

# Alternatives to Monte Carlo Simulations for Fractal Diffusion Models

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# Why Focus on Fractal Diffusion Simulations?

- Fun
- Check validity of analytically derived models
- Obtain simulation-based dimension estimates
- Application in anomalous diffusion research

# Waypoints

- Fractal dynamics introduction
- Graph-based set representation
- Random walk models
- Constraint convolution schema
- Application: oscillation detection

# Diffusion on Fractals

## Mass Scaling

$$M \sim L^{d_f}$$

## Time Scaling

$$t \sim L^{d_w}$$

## Observables

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- Return probability

$$\Pr(X_t = \mathbf{x}_0) \sim t^{-d_s/2}$$

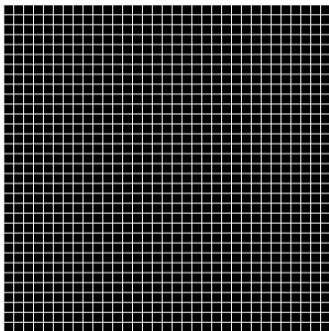
- Absolute moments

$$\mathbb{E} \|X_t - \mathbf{x}_0\|^\alpha \sim t^{\alpha/d_w}$$

# Grid Based Models

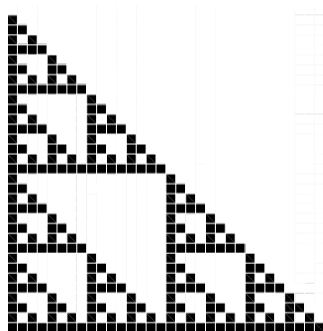
$$d = 2$$

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$$d_f = \log 3 / \log 2$$

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# Random Walk on Graph

$$\mathcal{G} = (\mathcal{F}, \mathcal{E}), \quad \mathcal{F} \subset \mathbb{Z}^d$$

$$p \in (0, 1), \quad c(\mathbf{x}) = \text{card } \mathcal{N}(\mathbf{x})$$

$$\Pr(X_{t+1} = \mathbf{y} \mid X_t = \mathbf{x}) = \begin{cases} p, & \mathbf{y} \in \mathcal{N}(\mathbf{x}) \\ 1 - c(\mathbf{x})p, & \mathbf{y} = \mathbf{x} \\ 0, & \text{otherwise} \end{cases}$$

$$\mathbf{x}, \mathbf{y} \in \mathcal{F}$$

# Standard Full Space Approach

- $P_t : \mathbb{Z}^d \mapsto [0, 1]$
- $P_t(\mathbf{x}) = \Pr(X_t = \mathbf{x} | X_0 = \mathbf{x}_0)$
- $P_0(\mathbf{x}) = 1(\mathbf{x} = \mathbf{x}_0)$

- $N_{\text{square}} = \begin{pmatrix} 0 & p & 0 \\ p & 1 - 4p & p \\ 0 & p & 0 \end{pmatrix}$

## Full Grid Probability Evolution

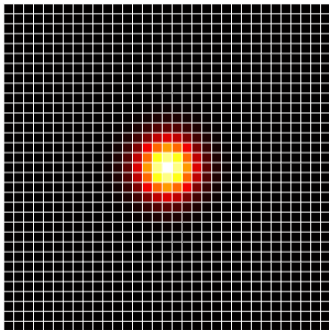
$$P_{t+1} = N * P_t$$

- $N_{\text{hexa}} = \begin{pmatrix} p & p & 0 \\ p & 1 - 6p & p \\ 0 & p & p \end{pmatrix}$

# Random Walk Over Grid Based Models

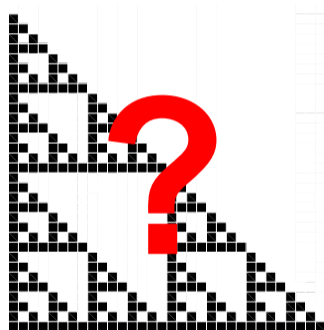
$$d = 2$$

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$$d_f = \log 3 / \log 2$$

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# Constrained Convolution Schema

- $P_t, F, C : \mathbb{Z}^d \mapsto \mathbb{R}$
- $P_t(\mathbf{x}) = \Pr(X_t = \mathbf{x} | X_0 = \mathbf{x}_0)$
- $P_0(\mathbf{x}) = 1(\mathbf{x} = \mathbf{x}_0)$
- $F(\mathbf{x}) = 1(\mathbf{x} \in \mathcal{F})$
- $C = F(N * F)$

- $N_{\text{square}} = \begin{pmatrix} 0 & 1 & 0 \\ 1 & 0 & 1 \\ 0 & 1 & 0 \end{pmatrix}$

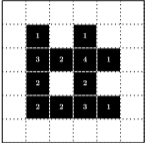
## Sparse Grid Probability Evolution

$$P_{t+1} = P_t(1 - pC) + pF(N * P_t)$$

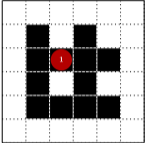
Proof is soon to be published

- $N_{\text{hexa}} = \begin{pmatrix} 1 & 1 & 0 \\ 1 & 0 & 1 \\ 0 & 1 & 1 \end{pmatrix}$

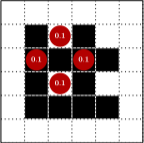
# Square Topology



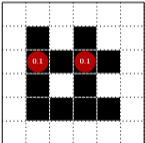
F, C



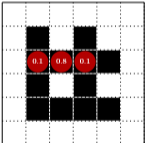
$P_0$



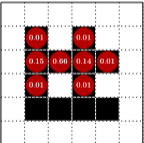
$\rho(P_0 * N)$



$\rho F(P_0 * N)$



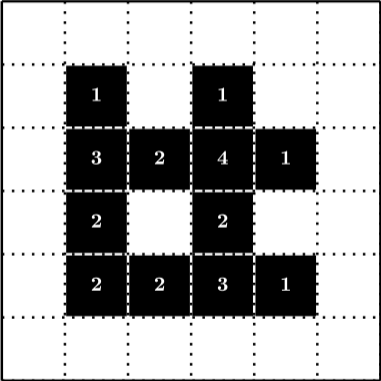
$P_1$



$P_2$

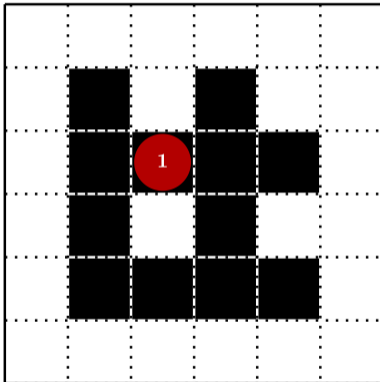
# Square Topology

- Set F with number of neighbours C



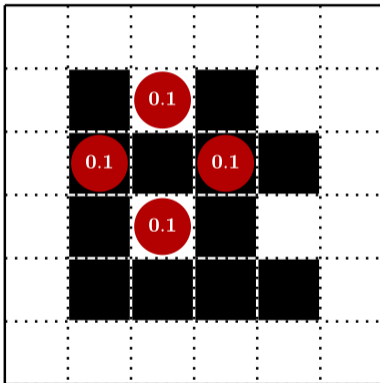
# Square Topology

- Initial distribution:  $P_0$



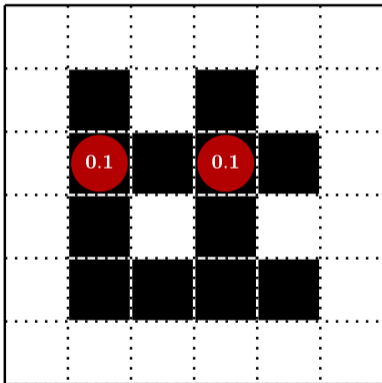
# Square Topology

- Intermediate step:  $p(P_0 * N), p = 0.1$



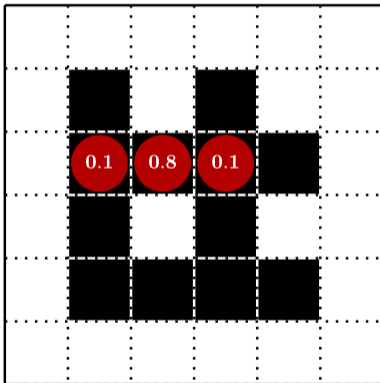
# Square Topology

- Intermediate step:  $\rho F (P_0 * N)$



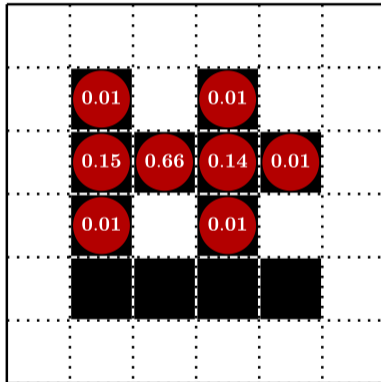
# Square Topology

- Distribution  $P_1 = P_0 (1 - \rho C) + \rho F (P_0 * N)$



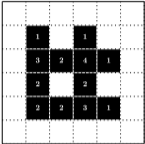
# Square Topology

- Distribution  $P_2$

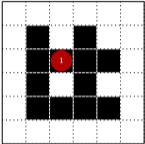




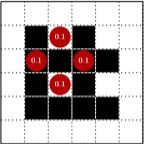
# Square Topology



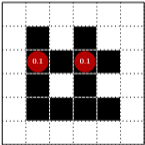
F, C



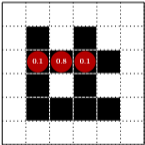
$P_0$



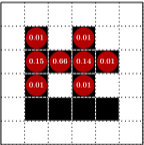
$\rho(P_0 * N)$



$\rho F(P_0 * N)$

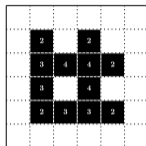


$P_1$

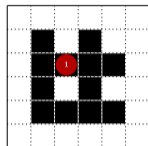


$P_2$

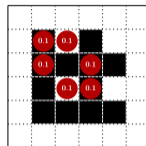
# Hexagonal Topology



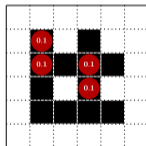
$F, C$



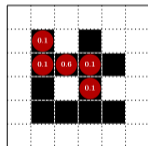
$P_0$



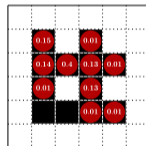
$\rho(P_0 * N)$



$\rho F(P_0 * N)$



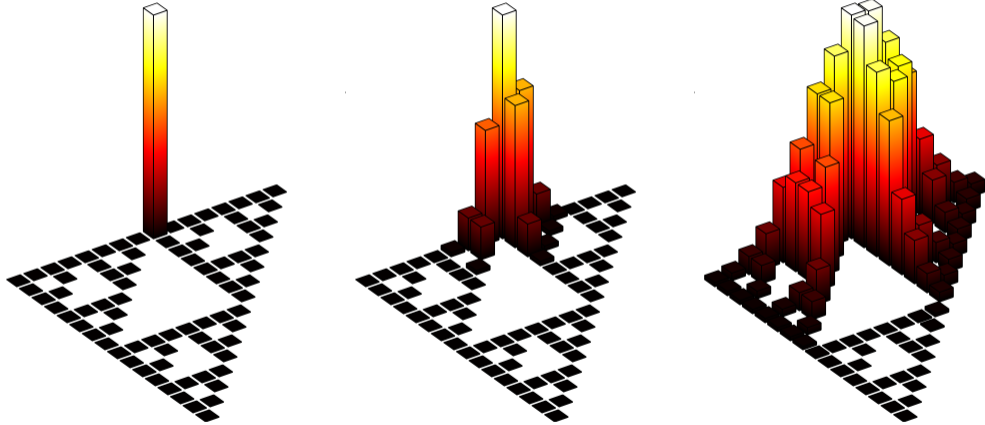
$P_1$



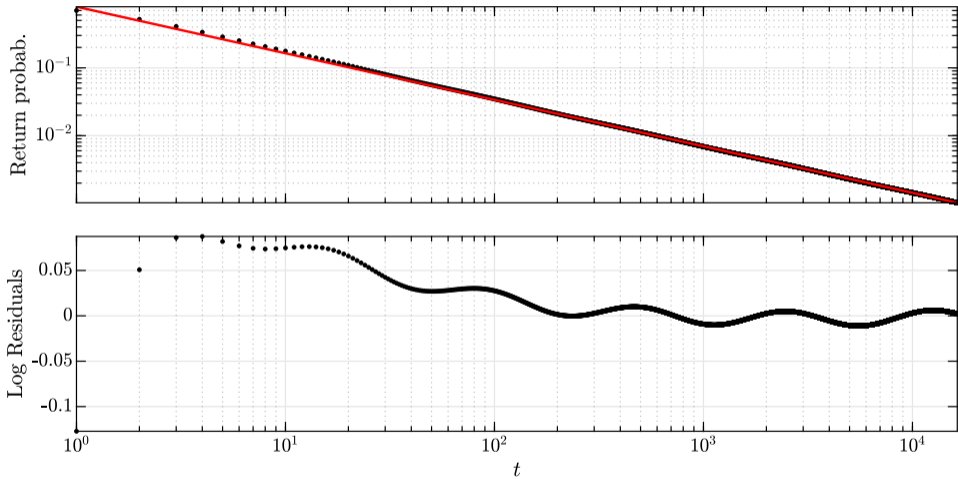
$P_2$

# Probability on Sierpinski Gasket

# Probability on Sierpinski Gasket



# Return Probability



- Random walk "simulation" without noise
- Calculation of full distribution  $P_t$
- Computationally friendly form
- Random walk analysis tool

Thank you for your attention!