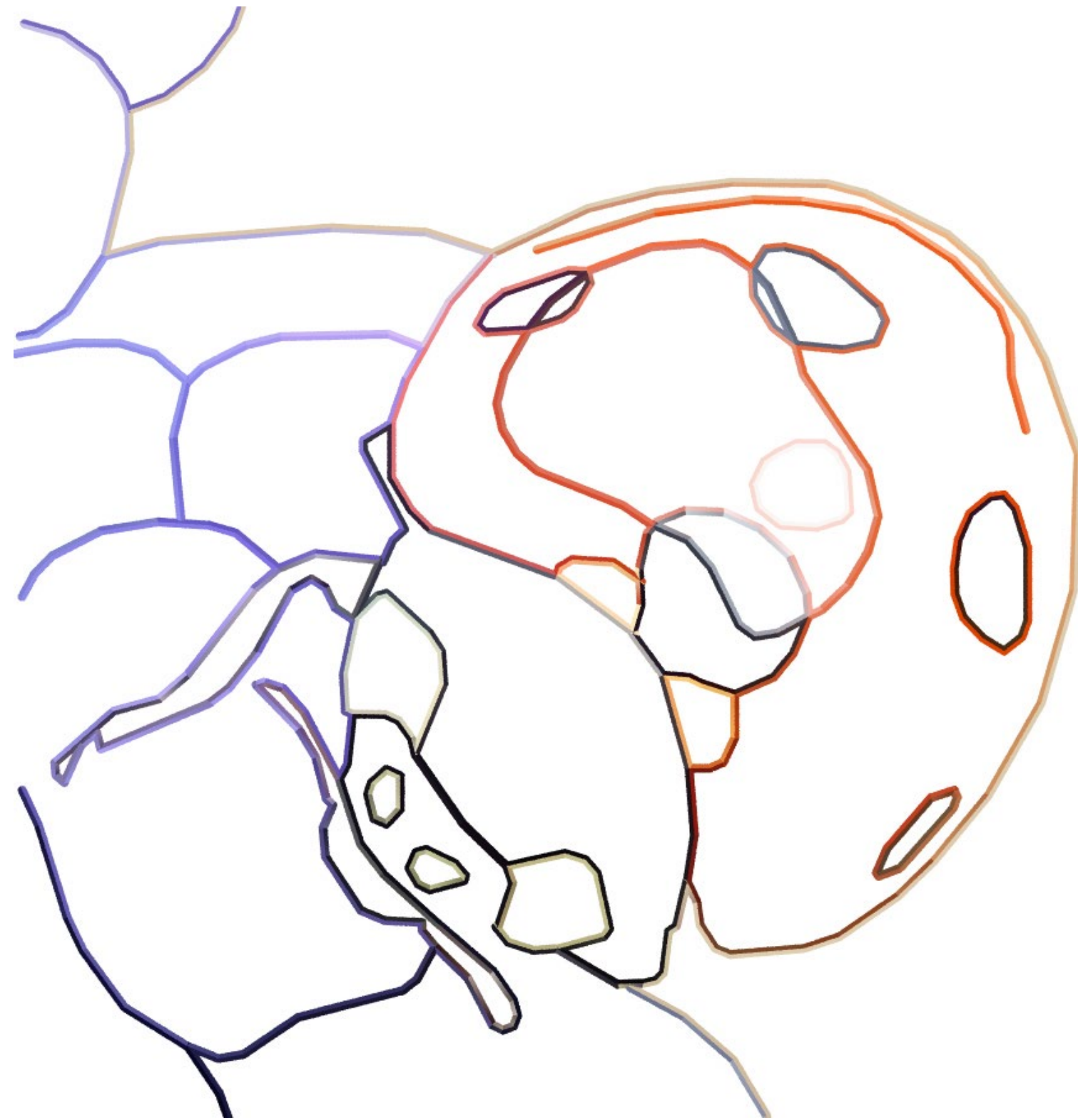


Combine Forward & Reverse

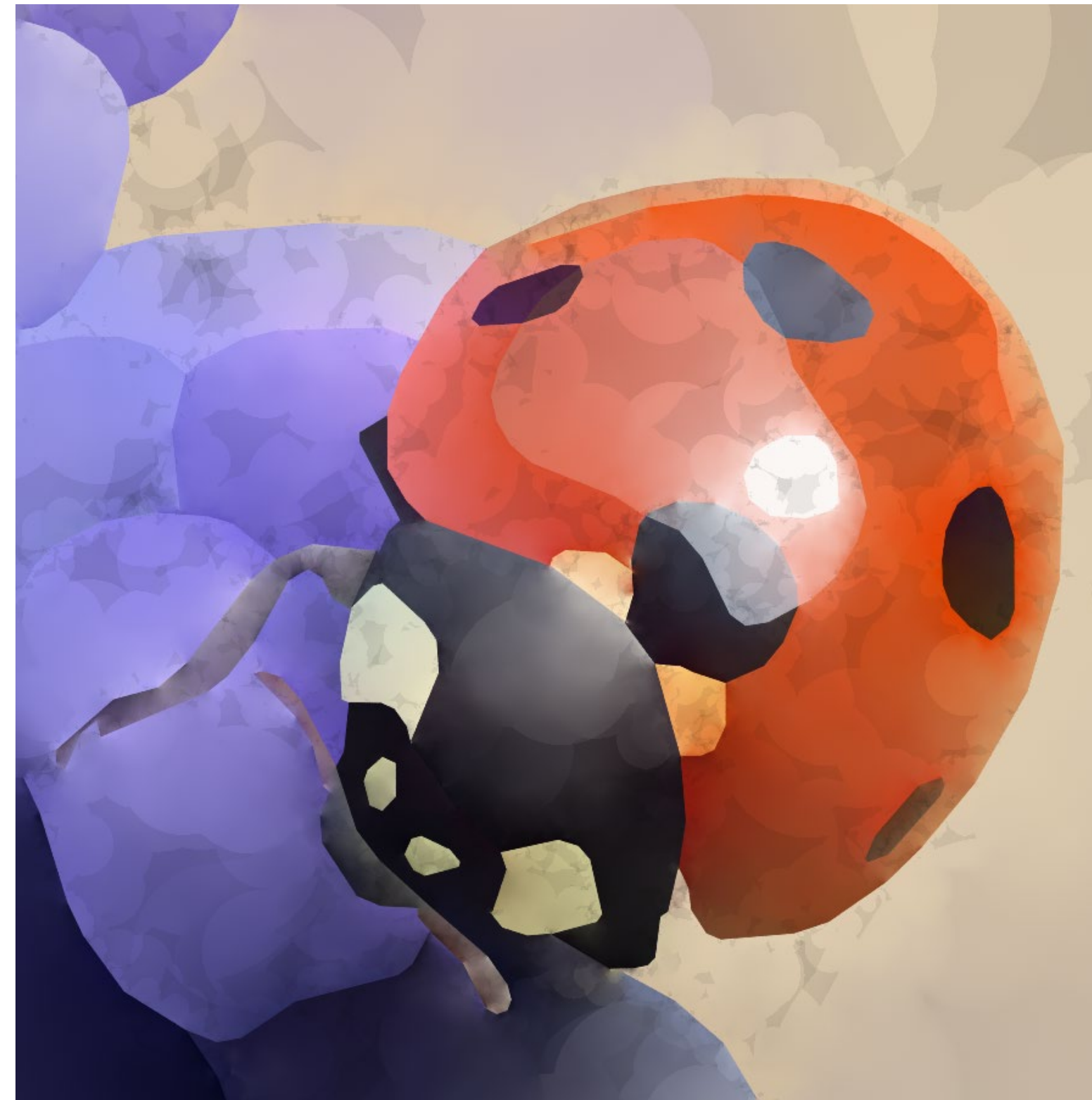
$$\langle \mathcal{G}(x_0 \leftrightarrow y_0) \rangle = \mathcal{G}^{B_{y_2}}(x_1 \leftrightarrow y_2)$$



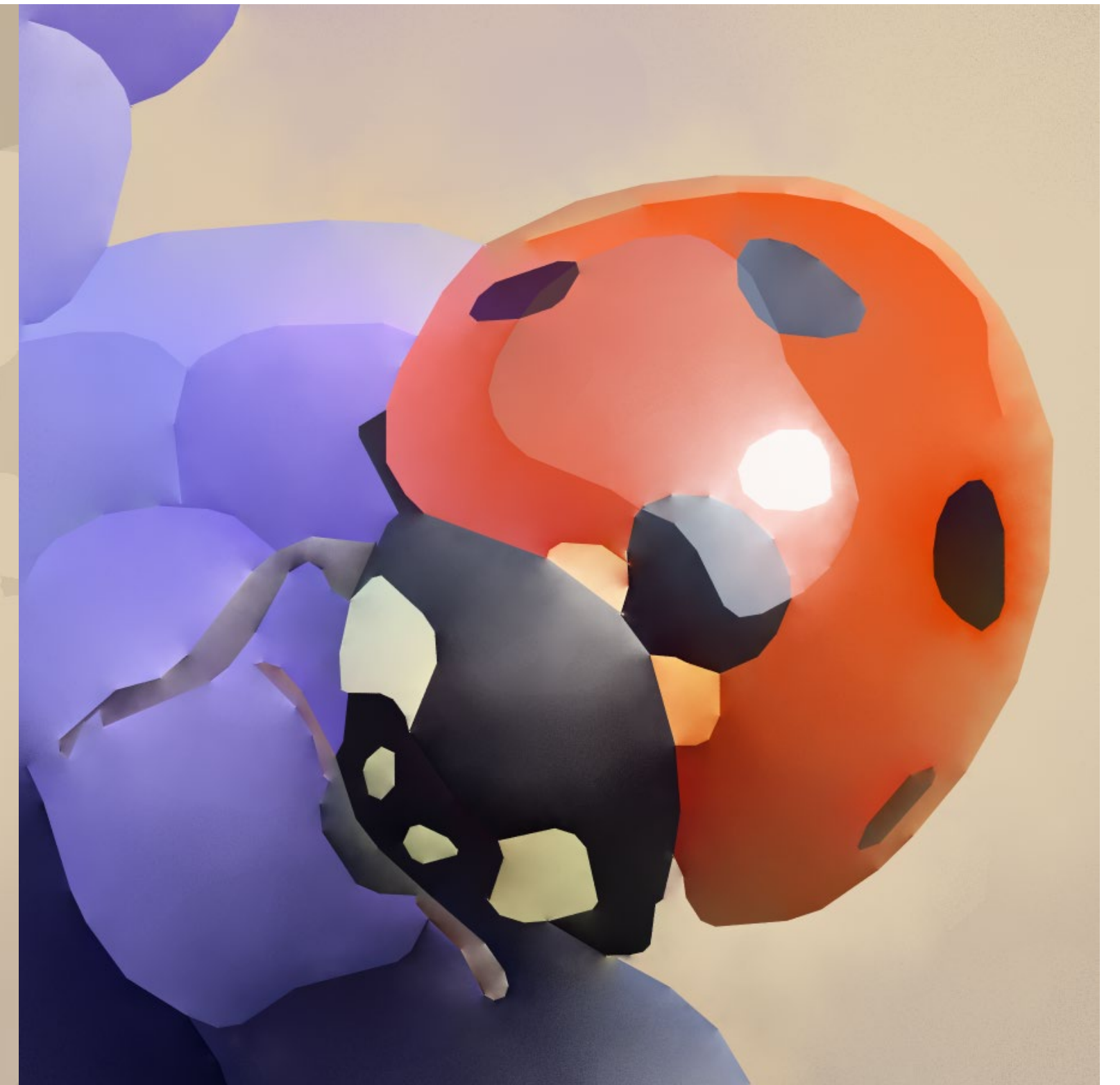
Final Gather



Boundary value



Reverse WoS (Insufficient samples)

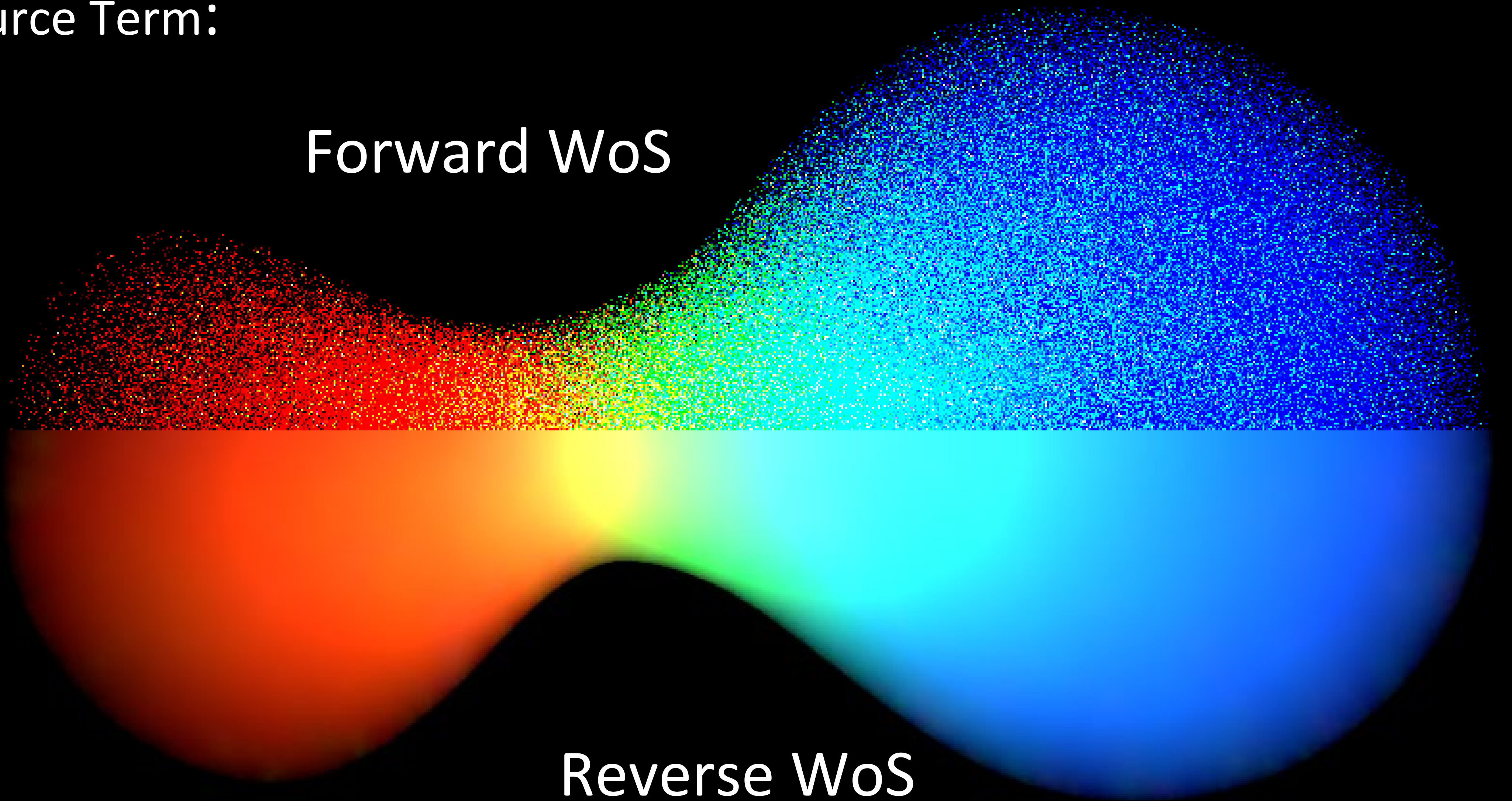


Final gather through forward walks

Equal Time Comparison:

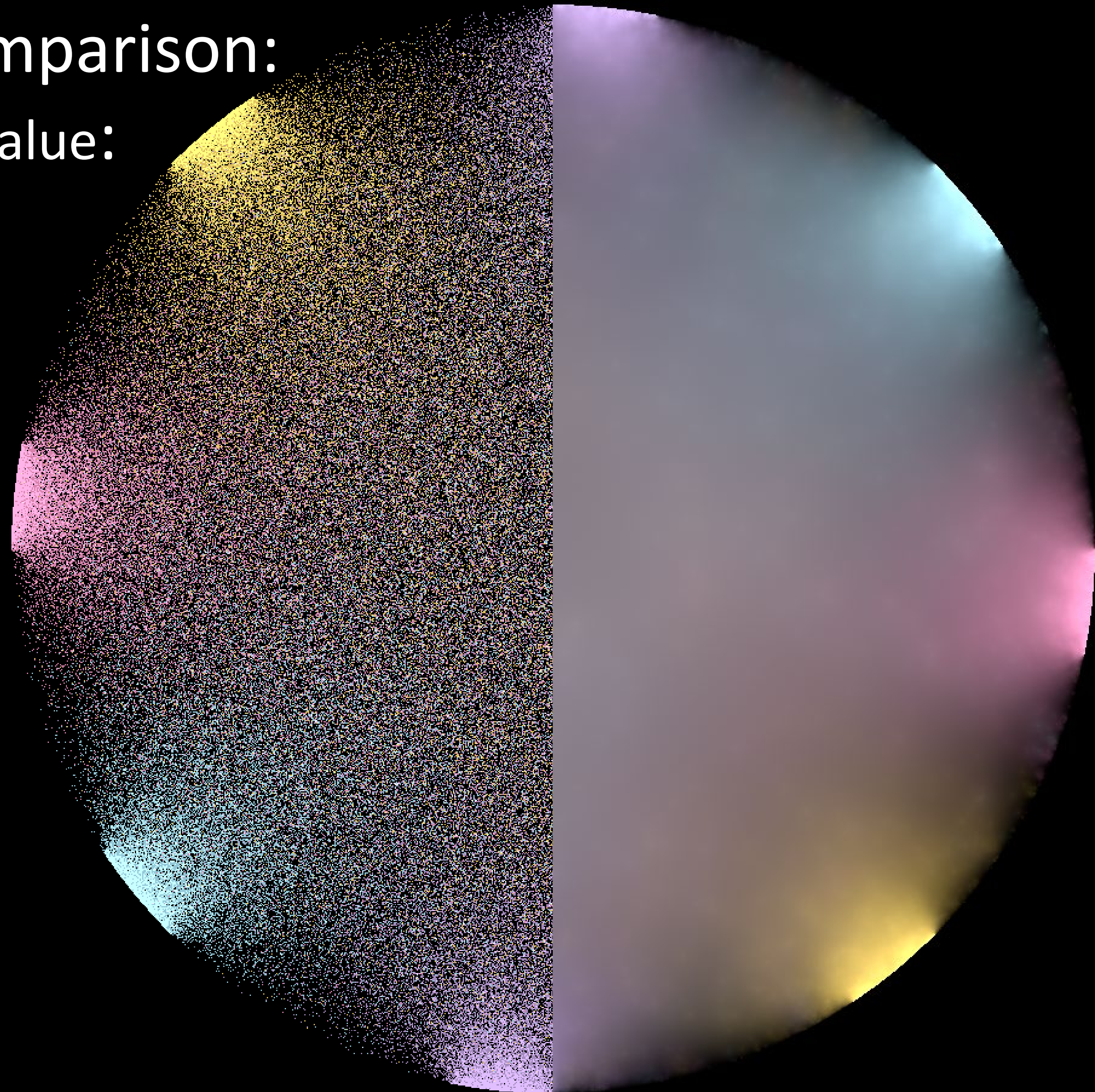
Sparse Source Term:

Forward WoS



Reverse WoS

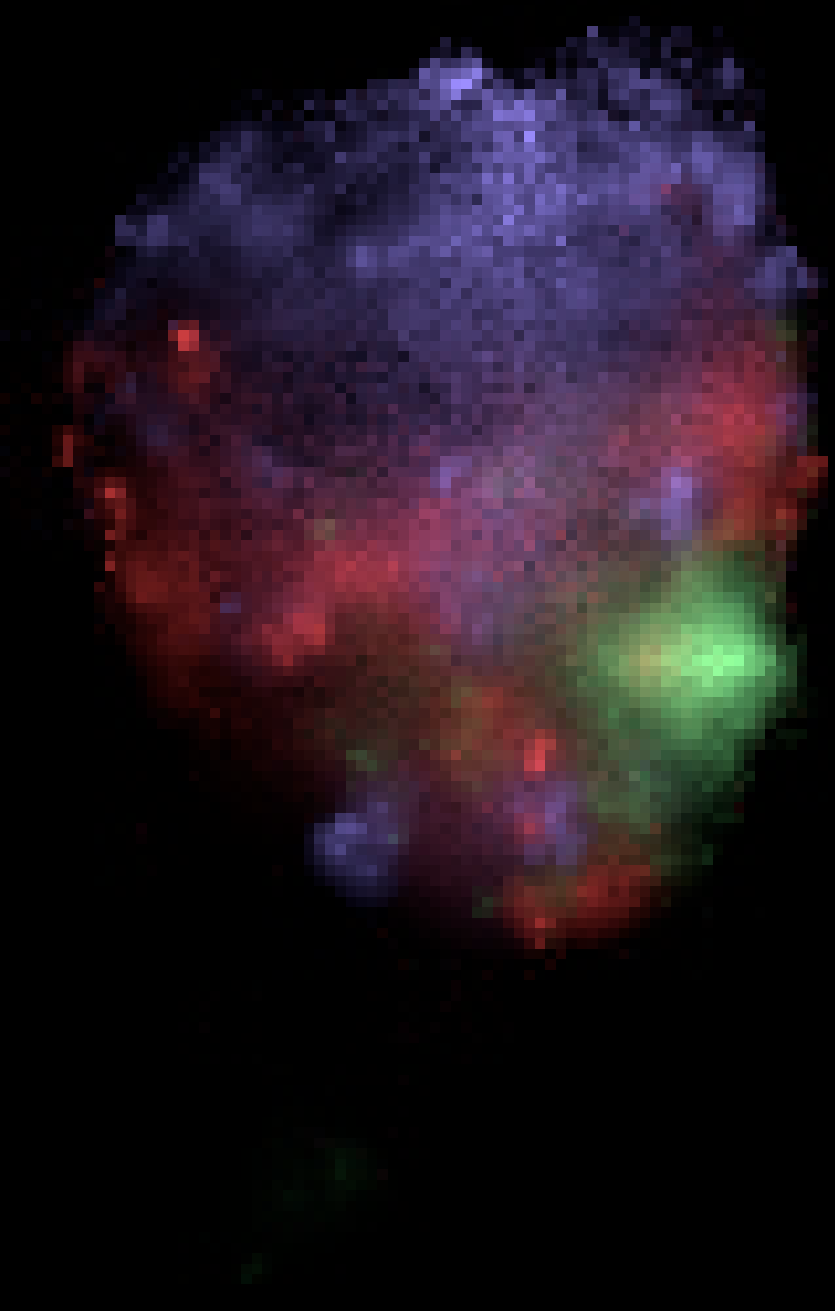
Equal Time Comparison:
Sparse Boundary Value:



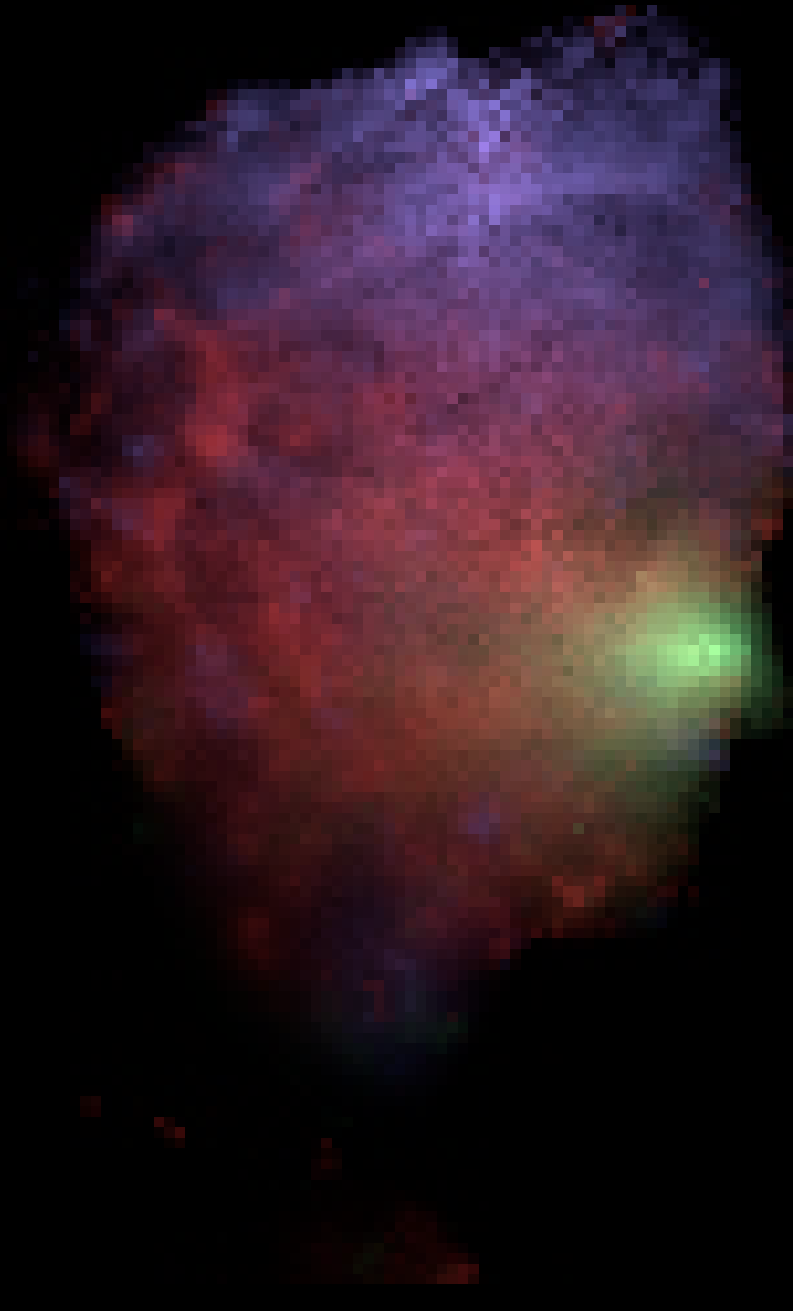
Forward WoS

Reverse WoS

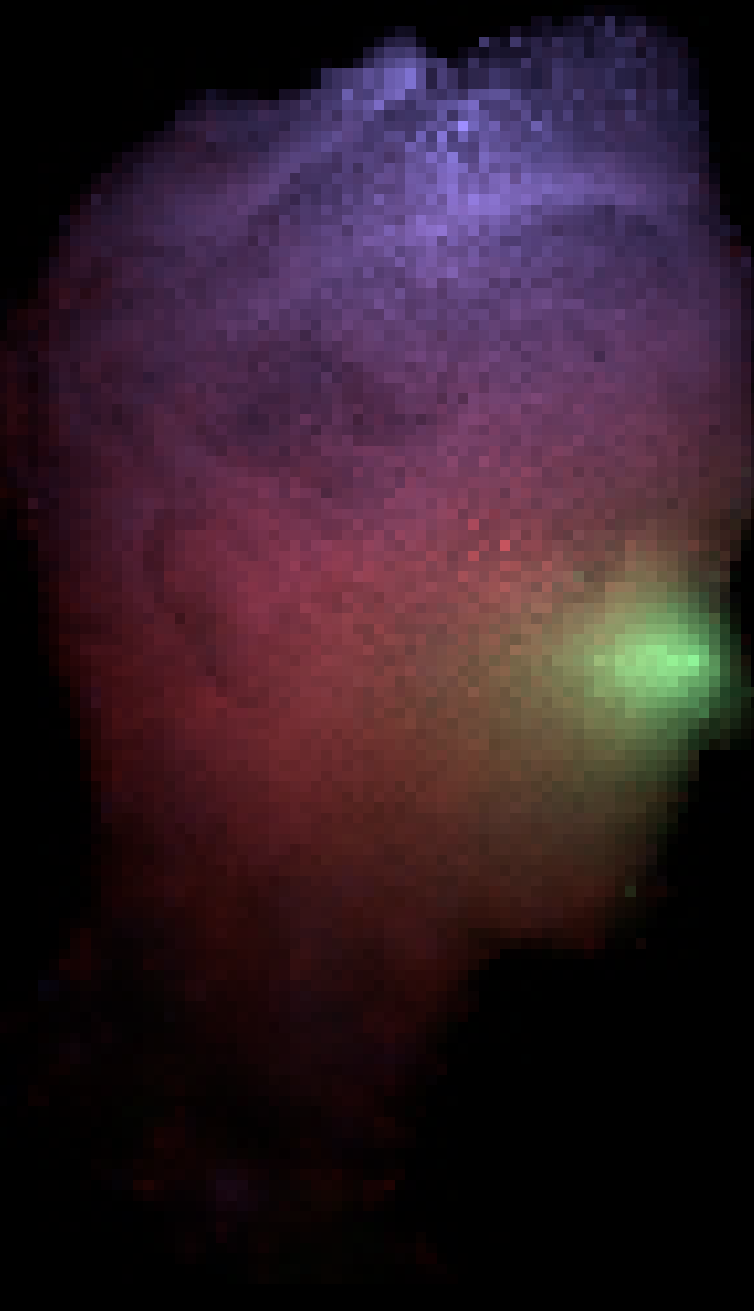
3D Diffusion:



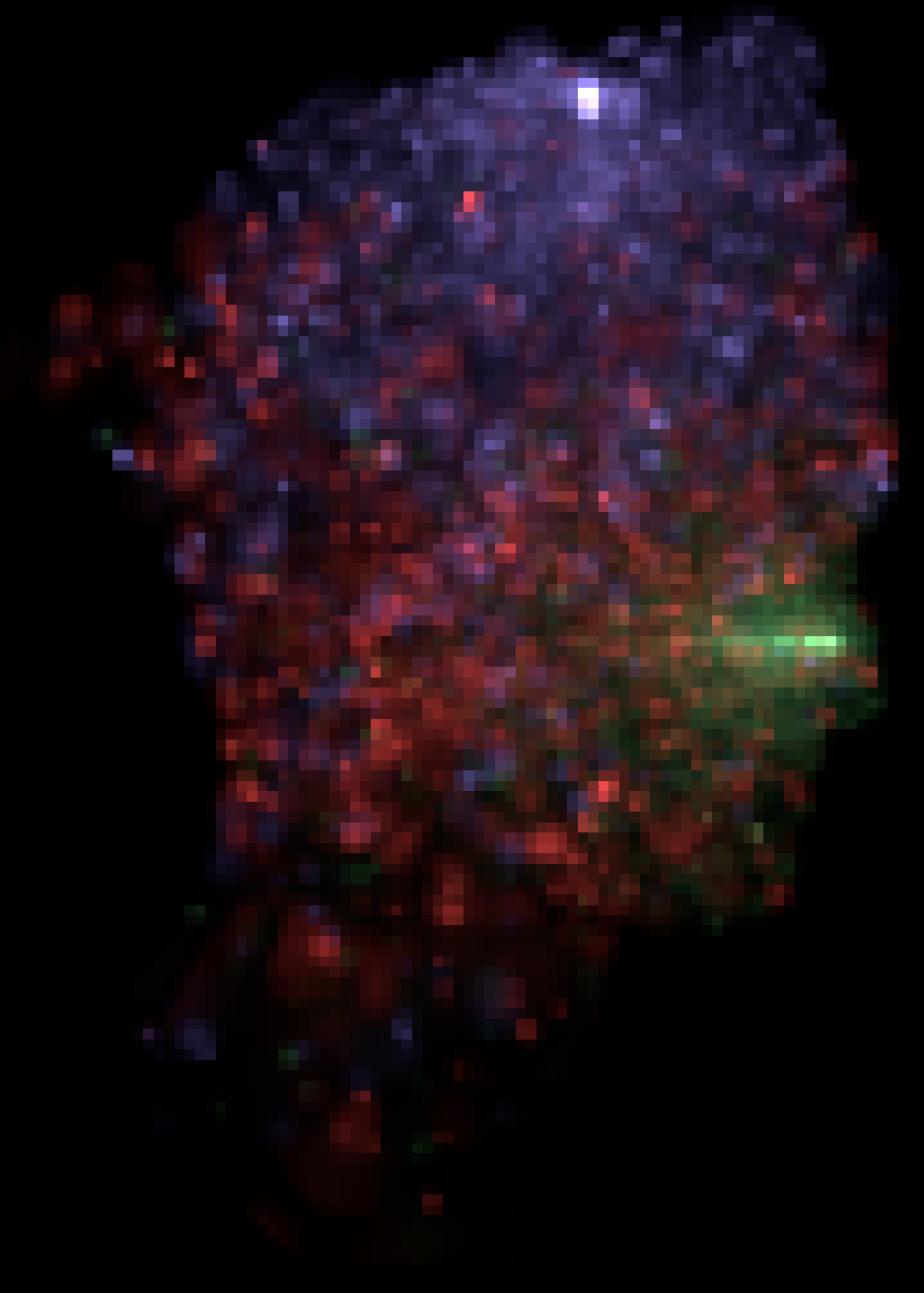
Ours(1000)



Ours(10000)



Ours(100,000)



VPM(100,000)

Other PDEs:

- Rendering equation
- Volume rendering equation

Shell Tracing:

- LEE, R. T. and O’SULLIVAN, C.
“Accelerated light propagation through participating media”.
- MULLER, T., PAPAS, M., GROSS, M., JAROSZ, W., and NOVAK, J.
“Efficient rendering of heterogeneous polydisperse granular media”.
- LEONARD, L., HOHLEIN, K., and WESTERMANN, R.
“Learning multiple-scattering solutions for sphere-tracing”.

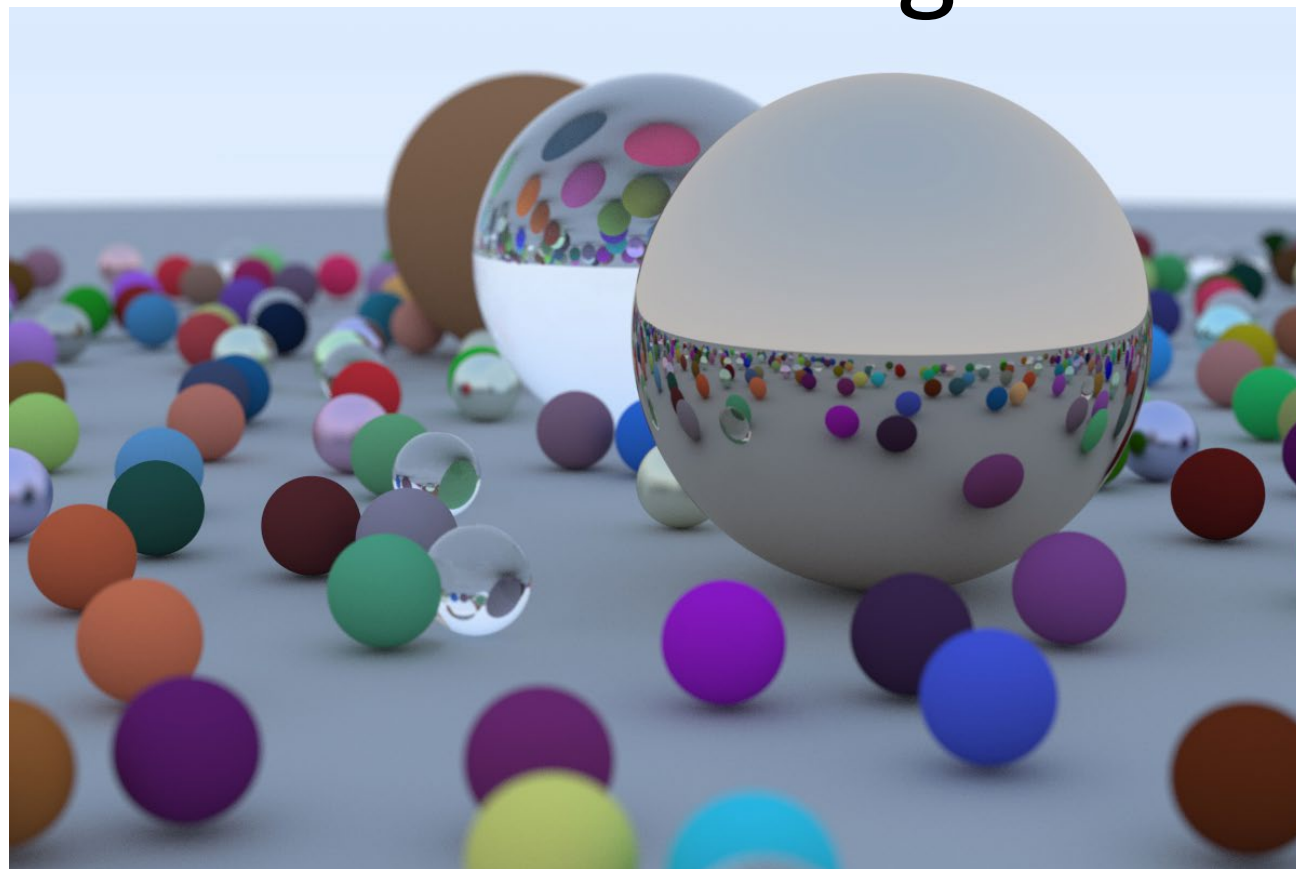


Other PDEs:

- Rendering equation
- Volume rendering equation
- Heat equation (time dependent)

Bidirectional Method

Forward Path Tracing



raytracing.github.io/books/RayTracingInOneWeekend.html

Bidirectional Path Tracing



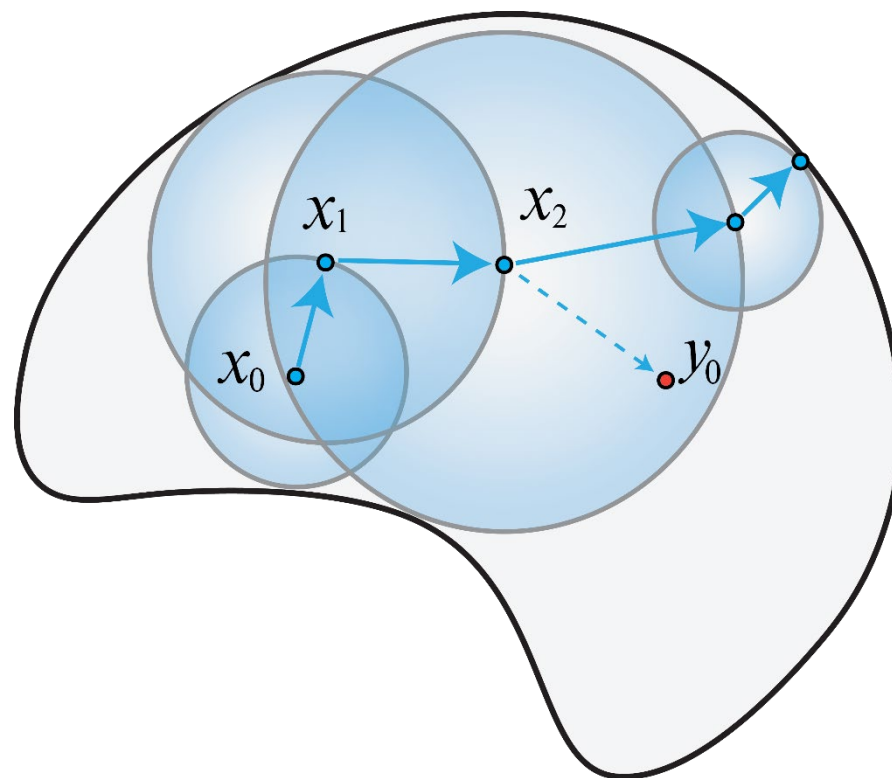
graphics.stanford.edu/papers/veach_thesis/thesis.pdf

Reverse Photon Mapping / VPLs



graphics.ucsd.edu/~henrik/papers/photon_map/

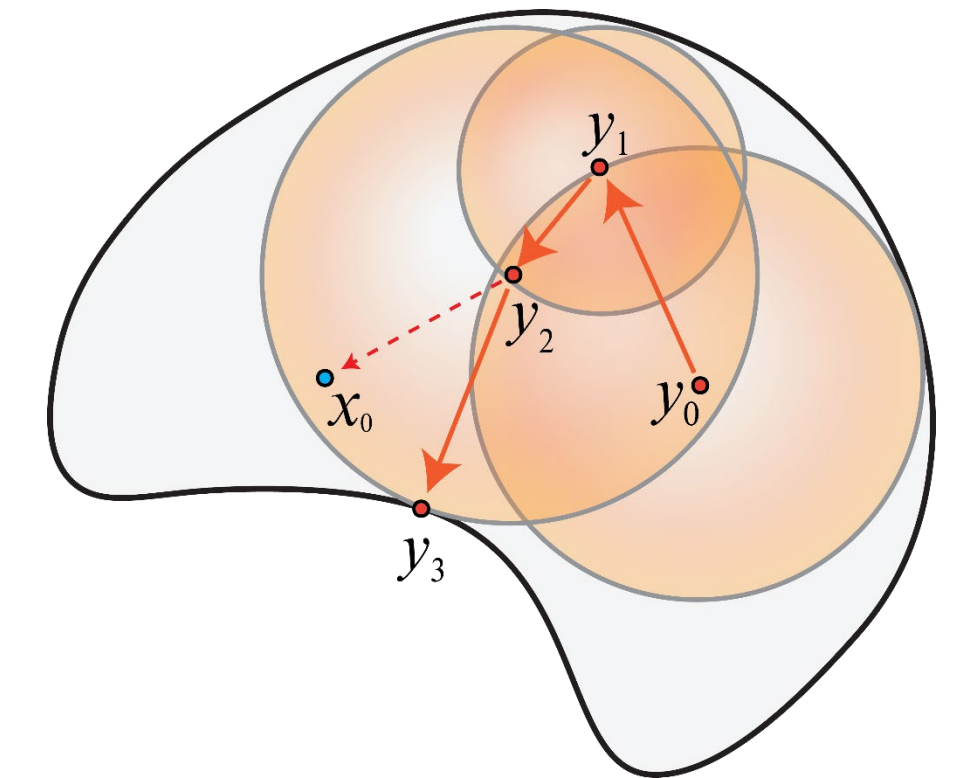
Forward WoS



Bidirectional WoS?



Reverse WoS



MIS

Different choices of forward & reverse steps lead to different path spaces

How to perform MIS between all different path spaces?

Thank you!