

# Chaos and Entropy of a Lattice Gas Cellular Automaton

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# Foundations of statistical physics

*... the **essential** features of the evolution do not depend on specific dynamical properties such as **positivity of the Lyapunov exponents**, ergodicity or mixing...*

*J. Lebowitz, Boltzmann's entropy and time's arrow, Phys. Today 1973*

# A simple stochastic model

- $M$  *macroscopic* configurations  $x_1, \dots, x_M$ .
- $P(x) = 1/M$  (microcanonical distribution).  $S = \log(M)$ .
- Coarse-grained stochastic dynamics:

$$W(x'|x) = \begin{cases} 1/\alpha & \text{for } \alpha M \text{ entries;} \\ 0 & \text{otherwise.} \end{cases}$$

- $W$  is irreducible.

# Entropy and Lyapunov exponents

The Kolmogorov-Sinai entropy (information production in the coarse-grained dynamics) is

$$K = - \sum_x P(x) \sum_{x'} W(x'|x) \log (W(x'|x)) = \log(\alpha) + \log(M).$$

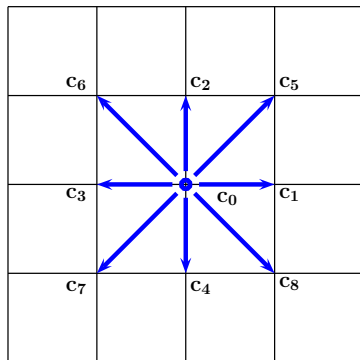
- $K$  is the sum of positive Lyapunov exponents.
- We can assume  $K \propto \lambda_0$ .
- Since  $\log(M) = S$  we get

$$\lambda_0 \propto \log(\alpha) + S.$$

# A multiparticle model: LGCA

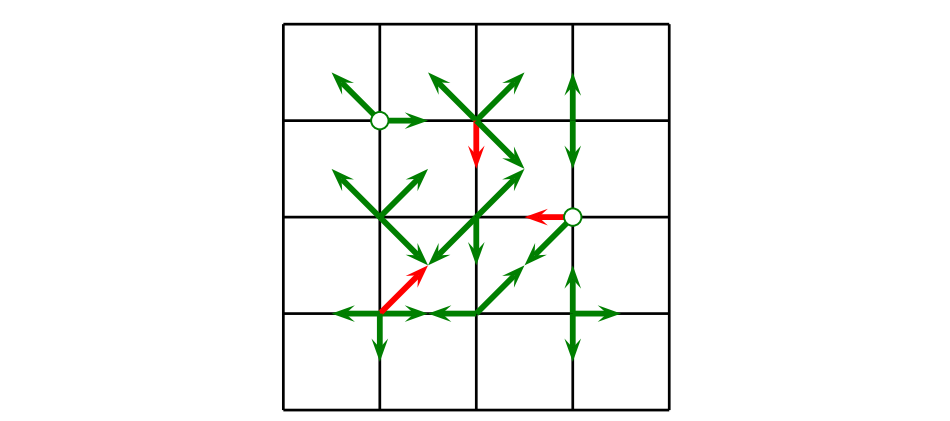
- Lattice gas cellular automata (**LGCA**) are discrete, deterministic dynamical models.
- Hydrodynamical behavior.
- Multispeed LGCA have **equilibrium thermodynamics**.
- It is possible to define the **largest Lyapunov exponent**  $\lambda_0$ .

## The $D2Q9$ LGCA: Discrete space and velocities

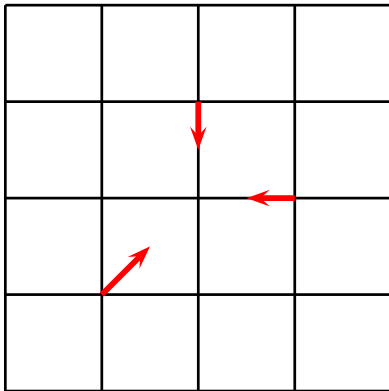


$$c_i = \begin{cases} 0 & i = 0, \\ 1 & i = 1, 2, 3, 4, \\ \sqrt{2} & i = 5, 6, 7, 8. \end{cases}$$

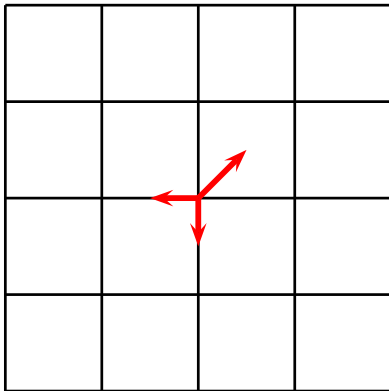
## Streaming and collision



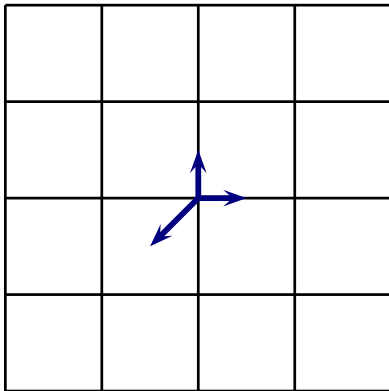
# Streaming and collision



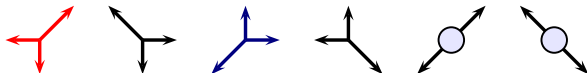
# Streaming and collision



# Streaming and collision



# Multiple collisions



Exclusion principle

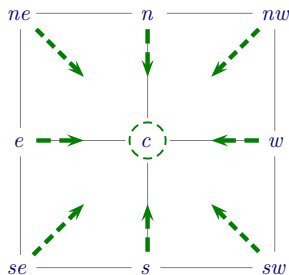
$$s(\mathbf{r}, t) = (s_8, s_7, \dots, s_0), \quad s_k = 0, 1.$$

$b$  states cellular automaton

$$s(\mathbf{r}, t) = \sum_{k=0}^8 s_k 2^k, \quad 0 \leq s(\mathbf{r}, t) < 2^9 = b,$$

$$s(\mathbf{r}, t+1) = \mathbf{CS}(\{s(\mathbf{r}', t)\}_{\mathbf{r}' \in \mathcal{V}(\mathbf{r})}).$$

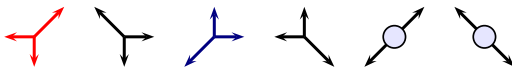
# Streaming and collision



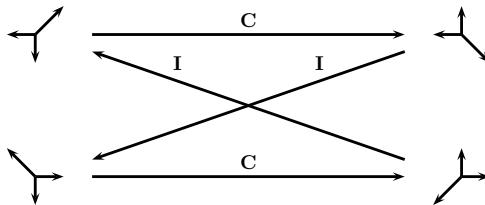
$$s_k(\mathbf{r}, t + 1) = C_k(s_0(c, t), s_8(ne, t), s_4(n, t), s_7(nw, t), \\ s_3(w, t), s_6(sw, t), s_2(s, t), s_5(se, t), s_1(s, t)) \\ k = 0, \dots, 8$$

# The $D2Q9$ LGCA

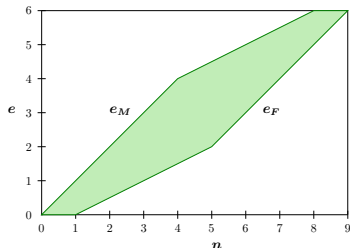
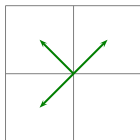
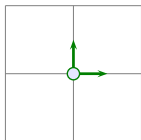
- Deterministic  $D2Q9$  LGCA.



- Reversible  $D2Q9$  LGCA.



# Fermi energy $e_F$ and maximum energy $e_M$



$$e_F = \begin{cases} 0 & 0 \leq n < 1, \\ (n-1)/2 & 1 \leq n < 5, \\ n-3 & 5 \leq n \leq 9, \end{cases}$$

$$e_M = \begin{cases} n & 0 \leq n < 4, \\ (n+4)/2 & 4 \leq n < 8, \\ 6 & 8 \leq n \leq 9. \end{cases}$$

# Boolean cellular automaton

$$s(\mathbf{r}) = (s_8(\mathbf{r}), \dots, s_0(\mathbf{r})), \quad s_k = 0, 1, \quad k = 0, \dots, 8,$$

$$\mathbf{s} = (s(\mathbf{r}_0), \dots, s(\mathbf{r}_{L-1})),$$

$$\mathbf{r}_m \in \Lambda, m = 0, \dots, L, \quad L = |\Lambda|$$

$$\mathbf{s} = (s_0, s_1, \dots, s_{\mathcal{L}-1}), \quad s_n = 0, 1,$$

$$n = 0, \dots, \mathcal{L} - 1, \quad \mathcal{L} = 9L, \quad \mathbf{s} \in B^{\mathcal{L}}, \quad B = \{0, 1\}$$

$$s_n(t+1) = F(\mathbf{s}(t)).$$

# Boolean derivatives

$$J_{np} = \frac{\partial s_n(t+1)}{\partial s_p(t)} = |F(\dots, s_p, \dots) - F(\dots, 1 - s_p, \dots)|,$$

$$n, p = 0, \dots, \mathcal{L} - 1,$$

$$\mathbf{v}(t+1) = J(\mathbf{s}(t)) \mathbf{v}(t), \quad \mathbf{v} \in \mathbb{R}^{\mathcal{L}},$$

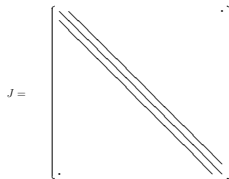
$$|\mathbf{v}(T)| \sim |\mathbf{v}(0)| e^{\lambda T},$$

$$\lambda = \frac{1}{T} \log \left[ \frac{|\mathbf{v}(T)|}{|\mathbf{v}(T-1)|} \frac{|\mathbf{v}(T-1)|}{|\mathbf{v}(T-2)|} \dots \frac{|\mathbf{v}(1)|}{|\mathbf{v}(0)|} \right]$$

$$= \frac{1}{T} \sum_{t=1}^T \log u(t) = \langle \log u(t) \rangle_T,$$

$$u(t) = |\mathbf{v}(t)| / |\mathbf{v}(t-1)|.$$

## Example: elementary cellular automaton rule 150



$$s_i(t+1) = s_{i-1} \oplus s_i \oplus s_{i+1}$$

$$\frac{\partial s_i(t+1)}{\partial s_{i+k}} = s_{i+k} \oplus \overline{s_{i+k}} = 1, \quad k = 0, \pm 1$$

$$\lambda = \log 3$$

# Kinetic theory

## Evolution equation of a LGCA

$$s_k(\mathbf{r} + \mathbf{c}_k, t + 1) = s_k(\mathbf{r}, t) + \delta(s(\mathbf{r}, t)),$$
$$\delta = 0, \pm 1.$$

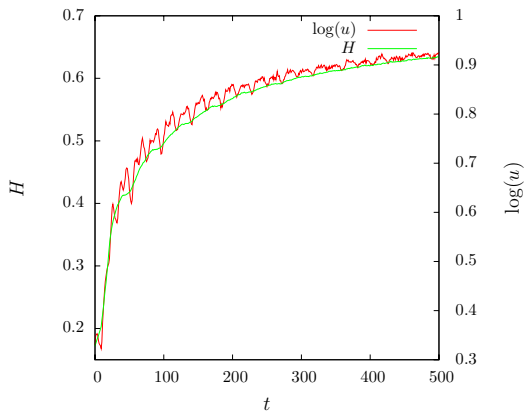
## Boltzmann's transport equation

$$f_k(\mathbf{r}, t) = \langle s_k(\mathbf{r}, t) \rangle,$$
$$f_k(\mathbf{r} + \mathbf{c}_k, t + 1) = f_k(\mathbf{r}, t) + \Delta_k(s(\mathbf{r}, t)).$$

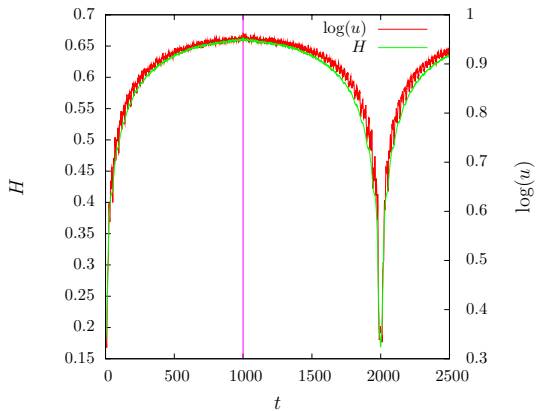
## Boltzmann's H theorem

$$H(t) = - \sum_{\mathbf{r}} \sum_k [f_k \log f_k + (1 - f_k) \log(1 - f_k)],$$
$$\frac{dH}{dt} \geq 0.$$

# Boltzmann's $H$ function and maximum Lyapunov exponent



# Time reversal



# Thermodynamic equilibrium

$$N = \sum_{\mathbf{r}} \sum_k f_k(\mathbf{r}), \quad E = \sum_{\mathbf{r}} \sum_k \epsilon_k f_k(\mathbf{r})$$

$$f_1 = f_2 = f_3 = f_4, \quad f_5 = f_6 = f_7 = f_8,$$

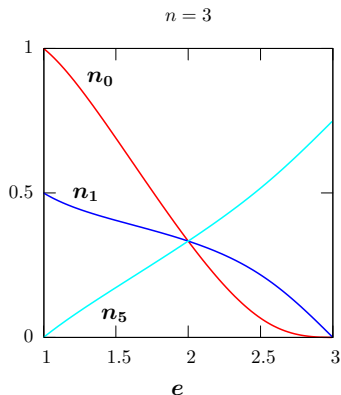
$$N_k = \sum_{\mathbf{r}} f_k(\mathbf{r}),$$

$$n_k = N_k/L, \quad n = N/L, \quad e = E/L$$

$$0 \leq n \leq 9, \quad 0 \leq e \leq 6$$

A. Salcido, R. Rechtman, in P. Cordero, B. Nachtergaele eds., *Nonlinear Phenomena in Fluids, Solids and Other Complex Systems*, Elsevier, 1991.

# Entropy density and equilibrium distribution functions



$$S(E, N, L) = \log \Omega(E, N, L) = Ls$$

$$s(e, n) = - \sum_{k=0}^8 [\hat{n}_k \log \hat{n}_k + (1 - \hat{n}_k) \log(1 - \hat{n}_k)],$$

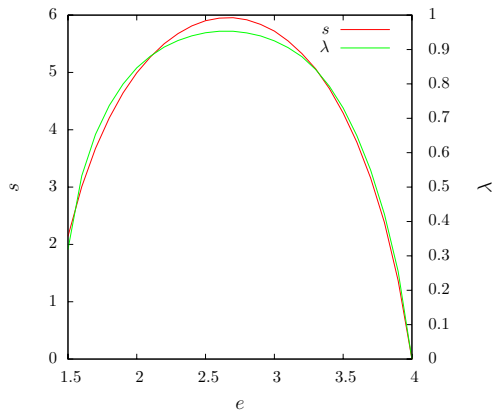
$$\hat{n}_k = [\exp(\beta \epsilon_k - \beta \mu) + 1]^{-1}.$$

$$\beta = \frac{\partial s}{\partial e}, \quad \mu = \frac{\partial s}{\partial n},$$

$$n = \sum_k \hat{n}_k = \hat{n}_0 + 4\hat{n}_1 + \hat{n}_5$$

$$e = \sum_k \epsilon_k \hat{n}_k = 2\hat{n}_1 + 4\hat{n}_5$$

# Equilibrium entropy and maximum Lyapunov exponent



# Conclusions

- LGCA are “**minimal**” models with a hydrodynamic limit. The  $D2Q9$  model includes thermal effects.
- Our version of  $D2Q9$  is a discrete, **deterministic** and **reversible** dynamical system.
- The Jacobian matrix, defined with Boolean derivatives, can be used to measure the **sensitivity** to a (finite) initial difference.
- Confirm the stochastic toy model:

- For the relaxation towards equilibrium,

$$\log u(t) = A + BH(t)$$

- In equilibrium,

$$\lambda = A + Bs$$

- Work is in progress on other systems.