

# Exclusion Processes and Pedestrians

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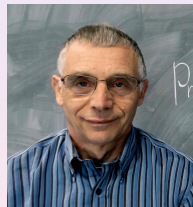
[Cecile.Appert-Rolland@th.u-psud.fr](mailto:Cecile.Appert-Rolland@th.u-psud.fr)

## Out-of-equilibrium

- Road Traffic
- Intracellular traffic
- Pedestrian traffic
- Experiments on pedestrian traffic (PEDIGREE Project)
  - Data analysis
  - Experiment-based models
- Today: Modeling based on exclusion processes



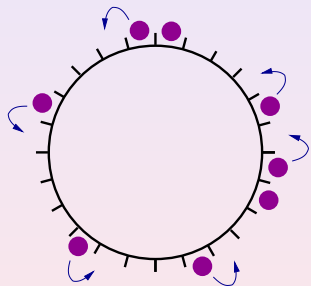
*Work realized in collaboration  
with Julien Cividini and Henk Hilhorst*



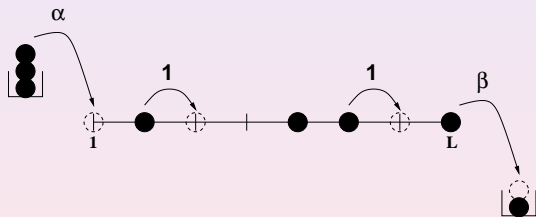
- 1 Introduction: exclusion processes
  - Update schemes
  - The frozen shuffle update
- 2 Crossing of two perpendicular pedestrian flows
  - Diagonal patterns and chevron effect
  - Mean-field approach
  - Wake and effective interaction
- 3 Domain-wall theory for deterministic TASEP with parallel update

# Exclusion Processes

TASEP = Totally Asymmetric Simple Exclusion Process



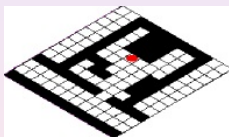
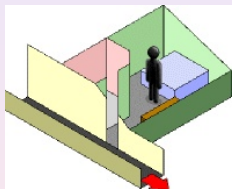
Periodic Boundary Conditions



Open Boundary Conditions

# Cellular automata models for pedestrians

- Floor field model  
[C. Burstedde et al, Physica A **295** (2001) 507-525]
- Ex: PEDGO Software



Cellular automaton = geometry + rules + update

# Updates (discrete time)

- random sequential update

- The unit time step is divided into microsteps  $\delta t = 1/N$
- One particle is randomly chosen and updated at each microtimestep
- ➔ close to a continuous time
- ➔ large fluctuations

- parallel update

- All particles are updated in parallel
- The state of a particle at time  $t + 1$  is determined by the state of the system at time  $t$
- ➔ road traffic ( $\Delta t = 1 = \text{reaction time}$ )
- ➔ conflicts are possible (crossing, lane changing...).

- ordered sequential update (forward, backward...)
  - ➔ depends on the geometry of the system
- random shuffle update [Wölki et al (2006); Smith & Wilson (2007) J. Phys. A]
  - The order of the updates is chosen randomly at each time step
  - Each particle is updated once per time step according to this predefined order.
    - ➔ proposed to model pedestrian traffic
    - ➔ no conflicts, bounded fluctuations
    - ➔ but still, the same pedestrian can be updated twice in a row
- frozen shuffle update

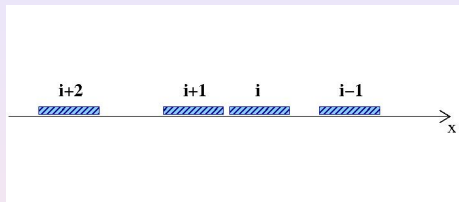


# Frozen shuffle update

- A phase  $\tau_i \in [0, 1[$  is attached to each newly created pedestrian. It remains unchanged during the whole simulation.
- At each time step, pedestrians are updated in the order of increasing phases (time  $t + \tau_i$ ).
  - $\tau_i$  = phase in the walking cycle, slight advance
  - pedestrians have the same speed, no large fluctuations
  - no conflicts
- Deterministic TASEP : all possible moves are accepted
  - ➔ stochasticity comes only from the phases  $\tau_i$ .
  - ➔ analytical predictions
    - TASEP+PBC: [C. A-R, Cividini & Hilhorst, J. Stat. Mech. (2011) P07009]
    - TASEP+OBC: [C. A-R, Cividini & Hilhorst, J. Stat. Mech. (2011) P10013]
- Can also be applied to more realistic models.



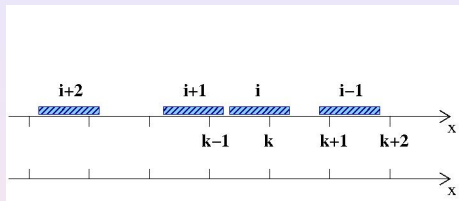
# Mapping a free-flow state to a continuous model



- Continuous model: rods of length one sliding along the  $x$ -axis with velocity 1.

- add an underlying discrete network and take snapshots at integer times  $s$ .
  - Put a particle on site  $k$  if rod  $i$  overlaps position  $x = k$ .
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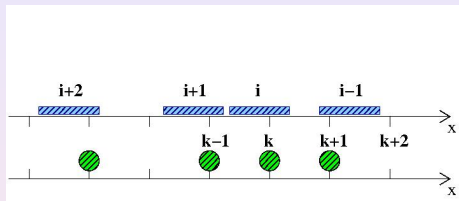
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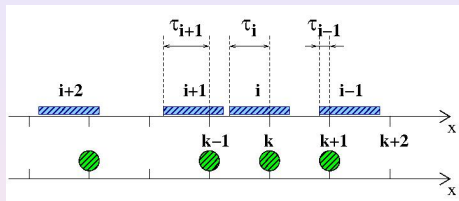
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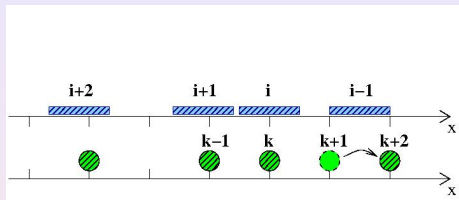
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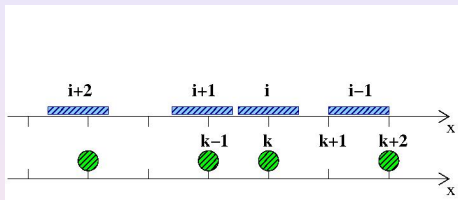
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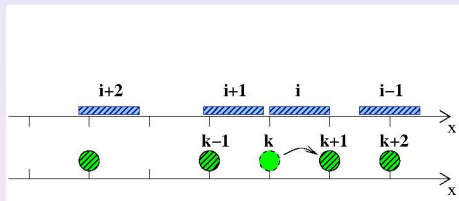
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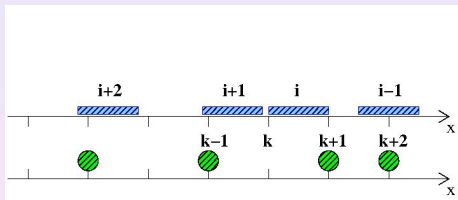


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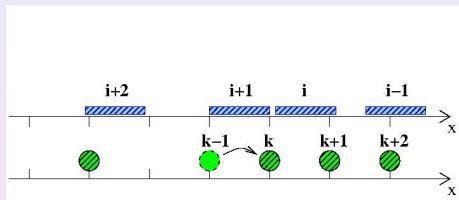
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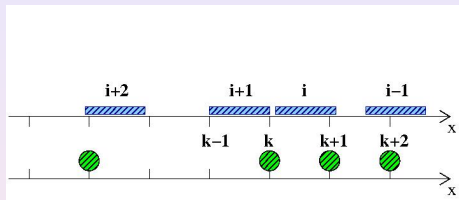
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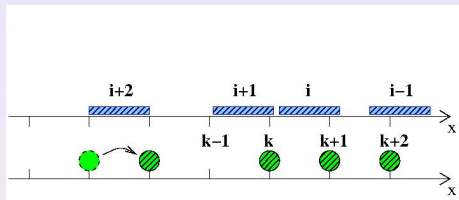
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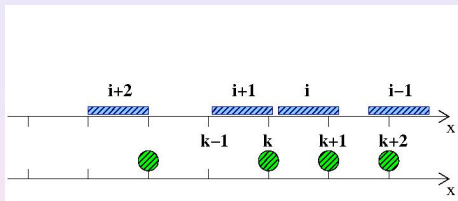
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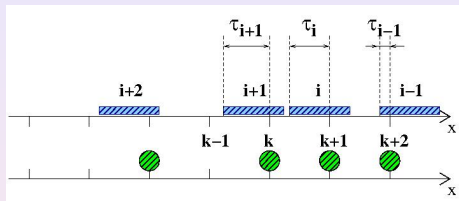
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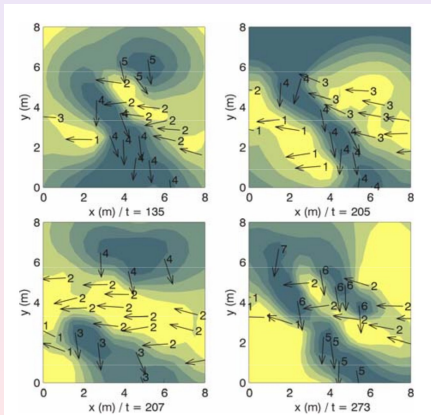
# Intersection of two perpendicular pedestrian flows

Diagonal instability:

- observed in experiments

in

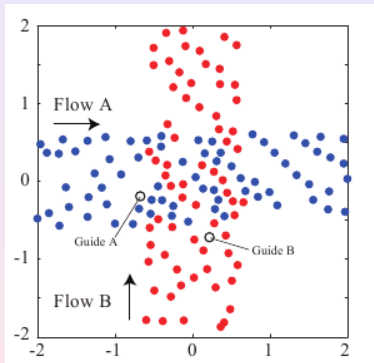
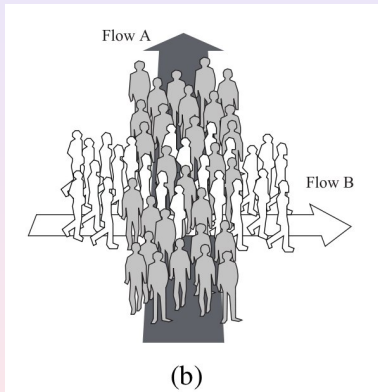
[Hoogendoorn & Daamen,  
TGF'03 (Springer) 2005,  
pp. 121]



# Intersection of two perpendicular pedestrian flows

Diagonal instability:

- observed in simulations



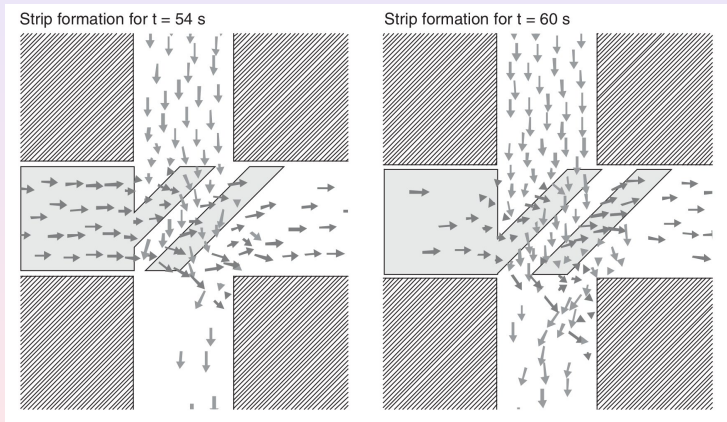
[Yamamoto & Okada, in 2011 IEEE Int. Conf. on Robotics and Automation (ICRA)]



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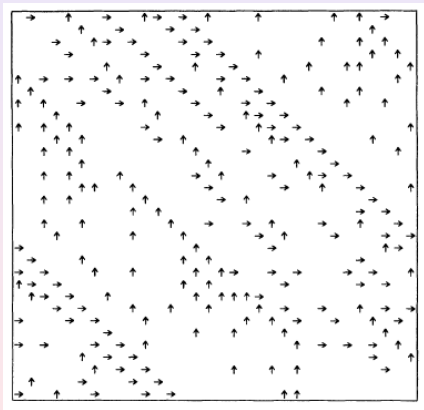


[Hoogendoorn & Bovy, Optim. Control Appl. Meth., 24 (2003) 153]

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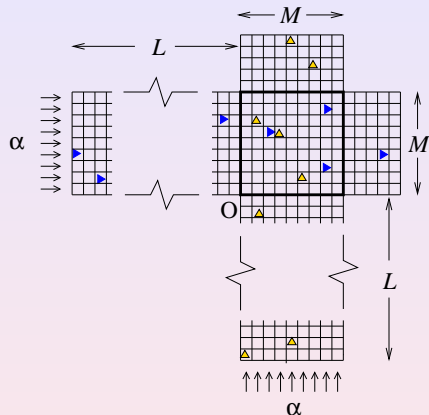


BML Model (city  
traffic)

PBC

[Biham, Middleton & Levine, PRA **46** (1992) R6124]

# Intersection of two corridors



- $\mathcal{E}$  = Eastbound particles
- $\mathcal{N}$  = Northbound particles

$n^{\mathcal{E}}(\mathbf{r}), n^{\mathcal{N}}(\mathbf{r})$  = boolean occupation variables

- As  $\alpha$  increases: jamming transition

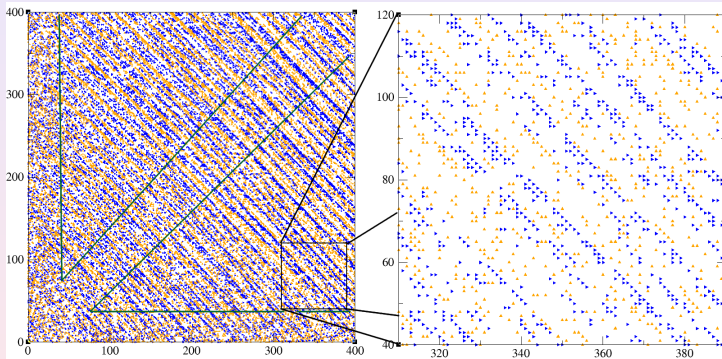
[H. J. Hilhorst, C. A-R, J. Stat. Mech. (2012) P06009]

➡ Here we consider only the free flow phase.

# Observations

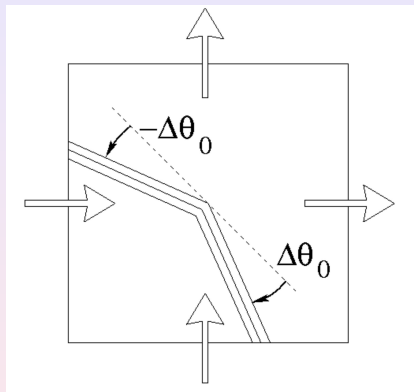
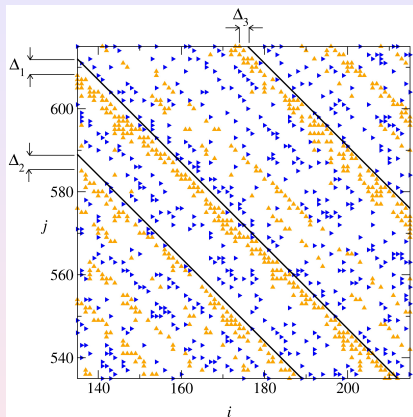
with frozen shuffle update

$M = 400$



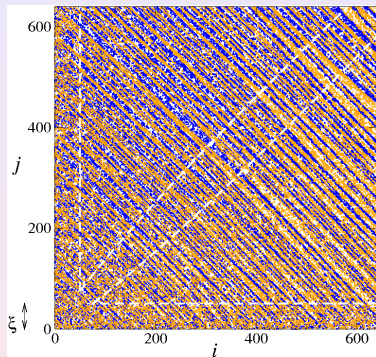
Number of encounters made by a particle:  $g = \rho M$   
= effective coupling constant governing pattern formation

# Observations

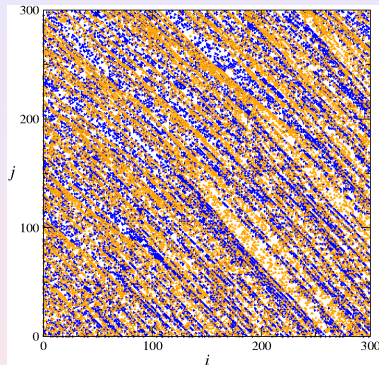


➡ Tilt  $\Delta\theta$

# Observations



- frozen shuffle update
- $M = 640$



- alternating parallel update
- $M = 300$

# Observations

PBC: no tilt

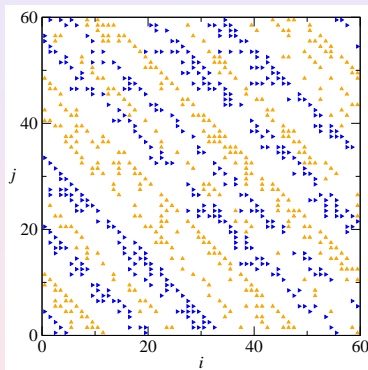
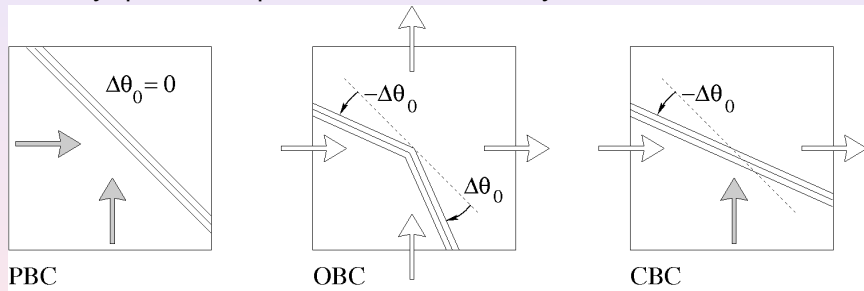


Figure from Chloé Barré

# Observations

Summary: pattern depends on the boundary conditions





# Mean field equations

We postulate some mean-field equations:

$$\begin{aligned}\rho_{t+1}^{\mathcal{E}}(\mathbf{r}) &= [1 - \rho_t^{\mathcal{N}}(\mathbf{r})]\rho_t^{\mathcal{E}}(\mathbf{r} - \mathbf{e}_x) + \rho_t^{\mathcal{N}}(\mathbf{r} + \mathbf{e}_x)\rho_t^{\mathcal{E}}(\mathbf{r}) \\ \rho_{t+1}^{\mathcal{N}}(\mathbf{r}) &= [1 - \rho_t^{\mathcal{E}}(\mathbf{r})]\rho_t^{\mathcal{N}}(\mathbf{r} - \mathbf{e}_y) + \rho_t^{\mathcal{E}}(\mathbf{r} + \mathbf{e}_y)\rho_t^{\mathcal{N}}(\mathbf{r})\end{aligned}$$

- pair correlations  $\langle n^{\mathcal{E}} n^{\mathcal{N}} \rangle$  have been factorized
- interaction terms  $\langle n^{\mathcal{X}} n^{\mathcal{X}} \rangle$  between same-type particles have been neglected (low density)

Simulations: same patterns as for the particle model

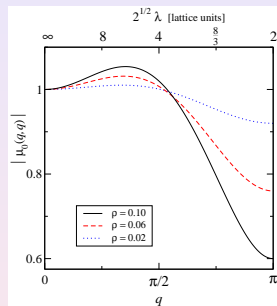
# Mean field equations

## ◆ PBC

- Linear stability analysis

$$\rho_t^{\mathcal{E}, \mathcal{N}}(\mathbf{r}) = \bar{\rho} + \delta \rho_t^{\mathcal{E}, \mathcal{N}}(\mathbf{r})$$

➔ Most unstable mode traveling in the (1, 1) direction with wavelength



$$\begin{aligned}\lambda_{\max} = 2\pi/|\mathbf{q}|_{\max} &= \sqrt{2}\pi / \arccos[(1 - 2\bar{\rho}) / (2 - 2\bar{\rho})] \\ &= 3\sqrt{2}[1 - (\sqrt{3}/\pi)\bar{\rho}] + \mathcal{O}(\bar{\rho}^2),\end{aligned}$$

# Mean field equations

## ◆ OBC

- Linear stability analysis:
  - ➔ In preparation [[Cividini & Hilhorst](#)]
  - ➔ no sign of the chevron effect

Chevron effect = non linear effect

# Chevron effect

- $v^{\mathcal{E}}(\mathbf{r})$  average eastward velocity (in the stationary state)
- $v^{\mathcal{N}}(\mathbf{r})$  average northward velocity

Hypothesis: Moving stripes are mutually impenetrable

➔ Possible only if

$$\tan \theta(\mathbf{r}) = \frac{v^{\mathcal{N}}(\mathbf{r})}{v^{\mathcal{E}}(\mathbf{r})}$$

# Chevron effect

## Particle model:

- *Definition of velocity*

$$v^{\mathcal{E},\mathcal{N}}(\mathbf{r}) = \frac{\mathbf{J}^{\mathcal{E},\mathcal{N}}(\mathbf{r})}{\langle n^{\mathcal{E},\mathcal{N}}(\mathbf{r}) \rangle}$$

where  $\mathbf{J}^{\mathcal{E},\mathcal{N}}(\mathbf{r}) =$  stationary current

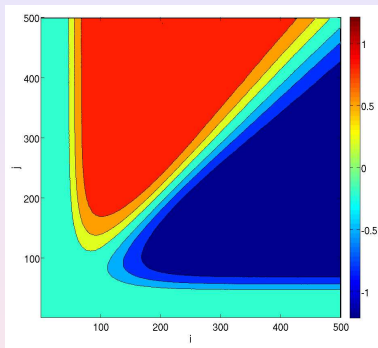
- If  $\mathbf{J}^{\mathcal{E},\mathcal{N}}(\mathbf{r}) = J$ , then

$$\tan \theta(\mathbf{r}) = \frac{\langle n^{\mathcal{E}}(\mathbf{r}) \rangle}{\langle n^{\mathcal{N}}(\mathbf{r}) \rangle}$$

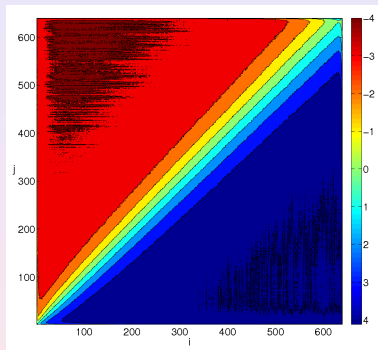
- Setting  $\theta = \frac{\pi}{4} + \Delta\theta$  and expanding yields

$$\Delta\theta(\mathbf{r}) \simeq \frac{\langle n^{\mathcal{E}}(\mathbf{r}) \rangle - \langle n^{\mathcal{N}}(\mathbf{r}) \rangle}{2\langle n^{\mathcal{N}}(\mathbf{r}) \rangle}$$

# Chevron effect



- Mean-field equations

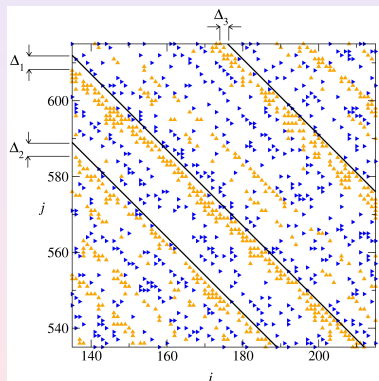


- alternating parallel update

Is the system able to sustain modes with tilted stripes?

# Chevron effect

In the upper triangle:



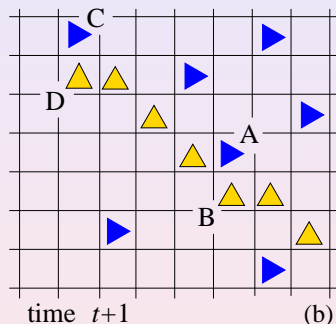
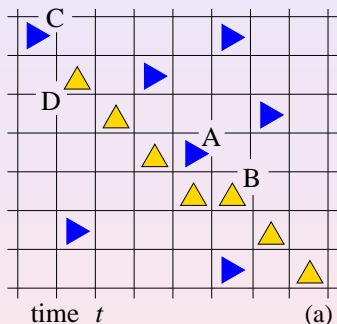
Asymmetry:

- $\mathcal{E}$  stripes (orange) are dense and narrow
- $\mathcal{N}$  stripes (blue) are sparse and wide

# Chevron effect: Identifying a tilted mode

Idealized tilted mode: (alternating parallel update)

Expected near the entrance of  $\mathcal{E}$  particles



figures taken before the hopping of  $\mathcal{E}$  particles

➔ Structure of the stripe is preserved



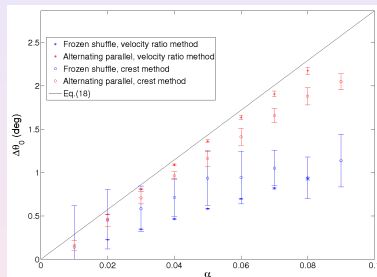
# Chevron effect: Identifying a tilted mode

$$\tan \theta = 1 - \rho_{\text{kink}} = 1 - \rho^{\mathcal{E}} = 1 - \mathcal{J}^{\mathcal{E}}$$

- To lowest order:

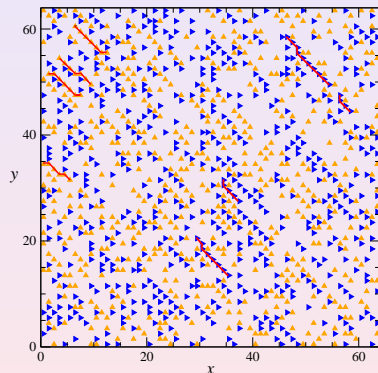
$$\Delta\theta(\mathbf{r}) = \frac{\alpha}{2} \left( \frac{180}{\pi} \right)^{\circ}$$

- Should give an upper bound



# Chevron effect: Identifying a tilted mode

In direct simulations:

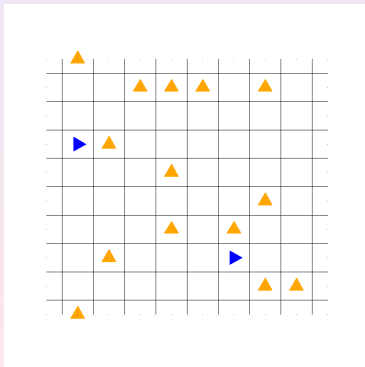


alternating parallel update  
with  $\alpha = 0.15$   
and  $M = 64$

# Effective interactions

From which microscopic mechanism does the (tilted) diagonal pattern emerge?

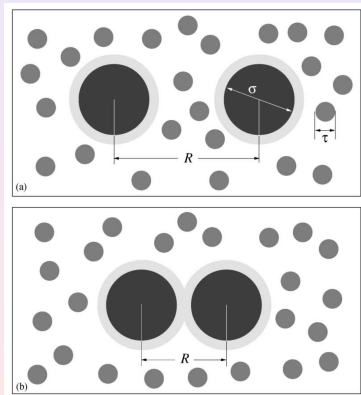
➡ effective interaction between two  $\mathcal{E}$  particles crossing a flow of  $\mathcal{N}$  particles



# Effective interactions

## Environment-mediated interactions:

Most well known : depletion forces



[C. Likos, Physics Reports **348** (2001) 267]

## Environment-mediated interactions: extensively studied

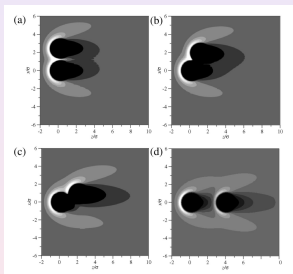
- in equilibrium soft matter

[C. Licos, *Effective interactions in soft condensed matter physics*,  
*Physics Reports* **348** (2001) 267]

# Effective interactions

Environment-mediated interactions: extensively studied

- in equilibrium soft matter
- and more recently in out-of-equilibrium systems



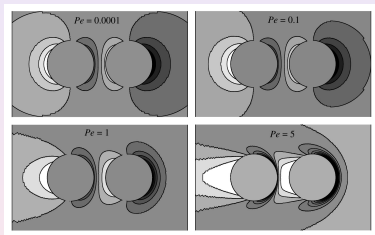
[Dzubiella, Löwen & Likos, Depletion forces in nonequilibrium, PRL **91** (2003) 1]

Approx: perturbation of the density field due to the two large particles  
= superposition of the perturbation due to each large particle separately.

# Effective interactions

Environment-mediated interactions: extensively studied

- in equilibrium soft matter
- and more recently in out-of-equilibrium systems



[Khair & Brady, On the motion of two particles translating with equal velocities through a colloidal dispersion, Proc. R. Soc. A **463** (2007) 223]

No superposition approximation  
but

- interactions in the bath are neglected
- probes are taken aligned

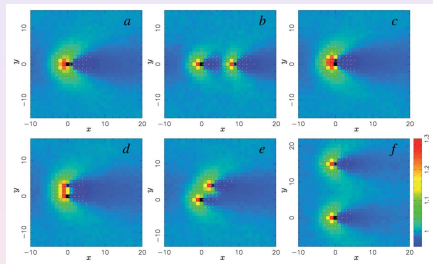
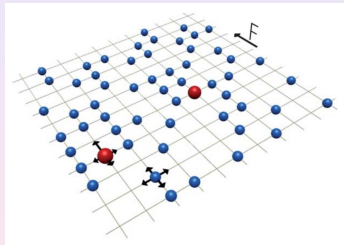
Focus: prediction of effective forces

- Forces are not relevant for pedestrians  
In our case: interaction comes from the dynamical rules



# Effective interactions

- Discrete model

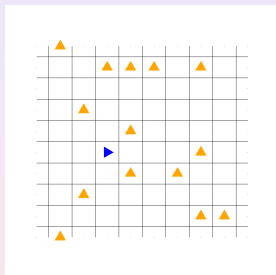


[Mejía-Monasterio & Oshanin, Bias- and bath-mediated pairing of particles driven through a quiescent medium, The royal soc. of chem. **7** (2011) 993]

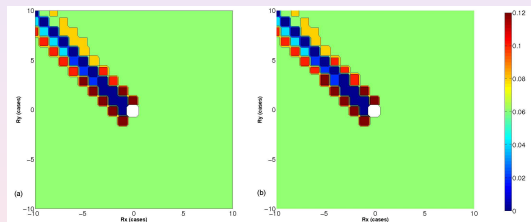
Numerical study  $\Rightarrow$  existence of an attractive interaction between the intruders resulting in a statistical pairing

# Wake of a single $\mathcal{E}$ particle

Ensemble averaged wake



Frozen shuffle update

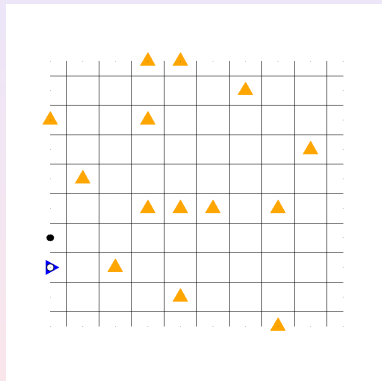


Theory

Simulation

# Wake of a single $\mathcal{E}$ particle

Microscopic structure of the wake:



Central part of the wake : the shadow

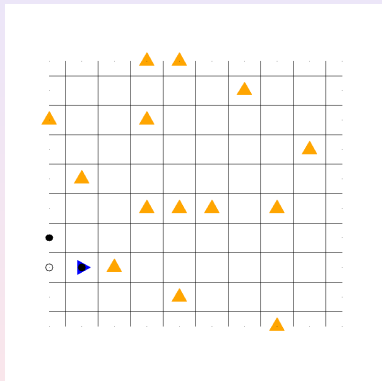
Construction:

- Before move: white dot
- After move: black dot

Again, at low density,  $\tan\theta \simeq 1 - \rho^{\mathcal{N}}$

# Wake of a single $\mathcal{E}$ particle

Microscopic structure of the wake:



Central part of the wake : the shadow

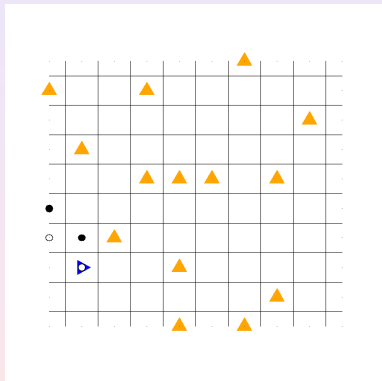
Construction:

- Before move: white dot
- After move: black dot

Again, at low density,  $\tan\theta \simeq 1 - \rho^{\mathcal{N}}$

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# Wake of a single $\mathcal{E}$ particle

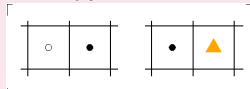
Microscopic structure of the wake:

Central part of the wake : the shadow

Construction:

- Before move: white dot
- After move: black dot

Two types of rows:

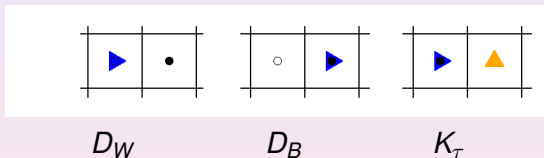


Again, at low density,  $\tan\theta \simeq 1 - \rho^{\mathcal{N}}$

# Two $\mathcal{E}$ particles in a flow of $\mathcal{N}$ particles

Frozen shuffle update:

Let us put a second  $\mathcal{E}$  particle (phase  $\tau_0$ ) in the shadow of the first one (phase 0) :

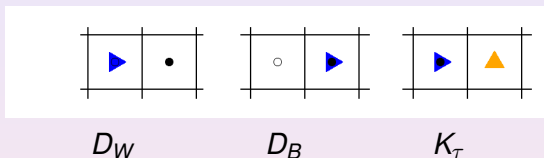


$$P_S(t) = P_W(t) + P_B(t) + \int_0^1 d\tau p_\tau(t)$$

# Two $\mathcal{E}$ particles in a flow of $\mathcal{N}$ particles

Frozen shuffle update:

Let us put a second  $\mathcal{E}$  particle (phase  $\tau_0$ ) in the shadow of the first one (phase 0) :



Low density limit :

$$\begin{cases} P_W^{\text{fs}}(t+1) &= (1 - \rho^{\text{fs}})P_W^{\text{fs}}(t) + \rho^{\text{fs}}(1 - \tau_0)P_B^{\text{fs}}(t) + P_{>\tau_0}^{\text{fs}}(t) \\ P_B^{\text{fs}}(t+1) &= (1 - 2\rho^{\text{fs}} + \rho^{\text{fs}}\tau_0)P_B^{\text{fs}}(t) + P_{<\tau_0}^{\text{fs}}(t) \\ P_{<\tau_0}^{\text{fs}}(t+1) &= \rho^{\text{fs}}\tau_0 P_W^{\text{fs}}(t) + \rho\tau_0 P_B^{\text{fs}}(t) \\ P_{>\tau_0}^{\text{fs}}(t+1) &= \rho^{\text{fs}}(1 - \tau_0)P_W^{\text{fs}}(t) \end{cases}$$

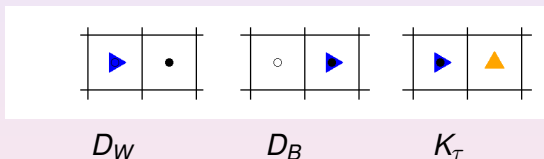
where  $P_{<\tau_0}^{\text{fs}}(t) \equiv \int_{\tau=0}^{\tau_0} p_\tau(t) d\tau$  and  $P_{>\tau_0}^{\text{fs}}(t) \equiv \int_{\tau=\tau_0}^1 p_\tau(t) d\tau$ .



# Two $\mathcal{E}$ particles in a flow of $\mathcal{N}$ particles

Frozen shuffle update:

Let us put a second  $\mathcal{E}$  particle (phase  $\tau_0$ ) in the shadow of the first one (phase 0) :



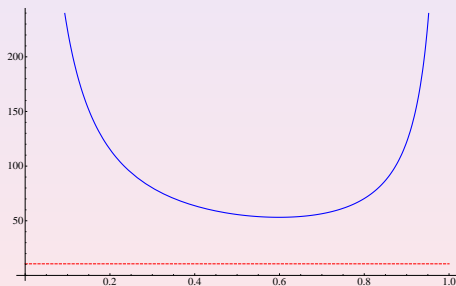
$$P_S^{\text{fs}}(t+1) = P_S^{\text{fs}}(t) - \rho^{\text{fs}}(1 - \tau_0)P_B^{\text{fs}}(t)$$

# Two $\mathcal{E}$ particles in a flow of $\mathcal{N}$ particles

Diagonalization of the transfer matrix

→ time evolution of  $P_S^{\text{fs}}(t)$  (linear combination of exponentials).

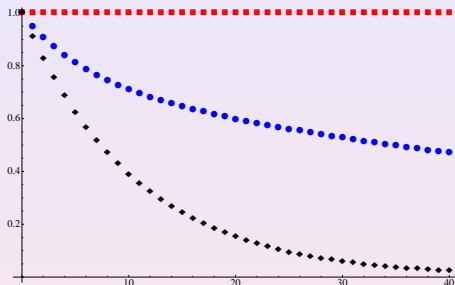
*Longest characteristic decay time as a function of  $\tau_0$*



*red: uncorrelated case*

# Two $\mathcal{E}$ particles in a flow of $\mathcal{N}$ particles

Probability that the second particle stays in the shadow during the first  $t$  timesteps



- blue: frozen shuffle update
- red: alternating parallel update
- black: uncorrelated

Alternating parallel update: overlap of shadows  $\Leftrightarrow$  several particles can be localized in the shadow of the first particle.

# Conclusion

- Exemple of an effective interaction that can be solved analytically.
- Calculations can be extended to other update schemes, provided the free flow phase is deterministically shifted forward with velocity 1 at each time step.
- The angle of the wake = angle of the long-lived global mode identified before.
- In the full problem, angle may be different and depend on the update, though the order of magnitude should be the same.

[J. Cividini and C. A-R, *Wake-mediated interaction between driven particles crossing a perpendicular flow*, J. Stat. Mech. (2013) P07015]

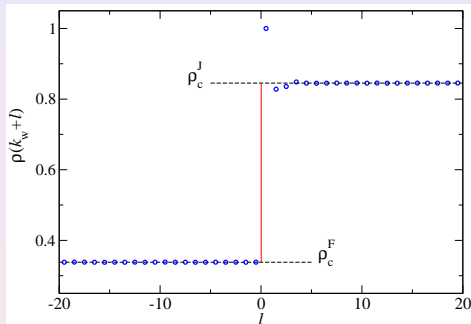
[Kolomeisky, Schütz, Kolomeisky and Straley, J. Phys. A: Math. Gen. **31** (1998) 6911]

- phenomenological picture
- can be more easily extended to non stationary states, variants of the ASEP
- physical understanding

# Domain wall picture

For deterministic updates for which free flow has velocity 1:

- microscopic definition of the wall
- wall position = position of the leftmost particle that has ever been blocked



$$\alpha = \beta = 0.4$$

$$L = 3000$$

frozen shuffle update

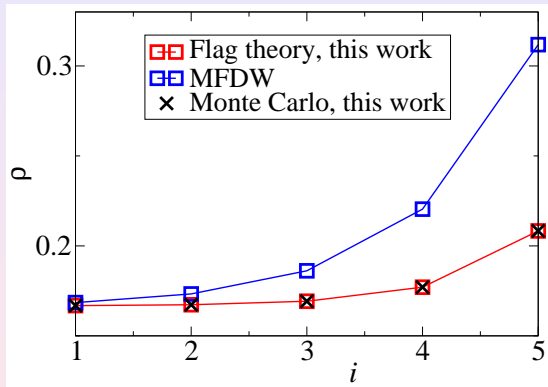
# Domain wall picture

Usual domain wall theory not appropriate for this case.

⇒ Extension of the domain wall theory to the case of deterministic parallel update

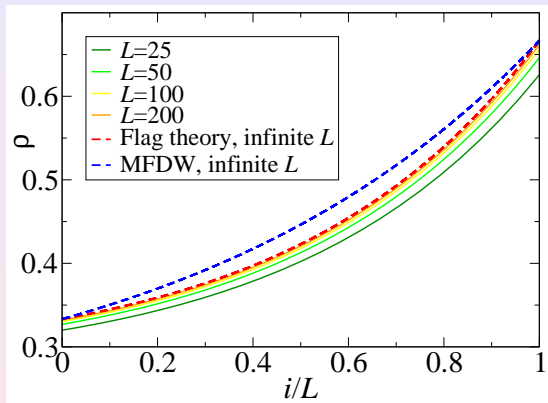
- There are correlations between successive steps of the domain wall
- ➔ Memory effect
- ➔ Two coupled master equations (+BC)
- Continuum limit of this exactly soluble model
- ➔ Fokker-Planck equation for the position of the wall
  - ⇒ with a diffusion constant different from the one that would be obtained by the usual DW theory
  - ⇒ in agreement with Monte-Carlo simulations.

# Domain wall picture





# Domain wall picture



# Conclusion

For more details:

[http://www.th.u-psud.fr/page\\_perso/Appert/](http://www.th.u-psud.fr/page_perso/Appert/)

Thank-you

## PEDIGREE Project 2009-2011 (LPT, IMT, CRCA, Bunraku)

- Experiments on pedestrian traffic
  - Ring  
[Moussaid et al, PLoS Computational Biology **8** (2012) 1002442]
  - 1D circle  
[Jelic et al, PRE **85** (2012) 036111]
- Models
  - Continuous model for bidirectional crowd motion  
[C. A.-R., P. Degond, S. Motsch, NHM **6** (2011) 351]
  - Following model [S. Lemerrier et al, Eurographics (2012)]



## Road Traffic

- Kinetic model for a bi-directional road [C. A.-R., H. Hilhorst, G. Schehr, J. Stat. Mech. (2010)]
- Response of a multi-lane highway to a local perturbation [C. A.-R., J. Du Boisberranger, Transp. Res. C (2013)]

## Intracellular Traffic

- **Intracellular transport** (collab. Sarrebrücken, L. Santen, M. Ebbinghaus, I. Weber, S. Klein)