

# Spatial Epidemics: Critical Behavior in Dimensions 1, 2, and 3

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## Background: Mean Field Models

Reed-Frost (SIR) Model  
Branching Envelopes  
Critical Behavior

## Spatial Epidemic Models

Spatial SIR Models  
Superprocess Limits  
Spatial Epidemic Models: Critical Scaling  
Branching Random Walk: Local Behavior  
Spatial Extent of SuperBM ( $d = 1$ )

## Reed-Frost (SIR) Model

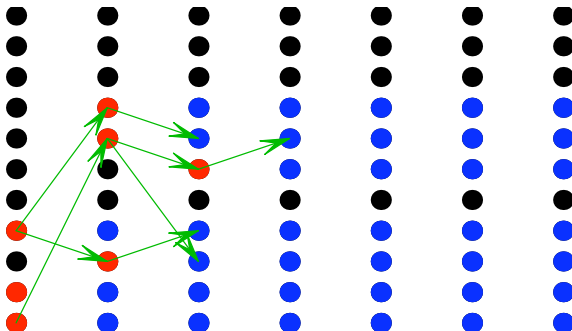
- ▶ Population Size  $N < \infty$
- ▶ Individuals **susceptible** (S), **infected** (I), or **recovered** (R).
- ▶ Recovered individuals immune from further infection.
- ▶ Infecteds recover in time 1.
- ▶ Infecteds infect susceptibles with probability  $p$ .

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**Critical Case:**  $p = 1/N$

## SIR Model: Example



## Reed-Frost and Random Graphs

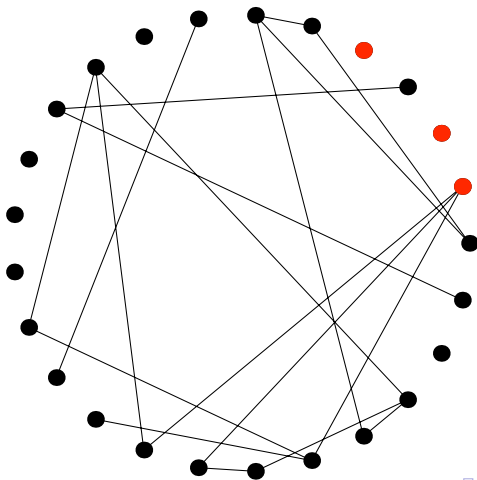
Reed-Frost model is equivalent to the Erdős-Renyi random graph model:

Individuals  $\longleftrightarrow$  Vertices

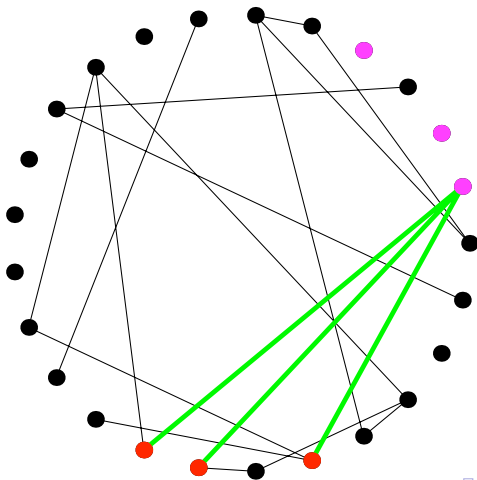
Infections  $\longleftrightarrow$  Edges

Epidemic  $\longleftrightarrow$  Connected Components

## Reed-Frost and Random Graphs

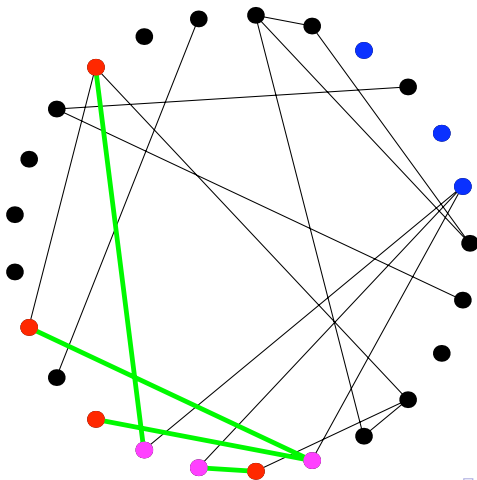


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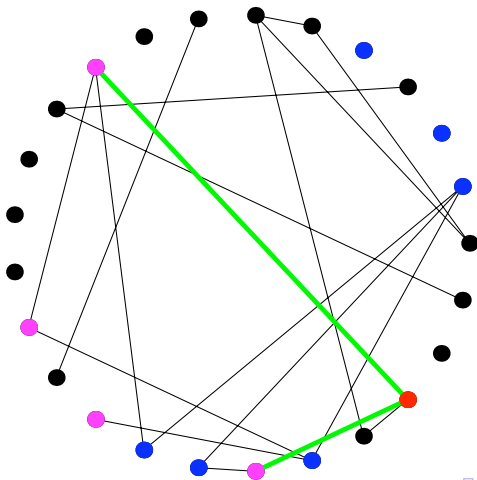




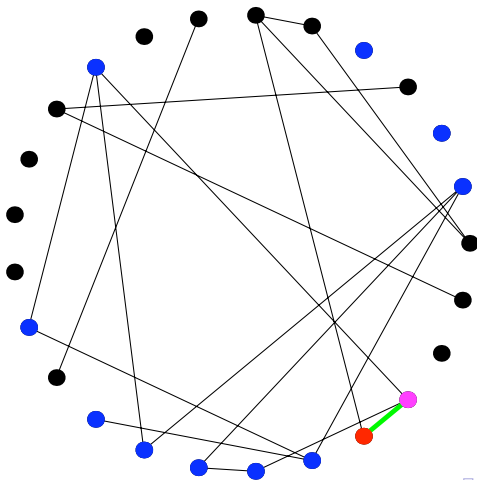
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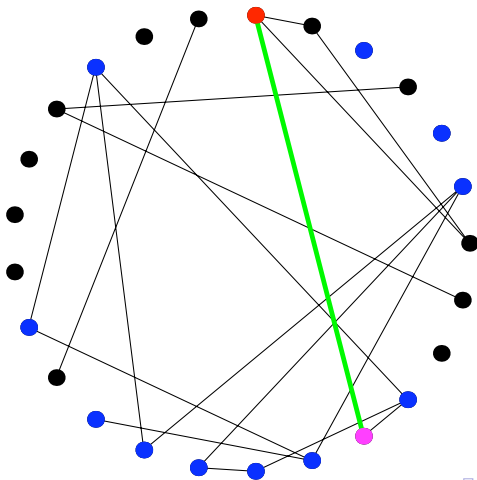
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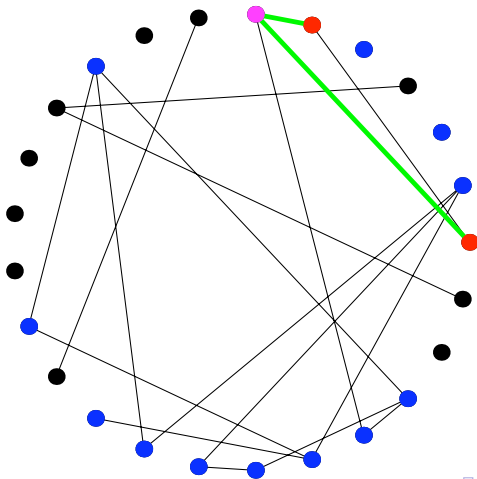
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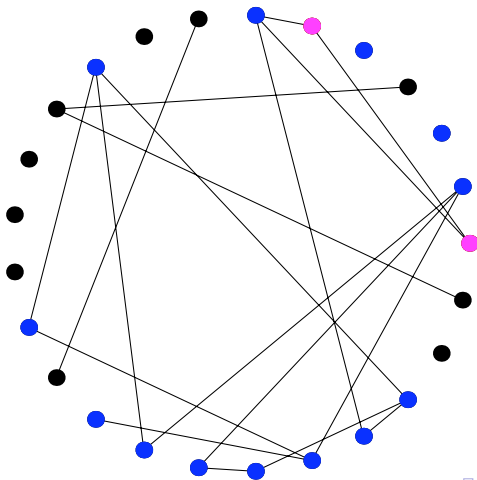
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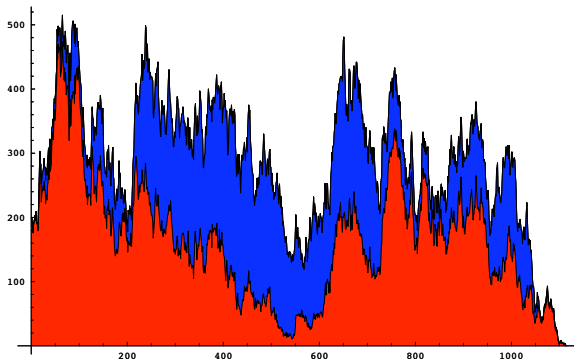
## Reed-Frost and Random Graphs



## Branching Envelope of an Epidemic

- ▶ Each epidemic has a **branching envelope** (GW process)
- ▶ Offspring distribution: Binomial- $(N, p)$
- ▶ Epidemic is dominated by its branching envelope
- ▶ When  $I_t \ll S_t$ , infected set grows  $\approx$  branching envelope

## Example: SIS Epidemic and its Branching Envelope



$$N = 80000$$
$$I_0 = 200$$
$$p = 1/80000$$



## Critical Behavior: Reed-Frost (SIR) Epidemics

- ▶ # Infected in Generation  $t := I_t^N$
- ▶ # Recovered in Generation  $t := R_t^N$
- ▶ Initial Condition:  $I_0^N \sim bN^\alpha$

**Theorem:** (Dolgoarshinnykh & L.) As population size  $N \rightarrow \infty$ ,

$$\begin{pmatrix} N^{-\alpha} I_t^N \\ N^{-2\alpha} R_t^N \end{pmatrix} \xrightarrow{\mathcal{D}} \begin{pmatrix} I(t) \\ R(t) \end{pmatrix}$$

The limit process satisfies  $I(0) = b$  and

$$dR(t) = I(t) dt$$

$$dI(t) = +\sqrt{I(t)} dB_t \quad \text{if } \alpha < 1/3$$

$$dI(t) = +\sqrt{I(t)} dB_t - I(t)R(t) dt \quad \text{if } \alpha = 1/3$$

## Critical Behavior: Reed-Frost (SIR) Epidemics

- ▶ Population size:  $N \rightarrow \infty$
- ▶ # Infected in Generation  $t$ :  $I_t^N$
- ▶ # Recovered in Generation  $t$ :  $R_t^N$
- ▶ Initial Condition:  $I_0^N \sim bN^\alpha$

Corollary: If  $\alpha = 1/3$  then

$$R_\infty^N / N^{2/3} \implies \tau(b)$$

where  $\tau(b) =$  first passage time of  $B(t) + t^2/2$  to  $b$ .

(Martin-Lof; Aldous)

## Critical Behavior: Heuristics

- ▶ Critical Epidemic with  $I_0 = m$  should last  $\approx m$  generations.
- ▶ Number  $R_t$  recovered should be  $\approx m^2$ .
- ▶ Offspring in branching envelope :: attempted infections.
- ▶ **Misfires**: Infections of immunes not allowed.
- ▶ **Critical Threshold**: # misfires/generations  $\approx O(1)$

### Critical SIR Epidemic:

$$E(\# \text{misfires in generation } t + 1) \approx I_t R_t / N$$

so there will be observable deviation from branching envelope when

$$I_t \approx N^{1/3} \quad \text{and} \quad R_t \approx N^{2/3}$$

## Spatial SIR Epidemic:

- ▶ Villages  $V_x$  at Sites  $x \in \mathbb{Z}^d$
- ▶ Village Size:= $N$
- ▶ Nearest Neighbor Disease Propagation
- ▶ SIR Rules **Locally**:
  - ▶ Infected individuals infect susceptibles at **same or neighboring site** with probability  $p_N$
  - ▶ Infecteds recover in time 1.
  - ▶ Recovered individuals immune from further infection.

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**Critical Case:** Infection probability  $p_N = 1/((2d + 1)N)$ .

## Percolation Representation

Spatial SIR epidemic is equivalent to **critical bond percolation** on the graph  $G_N := K_N \times \mathbb{Z}^d$  with nearest neighbor connections:

- ▶ Vertex set  $[N] \times \mathbb{Z}^d$
- ▶ Edges connect vertices  $(i, x)$  and  $(j, y)$  if  $\text{dist}(x, y) \leq 1$

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**Problem:** At critical point  $p = 1/(2d + 1)N$ ,

- ▶ How does connectivity probability decay?
- ▶ How does size of largest connected cluster scale with  $N$ ?
- ▶ Joint distribution of largest, 2nd largest,  $\dots$  ?

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**Conjecture:** Let  $R_N =$  maximum  $x$  such that a vertex  $(j, x)$  is connected to a vertex  $(i, 0)$ . Then  $R_N/N^{4/(6-d)}$  converges in distribution as  $N \rightarrow \infty$  in dimensions  $d = 1, 2, 3$ .



## Branching Envelope

The branching envelope of a spatial SIR epidemic is a **branching random walk**: In each generation,

- ▶ A particle at  $x$  puts offspring at  $x$  or neighbors  $x + e$ .
- ▶ #Offspring are independent Binomial- $(N, p_N)$  or Poisson- $Np_N$
- ▶ **Critical BRW** :  $p = p_N = 1/((2d + 1)N)$ .

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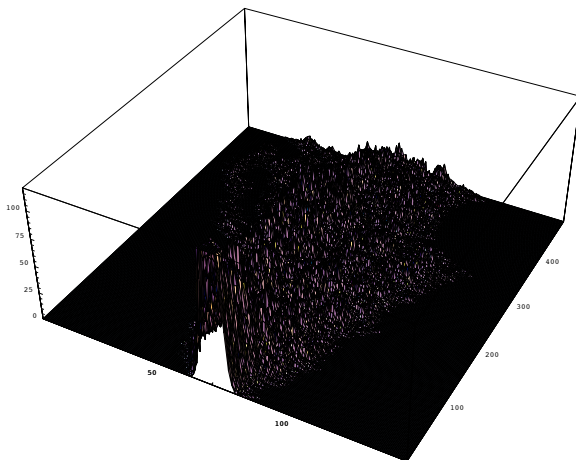
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## Associated Measure-Valued Processes

$$X_t^M = X_t^{M,N}:$$

measure that puts mass  $1/M$  at  $x/\sqrt{M}$  for each particle at site  $x$  at time  $t$ .

# Critical Spatial SIS Epidemic: Simulation



## Watanabe's Theorem

Let  $X_t^M$  be the measure-valued process associated to a critical nearest neighbor branching random walk. If

$$X_0^M \Longrightarrow X_0$$

then

$$X_{Mt}^M \Longrightarrow X_t$$

where  $X_t$  is the **Dawson-Watanabe** process (**superBM**). The DW process is a **measure-valued diffusion**.

**Note 1:** The total mass  $\|X_t\|$  is a Feller diffusion.

**Note 2:** Watanabe is the spatial analogue of Feller.

**Note 3:** In  $1D$ ,  $X_t$  has a continuous density  $X(t, x)$ .

## Scaling Limits: SIR Spatial Epidemics in $d = 1$

**Recall:**  $X_t^{M,N}$  is the measure that puts mass  $1/M$  at  $x/\sqrt{M}$  for each infected individual at site  $x$  at time  $t$ .

**Theorem:** Assume that the epidemic is critical and that  $M = N^\alpha$ . If  $X_0^{M,N} \Rightarrow X_0$  then

$$X_{Mt}^{M,N} \Longrightarrow X_t$$

where

- ▶ If  $\alpha < 2/5$  then  $X_t$  is the Dawson-Watanabe process.
- ▶ If  $\alpha = 2/5$  then  $X_t$  is the Dawson-Watanabe process with killing rate

$$\int_0^t X(s, x) ds$$

## Scaling Limits: SIR Spatial Epidemics in $d = 2, 3$

**Recall:**  $X_t^{M,N}$  is the measure that puts mass  $1/M$  at  $x/\sqrt{M}$  for each infected individual at site  $x$  at time  $t$ .

**Theorem:** Assume that the epidemic is critical and that  $M = N^\alpha$ . If  $X_0^{M,N} \Rightarrow X_0$  and  $X_0$  satisfies a smoothness condition then

$$X_{Mt}^{M,N} \Longrightarrow X_t$$

where

- ▶ If  $\alpha < 2/(6 - d)$  then  $X_t$  is the Dawson-Watanabe process.
- ▶ If  $\alpha = 2/(6 - d)$  then  $X_t$  is the Dawson-Watanabe process with killing rate  $L(t, x) =$  Sugitani local time density.

## Sugitani's Local Time

**Theorem:** Assume that  $d = 2$  or  $3$  and that the initial configuration  $X_0 = \mu$  of the super-BM  $X_t$  satisfies

**Smoothness Condition:**

$$\int_0^t \int_{x \in \mathbb{R}^d} \phi_t(x - y) d\mu(y)$$

is jointly continuous in  $t, x$ , where  $\phi_t(x)$  is the heat kernel (Gaussian density). Then for each  $t \geq 0$  the occupation measure

$$L_t := \int_0^t X_s ds$$

is absolutely continuous with jointly continuous density  $L(t, x)$ .

## Critical Scaling: Heuristics (SIR Epidemics, $d = 1$ )

- ▶ # Infected Per Generation:  $\approx M$
- ▶ Duration:  $\approx M$  generations.
- ▶ # Infected Per Site:  $\approx \sqrt{M}$
- ▶ # Recovered Per Site:  $\approx M\sqrt{M}$
- ▶ # Misfires Per Site:  $\approx M^2/N$
- ▶ # Misfires Per Generation:  $\approx M^{5/2}/N$

So if  $M \approx N^{2/5}$  then # Misfires Per Generation  $\approx 1$ .



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So if  $M \approx N^{2/5}$  then # Misfires Per Generation  $\approx 1$ .

But how do we know that the infected individuals in generation  $n$  don't "clump"?

## Critical Scaling: Heuristics (SIR Epidemics, $d = 3$ )

- ▶ # Infected Per Generation  $\approx$  Duration  $\approx M$
- ▶ # Sites Reachable  $\approx M^{3/2}$ .
- ▶ # Infected Per Infected Site:  $\approx O(1)$
- ▶ # Recovered Per Site:  $\approx M^2 / M^{3/2} = \sqrt{M}$
- ▶ # Misfires Per Generation:  $\approx M \times \sqrt{M} / N$

So if  $M \approx N^{2/3}$  then # Misfires Per Generation  $\approx 1$ .

## Proof Strategy I

**Lemma:** Assume that  $L_n$  and  $L$  are likelihood ratios under  $P_n$  and  $P$ , and define  $Q_n$  and  $Q$  by

$$\begin{aligned}dQ_n &= L_n dP_n, \\dQ &= L dP.\end{aligned}$$

Assume that  $X_n$  and  $X$  are random variables whose distributions under  $P_n$  and  $P$  satisfy

$$(X_n, L_n) \implies (X, L).$$

Then the  $Q_n$ -distribution of  $X_n$  converges weakly to the  $Q$ -distribution of  $X$ .

## Proof Strategy II

**Theorem:** (Dawson) The law  $Q$  of the Dawson-Watanabe process with location-dependent killing rate  $\theta(x, t)$  is mutually a.c. relative to the law  $P$  of the Dawson-Watanabe process with no killing (superBM), and the likelihood ratio is

$$dQ/dP = \exp \left\{ - \int \theta(t, x) dM(t, x) - \frac{1}{2} \int \langle X_t, \theta(t, \cdot)^2 \rangle dt \right\}$$

where  $M$  is the orthogonal martingale measure attached to the superBM  $X_t$ .

## Proof Strategy III

$P^M$  = Law of  $M$ th branching random walk.

$Q^{M,N}$  = Law of corresponding spatial epidemic.

$$\frac{dQ^{M,N}}{dP^M} = \prod_{\text{time } t} \prod_{\text{sites } x} \left( 1 + R^{M,N}(t, x) \right)$$

where  $R^{M,N}(t, x)$  is a function of the number of misfires at site  $x$  at time  $t$ . So the problem is to show that under  $P^M$ , as  $M \rightarrow \infty$ ,

$$\sum_t \sum_x R^{M,N}(t, x)$$

converges to the exponent in Dawson's likelihood ratio.

## Local Behavior for Branching Random Walk: $d = 1$

$Y_n^k(\cdot) =$  branching random walk on  $\mathbb{Z}$  with Poisson-1 offspring distribution and initial state  $Y_0^k$  scaling as in Watanabe's theorem.

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**Theorem:** If  $Y_0^k([\sqrt{k}x]) \rightarrow Y_0(x)$  where  $Y_0(x)$  is continuous with compact support then

$$\frac{Y_{kt}^k([\sqrt{k}x])}{\sqrt{k}} \Longrightarrow X(t, x)$$

where  $X(t, x)$  is the Dawson-Watanabe density process.

## Local Time for Branching Random Walk: $d = 2, 3$

$Y_n^k(\cdot) =$  branching random walk on  $\mathbb{Z}^d$  with Poisson-1 offspring distribution and initial state  $Y_0^k$  scaling as in Watanabe's theorem.

$$U_n^k(\cdot) = \sum_{i=0}^t Y_i^k(\cdot)$$



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**Theorem:** If  $Y_0^k([\sqrt{k}x]) \rightarrow Y_0(x)$  where  $Y_0$  satisfies hypotheses of Sugitani then in  $d = 2, 3$ ,

$$\frac{U_{kt}^k([\sqrt{k}x])}{k^{2-d/2}} \Longrightarrow L(t, x)$$

where  $L(t, x)$  is Sugitani local time.

## Occupation Statistics in $d = 2$

For branching random walk with Poisson-1 offspring distribution and nearest neighbor steps initiated by a single particle at  $(0, 0)$ , let

$$G_n = \{\text{process survives to generation } n\}$$

$$G_n^0 = \{\exists \text{ particle at origin in generation } n\}$$

$$\Omega_n = \#\{\text{occupied sites in generation } n\}$$

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**Theorem:** Conditional on  $G_n$ , the number  $\Omega_n$  of occupied sites is  $O_p(n/\log n)$ , that is, the conditional distributions of  $\Omega_n \log n/n$  are tight.

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**Theorem:** Conditional on  $G_n^0$ , the trajectory of a particle randomly chosen from those particles at the origin in generation  $n$  **is not** a pinned random walk.

## Spatial Extent of Super-BM in $d = 1$

- ▶  $X_t =$  Dawson-Watanabe process,
- ▶  $\mathcal{R} := \bigcup_{t \geq 0} \text{support}(X_t)$
- ▶  $u_D(x) := -\log P(\mathcal{R} \subset D \mid X_0 = \delta_x)$

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**Theorem (Dynkin):** For any finite interval  $D$ ,  $u_D(x)$  is the maximal nonnegative solution in  $D$  of the differential equation

$$u'' = u^2$$

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**Solution:** Weierstrass  $\mathcal{P}$ -Function

$$u_D(x) = \mathcal{P}_L(x/\sqrt{6}) = \frac{1}{6x^2} + \sum_{\omega \in L^*} \left\{ \frac{1}{6(x-\omega)^2} - \frac{1}{6\omega^2} \right\}$$

where the **period lattice**  $L$  is generated by  $Ce^{\pi i/3}$  for  $C > 0$  depending on  $D = [0, a]$  as follows:

$$C = \sqrt{6}a$$

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**General Initial Conditions:** For any finite Borel measure  $\mu$  with support  $\subset D$ ,

$$\begin{aligned} -\log P(\mathcal{R} \subset D \mid X_0 = \mu) &= \int u_{D(x)} \mu(dx) \\ &= \int \mathcal{P}_L(x/\sqrt{6}) \mu(dx) \end{aligned}$$