

# Lecture 27: Physics-informed deep neural networks

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## The Metropolis algorithm

# The Metropolis algorithm

$$p(x) = \frac{\pi(x)}{Z} ?$$

know the function

- Initialize:  $x_0$

- For  $n = 1, 2, \dots$

- **Generate** candidate sample:

$$x' \sim q(x' | x_{n-1})$$

proposed distribution

- **Calculate** the acceptance ratio:

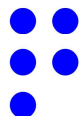
$$\alpha(x', x_{n-1}) = \min \left\{ \frac{\pi(x')}{\pi(x_{n-1})}, 1 \right\}$$

- **Accept/Reject:**

- Generate  $u \sim U([0, 1])$

- If  $u \leq \alpha$ , then accept:  $x_n \leftarrow x'$

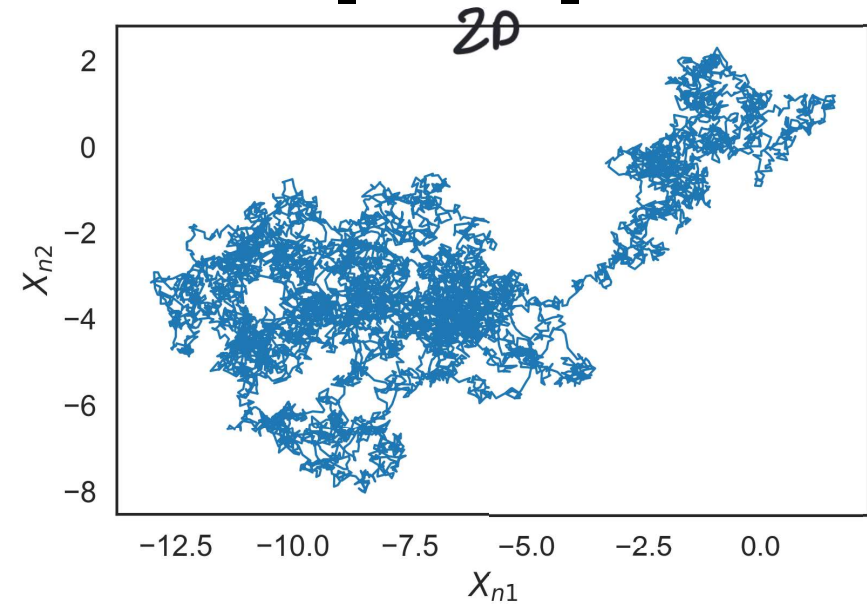
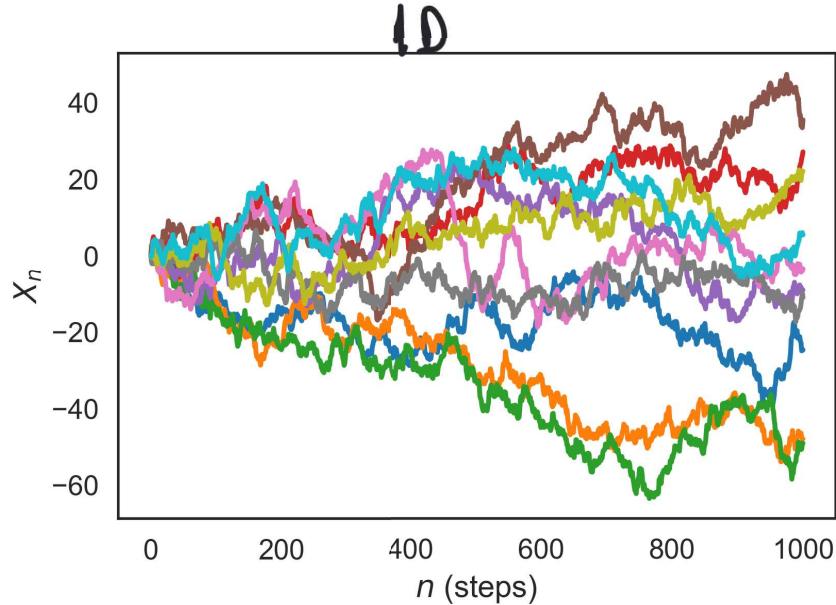
- If  $u > \alpha$ , then reject:  $x_n \leftarrow x_{n-1}$



# What are the restrictions on the Metropolis proposal?

$$q(x' | x_{n-1}) = q(x_{n-1} | x')$$

# The random walk proposal



$$q(x' | x_{n-1}) = \mathcal{N}(x' | x_{n-1}, \Sigma)$$

tunable parameter

# Why does the Metropolis algorithm work?

- Intuitively: It constructs a series of samples that eventually come from the desired probability density.
- Math reason: It constructs a stationary, reversible, aperiodic, Harris recurrent Markov chain that leaves the desired probability density invariant.
- Metropolis is an example of a Markov Chain Monte Carlo (MCMC) algorithm.