

# Pedestrian Flow Models with Slowdown Interactions

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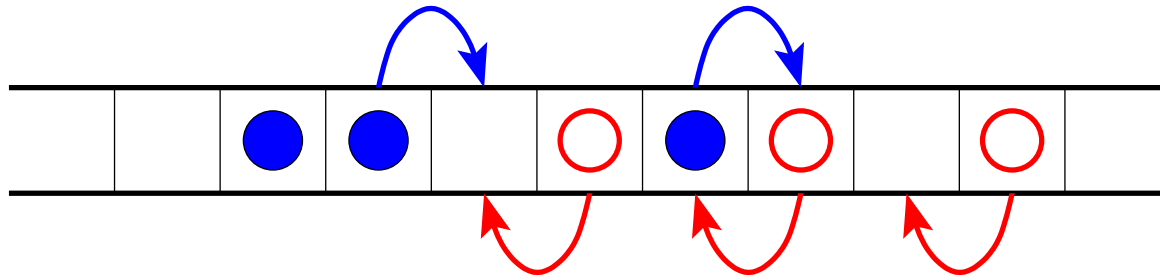
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# Outline

**Goal:** Systematic Derivation of PDE Models for Pedestrian Traffic Flow

- Microscopic Rules for the Interaction of Pedestrians Moving in Opposite Directions
- Microscopic Cellular Automata Model for Pedestrian Flow
- Derivation of the Coarse-Grained PDE
- Derivation of Nonlinear Diffusion
- Numerical Examples – Quantitative Agreement with Stochastic Simulations in Weaker Slowdown Regime

# Stochastic Lattice Model of Pedestrian Traffic



**Approach:** One-dimensional  $\{0, 1\}$  Lattice Configuration for

$$\sigma_k^{\pm}(t) = \begin{cases} 1, & \text{pedestrian moving to the right (left)} \\ 0, & \text{empty cell} \end{cases}$$

**If no pedestrians are moving in the opposite direction:** Equivalent to car traffic models

- Two pedestrians moving in the same direction cannot occupy the same cell
- Pedestrians moving into two opposite directions can occupy the same cell

To construct the microscopic cellular automata model, we consider explicit rules for the slowdown interaction.

We prescribe **transition probabilities** for four different pedestrian configurations in the cells neighboring to the right-moving pedestrian with  $\sigma_k^+ = 1$  (assuming that  $\sigma_{k+1}^+ = 0$ ):

$$\begin{cases} c_0 \Delta t, & \text{if } \sigma_k^- = \sigma_{k+1}^- = 0 \text{ (no left-moving pedestrians in cells } k \text{ or } k+1) \\ c_1 \Delta t, & \text{if } \sigma_k^- = 1, \sigma_{k+1}^- = 0 \text{ (a left-moving pedestrian is in cell } k) \\ c_2 \Delta t, & \text{if } \sigma_k^- = 0, \sigma_{k+1}^- = 1 \text{ (a left-moving pedestrian is in cell } k+1) \\ c_3 \Delta t, & \text{if } \sigma_k^- = \sigma_{k+1}^- = 1 \text{ (left-moving pedestrians in cells } k \text{ and } k+1) \end{cases}$$

From the common sense considerations, the velocities should obey the following relationship:  $c_3 < c_2 \gtrsim c_1 < c_0$

Transition probabilities for the left-moving pedestrian  $\sigma_k^- = 1$  can be obtain in a similar manner.

**Probability** of a right-moving pedestrian to move from cell  $k$  to cell  $k+1$  within  $\Delta t$  is

$$P_{k \rightarrow k+1}^+ = \Delta t \left[ c_0 \sigma_k^+ (1 - \sigma_{k+1}^+) (1 - \sigma_k^-) (1 - \sigma_{k+1}^-) \right. \\ \left. + c_1 \sigma_k^+ (1 - \sigma_{k+1}^+) \sigma_k^- (1 - \sigma_{k+1}^-) \right. \\ \left. + c_2 \sigma_k^+ (1 - \sigma_{k+1}^+) (1 - \sigma_k^-) \sigma_{k+1}^- + c_3 \sigma_k^+ (1 - \sigma_{k+1}^+) \sigma_k^- \sigma_{k+1}^- \right]$$

**Probability** of a left-moving pedestrian to move from cell  $k$  to cell  $k-1$  within  $\Delta t$  is

$$P_{k \rightarrow k-1}^- = \Delta t \left[ c_0 \sigma_k^- (1 - \sigma_{k-1}^-) (1 - \sigma_{k-1}^+) (1 - \sigma_k^+) \right. \\ \left. + c_1 \sigma_k^- (1 - \sigma_{k-1}^-) (1 - \sigma_{k-1}^+) \sigma_k^+ \right. \\ \left. + c_2 \sigma_k^- (1 - \sigma_{k-1}^-) \sigma_{k-1}^+ (1 - \sigma_k^+) + c_3 \sigma_k^- (1 - \sigma_{k-1}^-) \sigma_{k-1}^+ \sigma_k^+ \right]$$

Goal: Predict the Density of the Pedestrian Traffic,  $\mathbb{E}\sigma_k^+(t)$  and  $\mathbb{E}\sigma_k^-(t)$

$$\begin{aligned} \frac{d\mathbb{E}\sigma_k^+}{dt} = & \mathbb{E} \left[ c_0 \sigma_{k-1}^+ (1 - \sigma_k^+) (1 - \sigma_{k-1}^-) (1 - \sigma_k^-) - c_0 \sigma_k^+ (1 - \sigma_{k+1}^+) (1 - \sigma_k^-) (1 - \sigma_{k+1}^-) \right. \\ & + c_1 \sigma_{k-1}^+ (1 - \sigma_k^+) \sigma_{k-1}^- (1 - \sigma_k^-) - c_1 \sigma_k^+ (1 - \sigma_{k+1}^+) \sigma_k^- (1 - \sigma_{k+1}^-) \\ & + c_2 \sigma_{k-1}^+ (1 - \sigma_k^+) (1 - \sigma_{k-1}^-) \sigma_k^- - c_2 \sigma_k^+ (1 - \sigma_{k+1}^+) (1 - \sigma_k^-) \sigma_{k+1}^- \\ & \left. + c_3 \sigma_{k-1}^+ (1 - \sigma_k^+) \sigma_{k-1}^- \sigma_k^- - c_3 \sigma_k^+ (1 - \sigma_{k+1}^+) \sigma_k^- \sigma_{k+1}^- \right] \end{aligned}$$

$$\begin{aligned} \frac{d\mathbb{E}\sigma_k^-}{dt} = & \mathbb{E} \left[ c_0 \sigma_{k+1}^- (1 - \sigma_k^-) (1 - \sigma_k^+) (1 - \sigma_{k+1}^+) - c_0 \sigma_k^- (1 - \sigma_{k-1}^-) (1 - \sigma_{k-1}^+) (1 - \sigma_k^+) \right. \\ & + c_1 \sigma_{k+1}^- (1 - \sigma_k^-) (1 - \sigma_k^+) \sigma_{k+1}^+ - c_1 \sigma_k^- (1 - \sigma_{k-1}^-) (1 - \sigma_{k-1}^+) \sigma_k^+ \\ & + c_2 \sigma_{k+1}^- (1 - \sigma_k^-) \sigma_k^+ (1 - \sigma_{k+1}^+) - c_2 \sigma_k^- (1 - \sigma_{k-1}^-) \sigma_{k-1}^+ (1 - \sigma_k^+) \\ & \left. + c_3 \sigma_{k+1}^- (1 - \sigma_k^-) \sigma_k^+ \sigma_{k+1}^+ - c_3 \sigma_k^- (1 - \sigma_{k-1}^-) \sigma_{k-1}^+ \sigma_k^+ \right] \end{aligned}$$

These equations are exact, but not closed!

# Mesoscopic Model

- Notations:  $\rho_k^\pm(t) := \mathbb{E}\sigma_k^\pm$

- Assumptions:  $\mathbb{E}[\sigma_{k-1}^+ \sigma_k^+ \sigma_{k-1}^- \sigma_k^-] \approx \mathbb{E}[\sigma_{k-1}^+] \mathbb{E}[\sigma_k^+] \mathbb{E}[\sigma_{k-1}^-] \mathbb{E}[\sigma_k^-]$

$$\begin{aligned} \frac{d\rho_k^+}{dt} = & c_0 \rho_{k-1}^+ (1 - \rho_k^+) (1 - \rho_{k-1}^-) (1 - \rho_k^-) - c_0 \rho_k^+ (1 - \rho_{k+1}^+) (1 - \rho_k^-) (1 - \rho_{k+1}^-) \\ & + c_1 \rho_{k-1}^+ (1 - \rho_k^+) \rho_{k-1}^- (1 - \rho_k^-) - c_1 \rho_k^+ (1 - \rho_{k+1}^+) \rho_k^- (1 - \rho_{k+1}^-) \\ & + c_2 \rho_{k-1}^+ (1 - \rho_k^+) (1 - \rho_{k-1}^-) \rho_k^- - c_2 \rho_k^+ (1 - \rho_{k+1}^+) (1 - \rho_k^-) \rho_{k+1}^- \\ & + c_3 \rho_{k-1}^+ (1 - \rho_k^+) \rho_{k-1}^- \rho_k^- - c_3 \rho_k^+ (1 - \rho_{k+1}^+) \rho_k^- \rho_{k+1}^- \end{aligned}$$

$$\begin{aligned} \frac{d\rho_k^-}{dt} = & c_0 \rho_{k+1}^- (1 - \rho_k^-) (1 - \rho_k^+) (1 - \rho_{k+1}^+) - c_0 \rho_k^- (1 - \rho_{k-1}^-) (1 - \rho_{k-1}^+) (1 - \rho_k^+) \\ & + c_1 \rho_{k+1}^- (1 - \rho_k^-) (1 - \rho_k^+) \rho_{k+1}^+ - c_1 \rho_k^- (1 - \rho_{k-1}^-) (1 - \rho_{k-1}^+) \rho_k^+ \\ & + c_2 \rho_{k+1}^- (1 - \rho_k^-) \rho_k^+ (1 - \rho_{k+1}^+) - c_2 \rho_k^- (1 - \rho_{k-1}^-) \rho_{k-1}^+ (1 - \rho_k^+) \\ & + c_3 \rho_{k+1}^- (1 - \rho_k^-) \rho_k^+ \rho_{k+1}^+ - c_3 \rho_k^- (1 - \rho_{k-1}^-) \rho_{k-1}^+ \rho_k^+ \end{aligned}$$

# Macroscopic PDE Model

- $k \in \mathcal{L}$ : cells with some fixed length  $h > 0$  in the lattice  $\mathcal{L}$
- $\Omega = [0, L]$  corresponds to  $\mathcal{L}$  (the number of cells  $N$  depends on  $h$ )
- $t \rightarrow ht$  and  $N \rightarrow \infty$

We rewrite the mesoscopic system in the following flux form:

$$\frac{d\rho_k^+}{dt} = -\frac{F_{k,k+1}^+ - F_{k-1,k}^+}{h}, \quad \frac{d\rho_k^-}{dt} = \frac{F_{k,k+1}^- - F_{k-1,k}^-}{h}$$

where

$$F_{k,k+1}^+ = \rho_k^+ (1 - \rho_{k+1}^+) \left[ (1 - \rho_{k+1}^-) (c_0(1 - \rho_k^-) + c_1\rho_k^-) + \rho_{k+1}^- (c_2(1 - \rho_k^-) + c_3\rho_k^-) \right]$$

$$F_{k,k+1}^- = \rho_{k+1}^- (1 - \rho_k^-) \left[ (1 - \rho_k^+) (c_0(1 - \rho_{k+1}^+) + c_1\rho_{k+1}^+) + \rho_k^+ (c_2(1 - \rho_{k+1}^+) + c_3\rho_{k+1}^+) \right]$$



$$\frac{d\rho_k^+}{dt} = -\frac{F_{k,k+1}^+ - F_{k-1,k}^+}{h}, \quad \frac{d\rho_k^-}{dt} = \frac{F_{k,k+1}^- - F_{k-1,k}^-}{h}$$

- Multiply these equations by  $\varphi_k := \varphi(kh)$ , where  $\varphi$  is a  $C_0^1$  test function
- Use the summation by parts over  $\Omega$ :

$$\sum_k \varphi_k \frac{d\rho_k^\pm}{dt} = \pm \sum_k F_{k,k+1}^\pm \frac{\varphi_{k+1} - \varphi_k}{h}$$

- Multiply by  $h$  and expand  $\varphi_{k+1}$  into a Taylor series about  $kh$ :

$$\sum_k \varphi_k \frac{d\rho_k^\pm}{dt} h = \pm \sum_k F_{k,k+1}^\pm [\varphi_k' + \mathcal{O}(h)] h$$

Define pedestrian densities on  $\Omega$  as follows:

- Define the function  $\rho^\pm(x, t)$  as a continuous piecewise linear interpolation (in the spatial variable) of  $\rho_k^\pm(t)$
- Take the limit as  $h \rightarrow 0+$

Due to the boundedness of both  $\rho^\pm$  and  $\frac{d\rho_k^\pm}{dt}$  we obtain a **weak formulation of the coarse-grained model**:

$$\int_{\Omega} \varphi(x) \frac{\partial}{\partial t} \rho^\pm(x, t) dx = \pm \int_{\Omega} F^\pm(\rho^+, \rho^-) \varphi'(x) dx$$

where  $F^\pm(\rho^+, \rho^-)$  are defined as the corresponding limits of  $F_{k,k+1}^\pm$ :

$$F^+(\rho^+, \rho^-) = f(\rho^+)g(\rho^-), \quad F^-(\rho^+, \rho^-) = f(\rho^-)g(\rho^+)$$

where

$$f(u) = u(1 - u), \quad g(u) = (c_3 - c_2 - c_1 + c_0)u^2 + (c_2 + c_1 - 2c_0)u + c_0$$

$$\int_{\Omega} \varphi(x) \frac{\partial}{\partial t} \rho^+(x, t) dx = \int_{\Omega} f(\rho^+) g(\rho^-) \varphi'(x) dx$$

$$\int_{\Omega} \varphi(x) \frac{\partial}{\partial t} \rho^-(x, t) dx = - \int_{\Omega} f(\rho^-) g(\rho^+) \varphi'(x) dx$$

Since  $\varphi$  is arbitrary, we have

$$\begin{cases} \rho_t^+ + [f(\rho^+)g(\rho^-)]_x = 0 \\ \rho_t^- - [f(\rho^-)g(\rho^+)]_x = 0 \end{cases}$$

where

$$f(u) = u(1 - u), \quad g(u) = (c_3 - c_2 - c_1 + c_0)u^2 + (c_2 + c_1 - 2c_0)u + c_0$$

Note that the velocities  $c_1$  and  $c_2$  enter only as a sum, and, therefore, it is not necessary to specify them separately

## Properties of the PDE Model

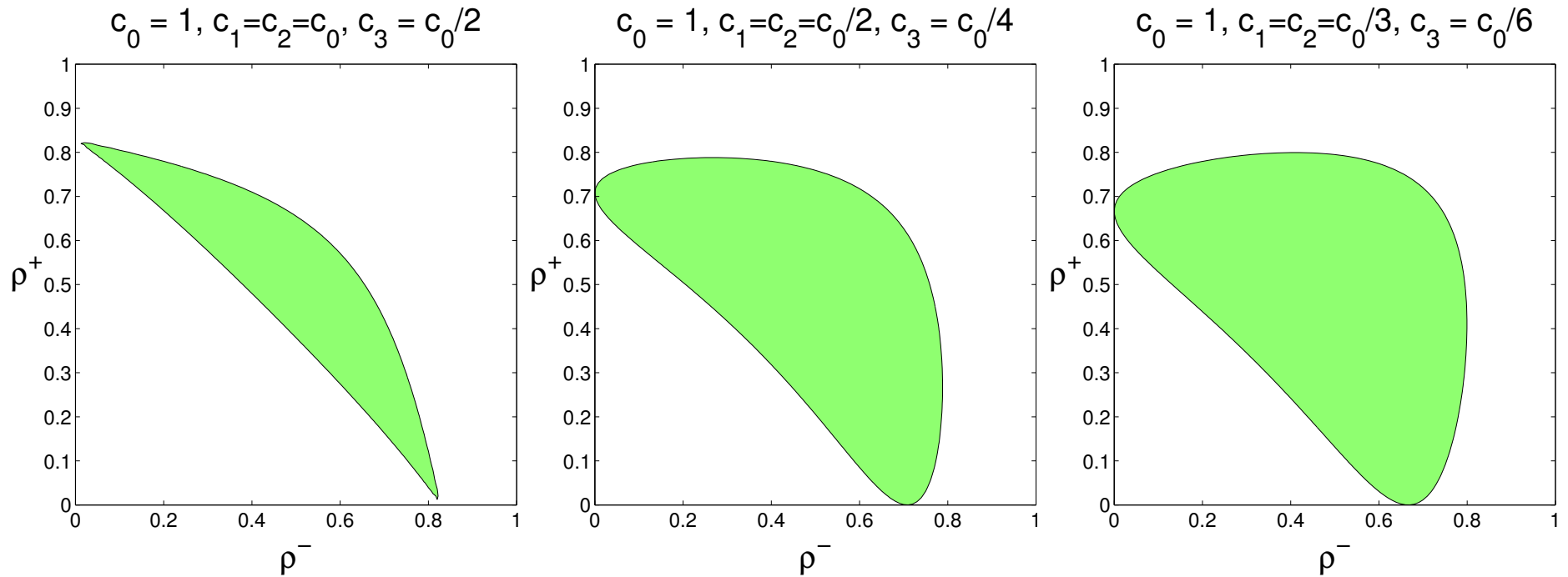
The PDE system is only conditionally hyperbolic:

$$J(f, g) := \begin{pmatrix} f'(\rho^+)g(\rho^-) & f(\rho^+)g'(\rho^-) \\ -f(\rho^-)g'(\rho^+) & -f'(\rho^-)g(\rho^+) \end{pmatrix}$$

has real eigenvalues only if

$$\left[ f'(\rho^-)g(\rho^+) + f'(\rho^+)g(\rho^-) \right]^2 - 4f(\rho^-)f(\rho^+)g'(\rho^-)g'(\rho^+) > 0$$

- For any particular choice of the velocities  $c_0, c_1, c_2$  and  $c_3$  there is a region on non-hyperbolicity in the  $(\rho^-, \rho^+)$  plane
- The non-hyperbolicity can only manifest itself when pedestrians moving in the opposite directions are both present in a particular location



- The **non-hyperbolic region** depends only on the ratio of velocities  $c_1/c_0, c_2/c_0$  and  $c_3/c_0$ , but not on the particular value of  $c_0$
- The non-hyperbolic region becomes larger as the slowdown effect becomes more pronounced (i.e., as the ratios  $c_1/c_0, c_2/c_0$  and  $c_3/c_0$  become smaller)
- The loss of hyperbolicity may induce instabilities, which are nonphysical and can be removed by introducing a **nonlinear diffusive correction** to the system

# Nonlinear Diffusive Correction

$$\frac{d\rho_k^+}{dt} = -\frac{F_{k,k+1}^+ - F_{k-1,k}^+}{h}, \quad \frac{d\rho_k^-}{dt} = \frac{F_{k,k+1}^- - F_{k-1,k}^-}{h}$$

where

$$F_{k,k+1}^+ = \rho_k^+ (1 - \rho_{k+1}^+) \left[ (1 - \rho_{k+1}^-) (c_0(1 - \rho_k^-) + c_1\rho_k^-) + \rho_{k+1}^- (c_2(1 - \rho_k^-) + c_3\rho_k^-) \right]$$

$$F_{k,k+1}^- = \rho_{k+1}^- (1 - \rho_k^-) \left[ (1 - \rho_k^+) (c_0(1 - \rho_{k+1}^+) + c_1\rho_{k+1}^+) + \rho_k^+ (c_2(1 - \rho_{k+1}^+) + c_3\rho_{k+1}^+) \right]$$

The derivation of the coarse-grained PDE system can also be obtained by formally using the Taylor expansions

$$\rho_{k\pm 1}^\pm = \rho_k^\pm \pm h(\rho_k^\pm)' + \frac{h^2}{2}(\rho_k^\pm)'' + \mathcal{O}(h^3)$$

followed by passing to the limit as  $h \rightarrow 0+$

- Keep  $h$  fixed
- Neglect the  $\mathcal{O}(h^3)$  terms

$$\begin{aligned} \rho_t^+ + [f(\rho^+)g(\rho^-)]_x &= h \left[ \frac{c_0}{2} \rho_{xx}^+ + (c_1 - c_0 + (c_3 - c_2 - c_1 + c_0)\rho^- \right. \\ &\quad \left. + (c_2 - c_1)\rho^+) \rho_x^- \rho_x^+ + \frac{1}{2}(c_1 - c_2)\rho^+(1 - \rho^+)\rho_{xx}^- \right. \\ &\quad \left. + \frac{1}{2} \left( (c_1 + c_2 - 2c_0)\rho^- + (c_3 - c_2 - c_1 + c_0)(\rho^-)^2 \right) \rho_{xx}^+ \right] \end{aligned}$$

$$\begin{aligned} \rho_t^- - [f(\rho^-)g(\rho^+)]_x &= h \left[ \frac{c_0}{2} \rho_{xx}^- + (c_1 - c_0 + (c_3 - c_2 - c_1 + c_0)\rho^+ \right. \\ &\quad \left. + (c_2 - c_1)\rho^-) \rho_x^+ \rho_x^- + \frac{1}{2}(c_1 - c_2)\rho^-(1 - \rho^-)\rho_{xx}^+ \right. \\ &\quad \left. + \frac{1}{2} \left( (c_2 + c_1 - 2c_0)\rho^+ + (c_3 - c_2 - c_1 + c_0)(\rho^+)^2 \right) \rho_{xx}^- \right] \end{aligned}$$

- Replace  $h$  with a small parameter  $\varepsilon$  and use the formulae for the fluxes

$$f(u) = u(1 - u), \quad g(u) = (c_3 - c_2 - c_1 + c_0)u^2 + (c_2 + c_1 - 2c_0)u + c_0$$

to obtain

$$\begin{cases} \rho_t^+ + [f(\rho^+)g(\rho^-)]_x = \frac{\varepsilon}{2} [g(\rho^-)\rho_x^+ + (c_1 - c_2)f(\rho^+)\rho_x^-]_x \\ \rho_t^- - [f(\rho^-)g(\rho^+)]_x = \frac{\varepsilon}{2} [g(\rho^+)\rho_x^- + (c_1 - c_2)f(\rho^-)\rho_x^+]_x \end{cases}$$

- The coefficients of the nonlinear diffusion are positive provided  $c_1 \geq c_2$  and both  $\rho^+$  and  $\rho^-$  are between 0 and 1
- Further simplifying assumption  $c_1 = c_2$  leads to

$$\begin{cases} \rho_t^+ + [f(\rho^+)g(\rho^-)]_x = \frac{\varepsilon}{2} [g(\rho^-)\rho_x^+]_x \\ \rho_t^- - [f(\rho^-)g(\rho^+)]_x = \frac{\varepsilon}{2} [g(\rho^+)\rho_x^-]_x \end{cases}$$

- The assumption  $c_1 = c_2$  is rather mild since the velocities  $c_1$  and  $c_2$  only enter as a sum into the fluxes



$$\begin{cases} \rho_t^+ + [f(\rho^+)g(\rho^-)]_x = \frac{\varepsilon}{2} [g(\rho^-)\rho_x^+]_x \\ \rho_t^- - [f(\rho^-)g(\rho^+)]_x = \frac{\varepsilon}{2} [g(\rho^+)\rho_x^-]_x \end{cases}$$

- The nonlinear diffusion reflects the presence of pedestrians moving in the opposite direction
  - If  $\rho^- = 0$  (i.e., no pedestrians moving to the left are present), the diffusion reduces to the usual linear diffusion  $0.5\varepsilon c_0 \rho_{xx}^+$
  - If  $\rho^- = 1$ , then the diffusion also becomes linear  $0.5\varepsilon c_3 \rho_{xx}^+$ , but with a smaller coefficient (since  $c_3 < c_0$ ) reflecting a high density presence of the pedestrians moving in the opposite direction
- $\varepsilon$  needs to be established experimentally

# Semi-Discrete Central-Upwind Scheme

$$\rho_t + \mathbf{F}(\rho)_x = (Q(\rho)\rho_x)_x$$

- $\rho := (\rho^+, \rho^-)^T$
- $\mathbf{F}(\rho) := (f(\rho^+)g(\rho^-), f(\rho^-)g(\rho^+))^T$
- $Q(\rho) = \frac{\varepsilon}{2} \begin{pmatrix} g(\rho^-) & (c_1 - c_2)f(\rho^+) \\ (c_1 - c_2)f(\rho^-) & g(\rho^+) \end{pmatrix}$
- $\bar{\rho}_j^n \approx \frac{1}{\Delta x} \int_{C_j} \rho(x, t^n) dx$  : cell averages over  $C_j := (x_{j-\frac{1}{2}}, x_{j+\frac{1}{2}})$
- The solution is approximated by a piecewise linear conservative, second-order accurate, non-oscillatory reconstruction:

$$\tilde{\rho}^n(x) = \bar{\rho}_j^n + (\rho_x)_j^n (x - x_j) \quad \text{for } x \in C_j$$

The solution is evolved by the semi-discrete central-upwind scheme

[Kurganov, Tadmor; 2000]

[Kurganov, Noelle, Petrova; 2001]

[Kurganov, Lin; 2007]

$$\frac{d\bar{\rho}_j(t)}{dt} = -\frac{H_{j-\frac{1}{2}}(t) - H_{j+\frac{1}{2}}(t)}{\Delta x} + \frac{P_{j-\frac{1}{2}}(t) - P_{j+\frac{1}{2}}(t)}{\Delta x}$$

with the numerical fluxes given by

$$H_{j+\frac{1}{2}} = \frac{a_{j+\frac{1}{2}}^+ F(\rho_j^E) - a_{j+\frac{1}{2}}^- F(\rho_{j+1}^W)}{a_{j+\frac{1}{2}}^+ - a_{j+\frac{1}{2}}^-} + \frac{a_{j+\frac{1}{2}}^+ a_{j+\frac{1}{2}}^-}{a_{j+\frac{1}{2}}^+ - a_{j+\frac{1}{2}}^-} [\rho_{j+1}^W - \rho_j^E]$$

$$P_{j+\frac{1}{2}} = Q(\rho_{j+\frac{1}{2}}) \frac{\bar{\rho}_{j+1} - \bar{\rho}_j}{\Delta x}, \quad \rho_{j+\frac{1}{2}} = \frac{\rho_j^E + \rho_{j+1}^W}{2}$$

The reconstructed point values are

$$\rho_j^E := \bar{\rho}_j + \frac{\Delta x}{2}(\rho_x)_j, \quad \rho_j^W := \bar{\rho}_j - \frac{\Delta x}{2}(\rho_x)_j$$

The discontinuities appearing at the reconstruction step at the interface points  $\{x_{j+\frac{1}{2}}\}$  propagate at finite speeds estimated by

$$a_{j+\frac{1}{2}}^+ := \max \left\{ \lambda_2 \left( \frac{\partial \mathbf{F}}{\partial \mathbf{U}}(\rho_{j+\frac{1}{2}}^E) \right), \lambda_2 \left( \frac{\partial \mathbf{F}}{\partial \mathbf{U}}(\rho_{j+\frac{1}{2}}^W) \right), 0 \right\}$$

$$a_{j+\frac{1}{2}}^- := \min \left\{ \lambda_1 \left( \frac{\partial \mathbf{F}}{\partial \mathbf{U}}(\rho_{j+\frac{1}{2}}^E) \right), \lambda_1 \left( \frac{\partial \mathbf{F}}{\partial \mathbf{U}}(\rho_{j+\frac{1}{2}}^W) \right), 0 \right\}$$

$\lambda_1 < \lambda_2$ : eigenvalues of the Jacobian  $\frac{\partial \mathbf{F}}{\partial \rho}$

The eigenvalues of the Jacobian matrix are calculated as follows...

We denote by

$$R = f'(\rho^+)g(\rho^-) - f'(\rho^-)g(\rho^+)$$

$$D = [f'(\rho^-)g(\rho^+) + f'(\rho^+)g(\rho^-)]^2 - 4f(\rho^-)f(\rho^+)g'(\rho^-)g'(\rho^+)$$

and consider the two possible cases:

- If both  $D_j^E \geq 0$  and  $D_{j+1}^W \geq 0$  (**hyperbolic regime**), then

$$a_{j+\frac{1}{2}}^+ = \frac{1}{2} \max \left\{ R_j^E + \sqrt{D_j^E}, R_{j+1}^W + \sqrt{D_{j+1}^W}, 0 \right\}$$

$$a_{j+\frac{1}{2}}^- = \frac{1}{2} \min \left\{ R_j^E - \sqrt{D_j^E}, R_{j+1}^W - \sqrt{D_{j+1}^W}, 0 \right\}$$

- If either  $D_j^E < 0$  or  $D_{j+1}^W < 0$  (**nonhyperbolic regime**), then

$$a_{j+\frac{1}{2}}^+ = \frac{1}{2} \max \left\{ \sqrt{(R_j^E)^2 - D_j^E}, \sqrt{(R_{j+1}^W)^2 - D_{j+1}^W} \right\}$$

$$a_{j+\frac{1}{2}}^- = -a_{j+\frac{1}{2}}^+$$

- The choice of one-sided local speeds in the nonhyperbolic regime is ad-hoc
- We have not tried to stabilize the inviscid PDE solution by increasing the amount of numerical viscosity
- The solution has been stabilized by adding nonlinear diffusion terms rigorously derived from the mesoscopic formulation

## Example — “Red Light” Initial Conditions

Two (relatively small) groups of pedestrians standing still and starting to move toward each other at time  $t = 0$ :

- **Microscopic model:**

$$c_0 = 0.8m/s, \quad c_1 = c_2 = c_0/a, \quad c_3 = c_0/(2a), \quad a = 2 \quad \text{or} \quad a = 3$$

$$\sigma^+(k, 0) = \begin{cases} 1, & n_1 \leq k \leq n_2 \\ 0, & \text{otherwise} \end{cases} \quad \sigma^-(k, 0) = \begin{cases} 1, & N - n_2 \leq k \leq N - n_1 \\ 0, & \text{otherwise} \end{cases}$$

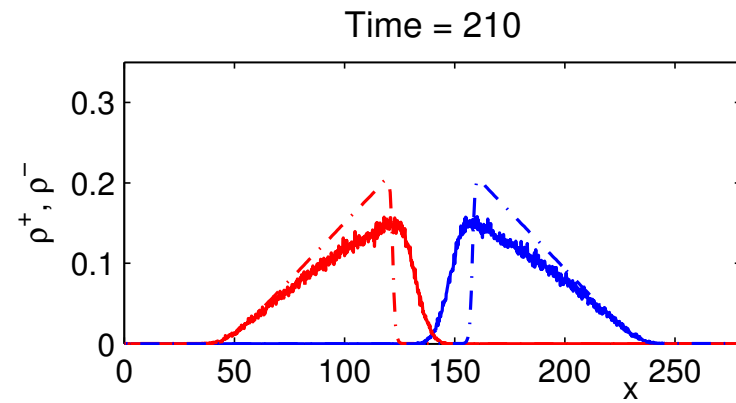
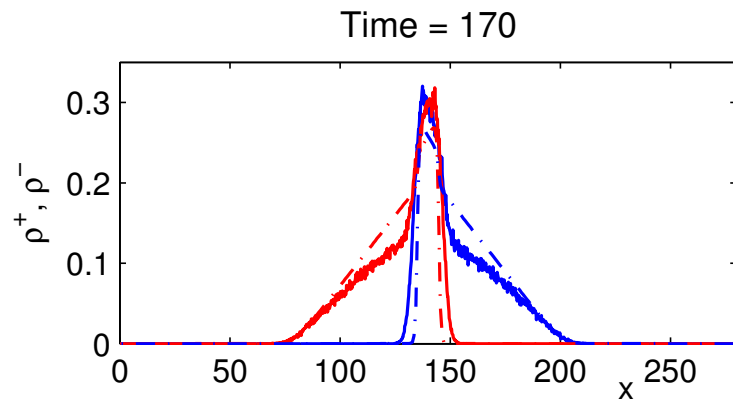
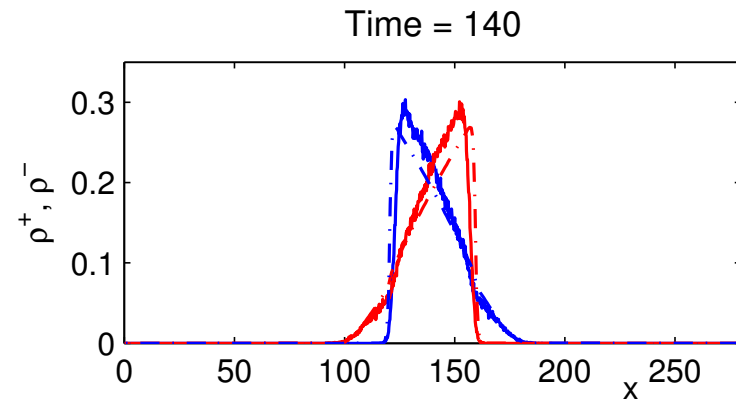
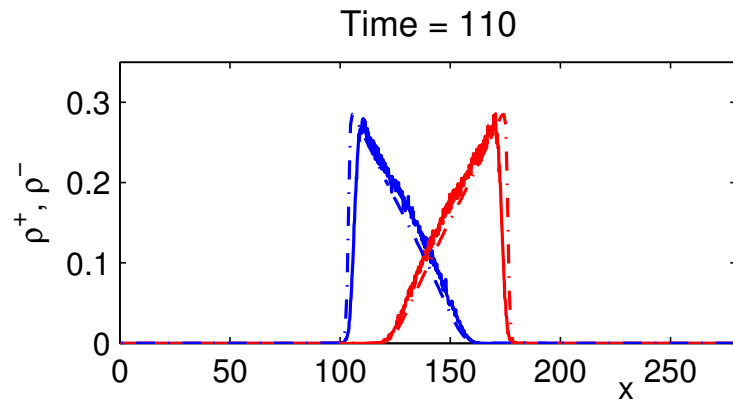
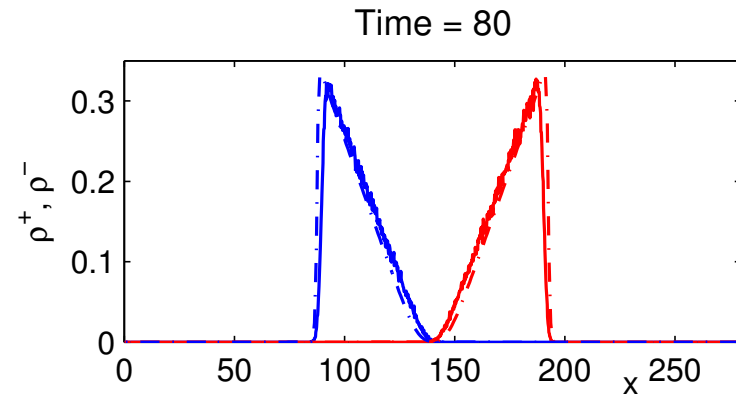
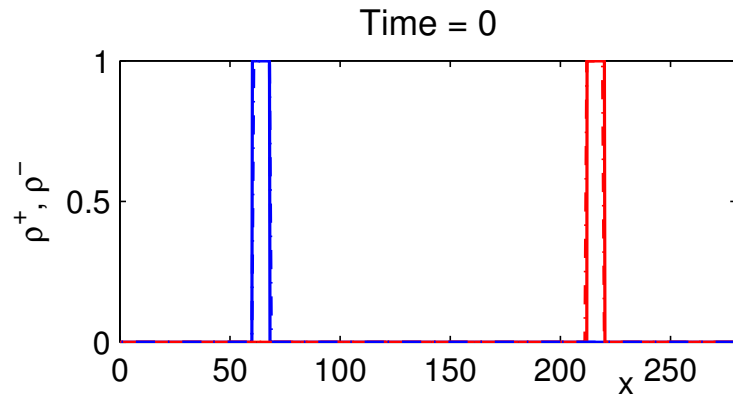
$$N = 1400, \quad n_1 = 301, \quad n_2 = 340, \quad h = 0.2m, \quad \Delta t = 0.01s, \quad MC = 5000$$

- **Macroscopic model:**

$$\rho^+(x, 0) = \begin{cases} 1, & 60 < x < 68 \\ 0, & \text{otherwise} \end{cases} \quad \rho^-(x, 0) = \begin{cases} 1, & 212 < x < 220 \\ 0, & \text{otherwise} \end{cases}$$

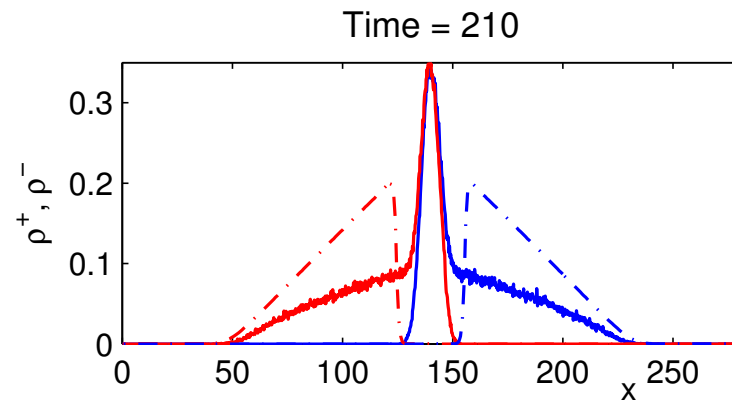
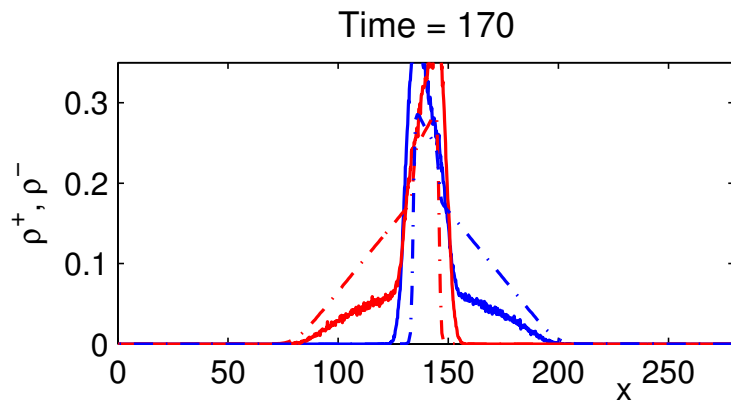
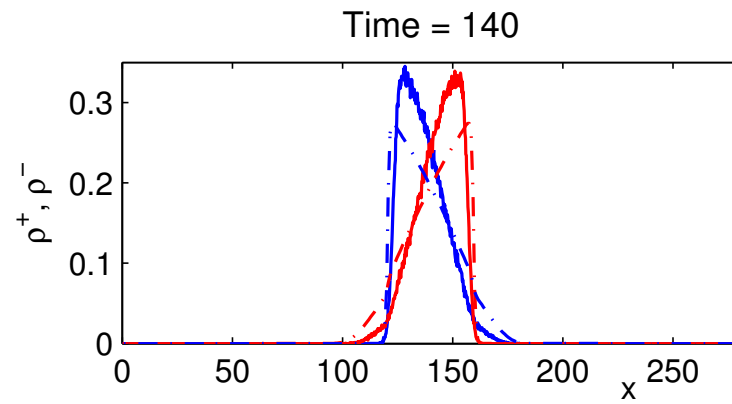
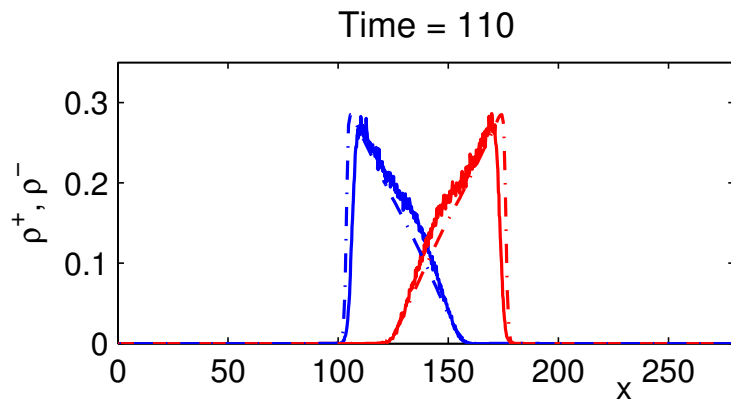
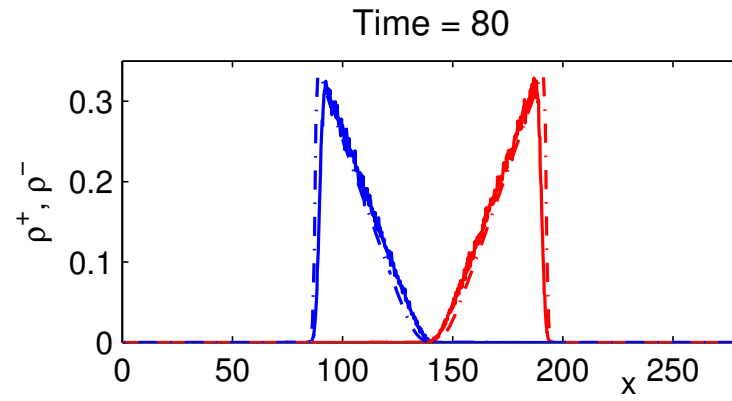
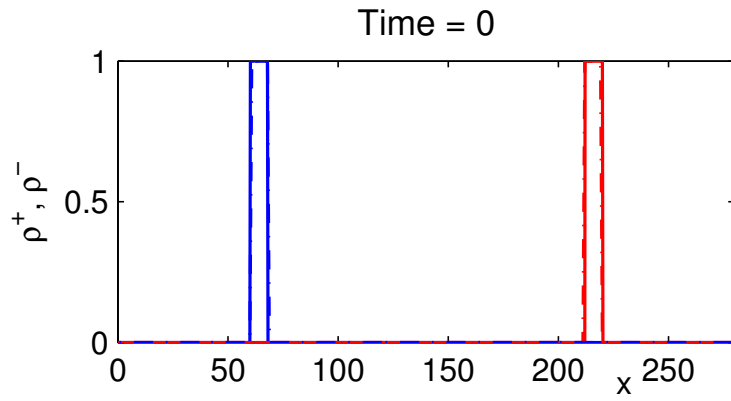
$$L = 280, \quad \Delta x = 0.8$$

$$a = 2$$





$$a = 3$$



## Example — Fully Mixed Initial Conditions

Pedestrian movement in a periodic domain, which is divided into 30 sectors with 15 cells in each sector (totally  $N = 450$  cells)

- **Microscopic model:**

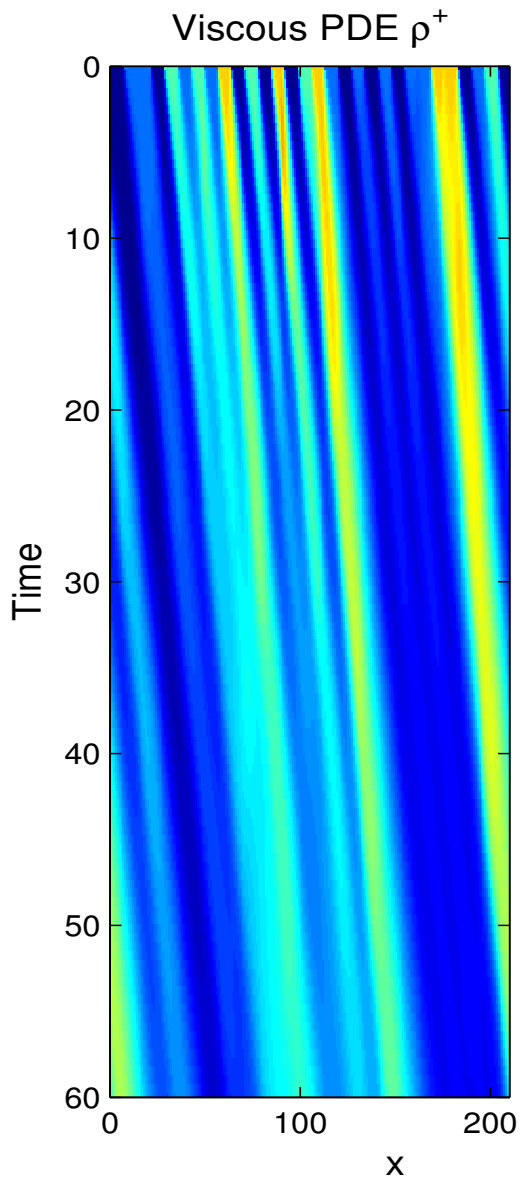
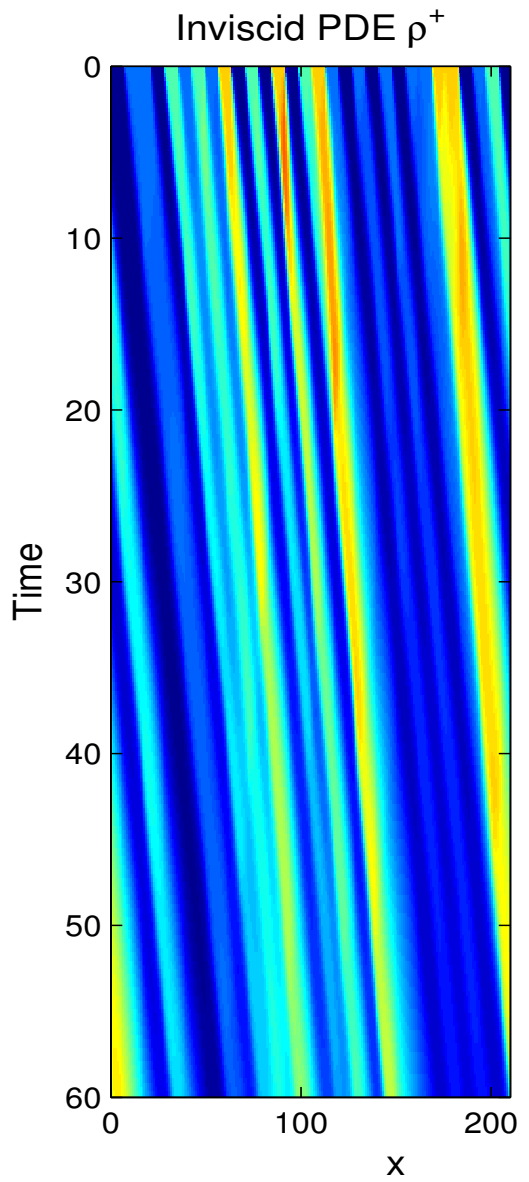
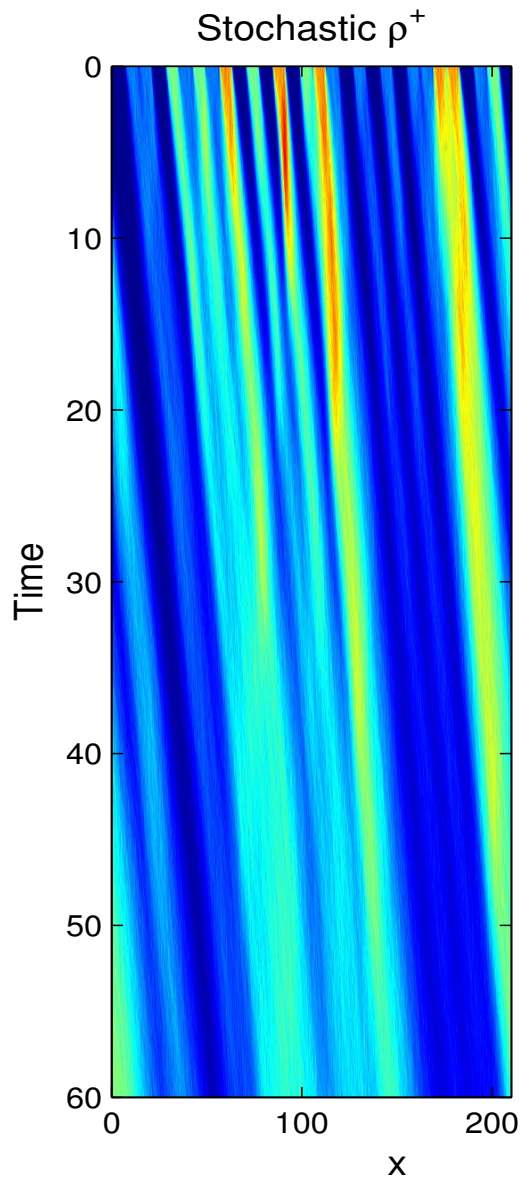
- the total number of 70 pedestrians (35 moving in each direction)
- the initial distribution of pedestrians is purely random (uniform)
- sectors are  $7m$  long;  $L = 210m$ ;  $\Delta t = 0.005$ ,  $MC = 3000$

- **Macroscopic model:**

$$\rho^\pm(x, 0) = \frac{n_i^\pm}{15} \quad \text{for} \quad \frac{i-1}{30}L < x < \frac{i}{30}L, \quad i = 1, \dots, 30$$

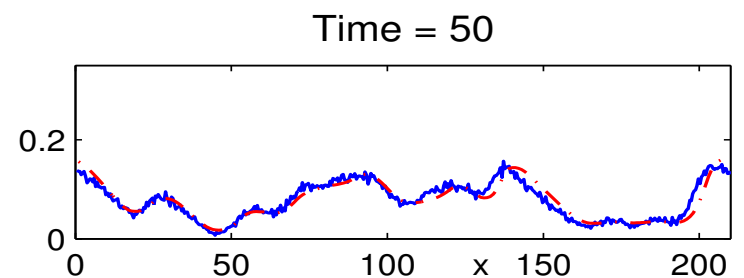
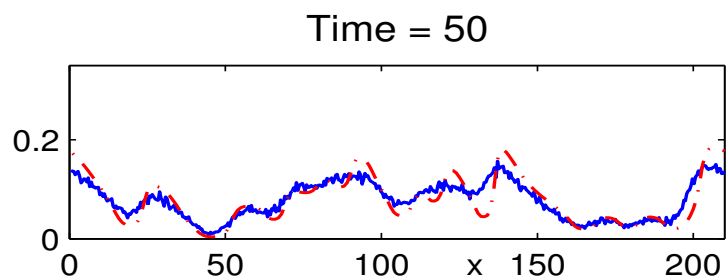
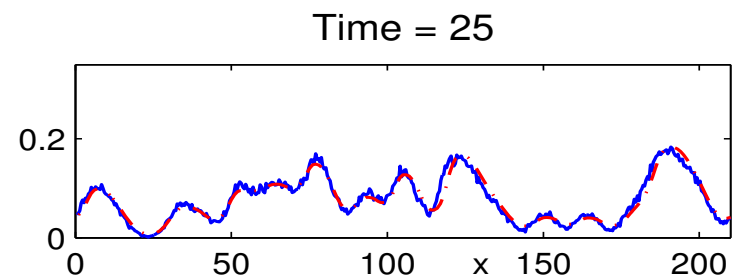
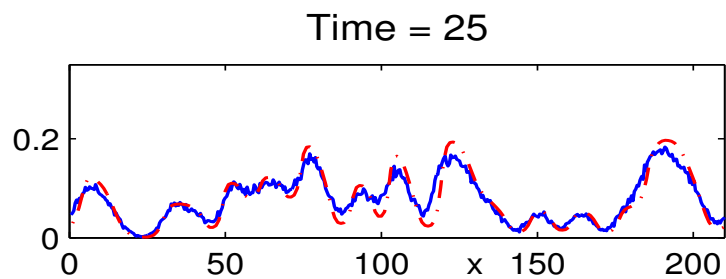
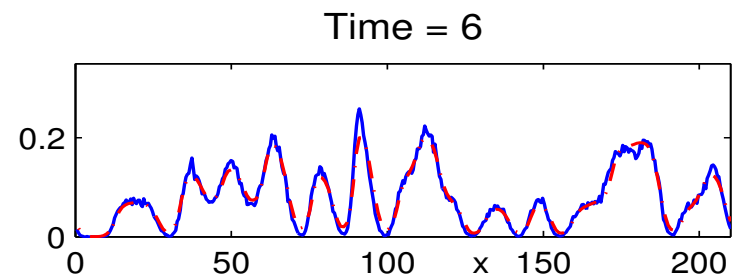
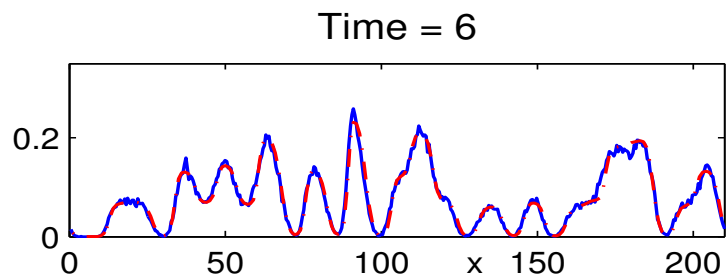
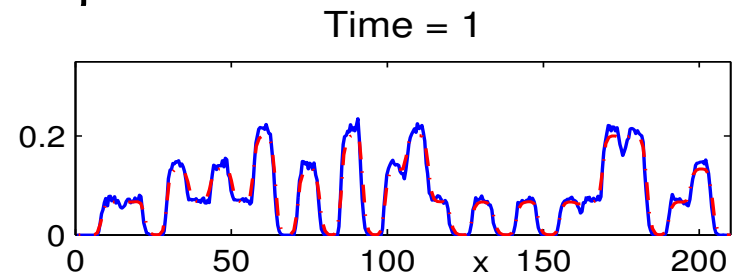
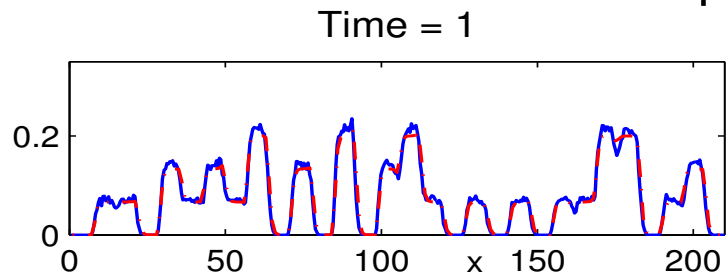
- $n_i^+$  and  $n_i^-$ : right- and left-moving pedestrians in the  $i$ th sector
- $c_0 = 1m/s$ ,  $c_1 = c_2 = c_0/a$ ,  $c_3 = c_0/(2a)$ ,  $a = 2$  or  $a = 3$
- $L = 210$ ,  $\Delta x = 1$ ,  $\varepsilon = 0.5$

$a = 2$  (right-moving pedestrians)

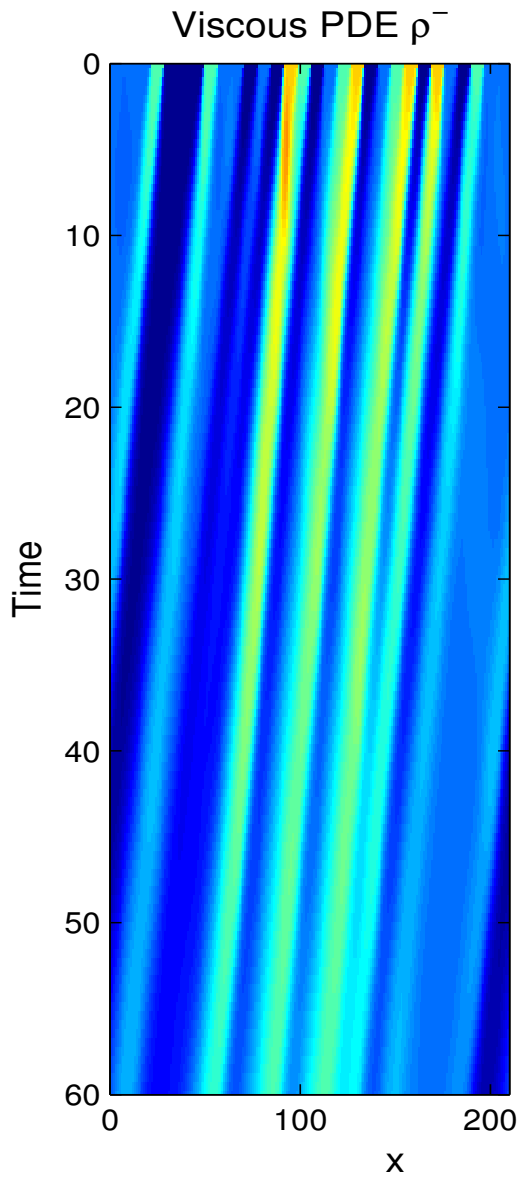
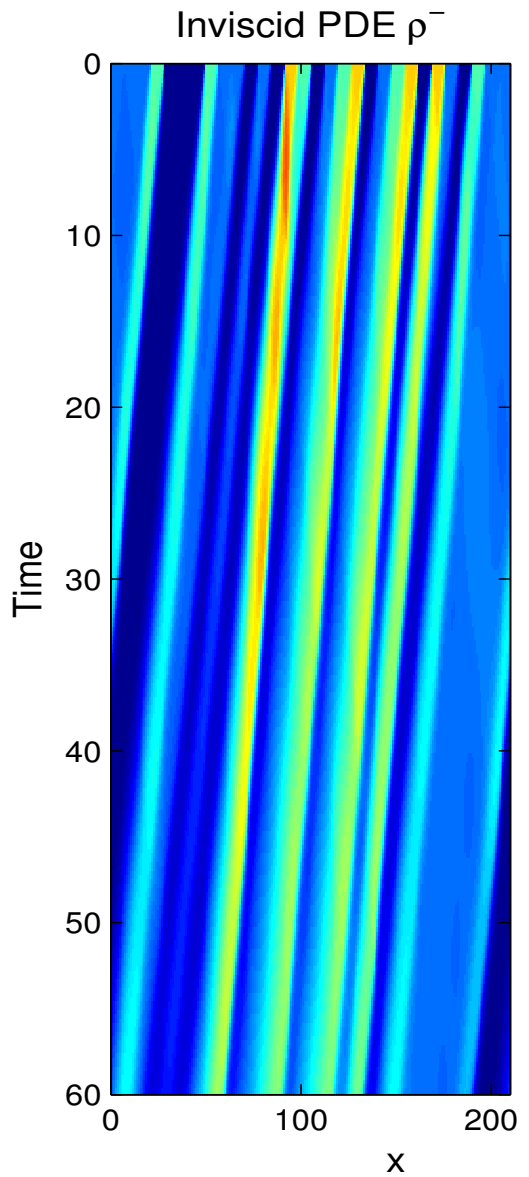
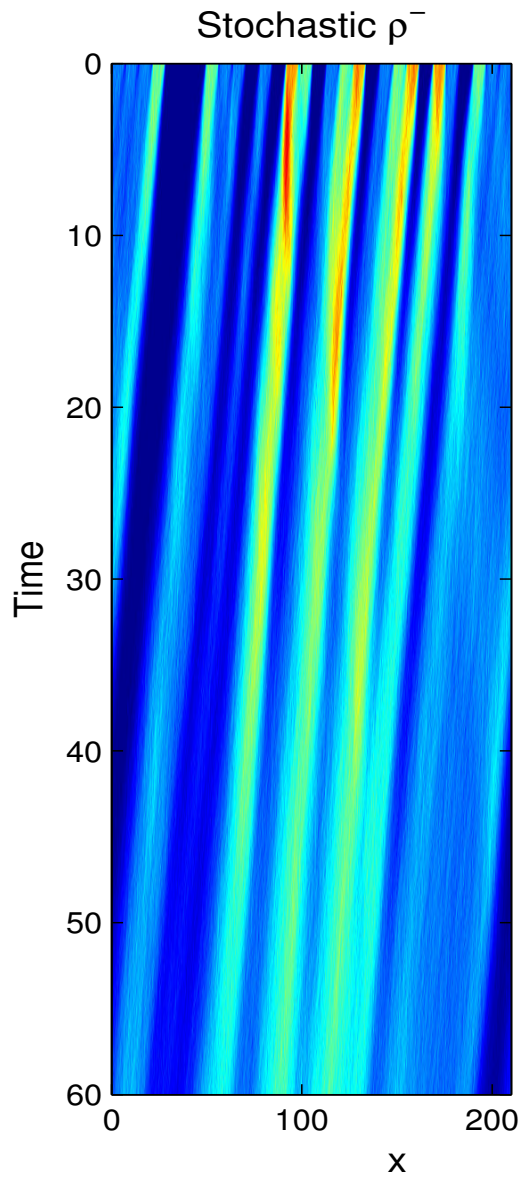


$a = 2$  (right-moving pedestrians)

### Snapshots of $\rho^+$

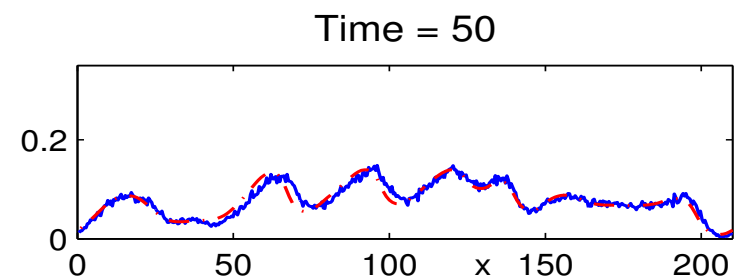
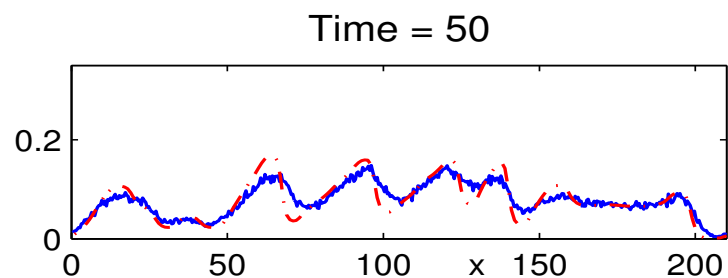
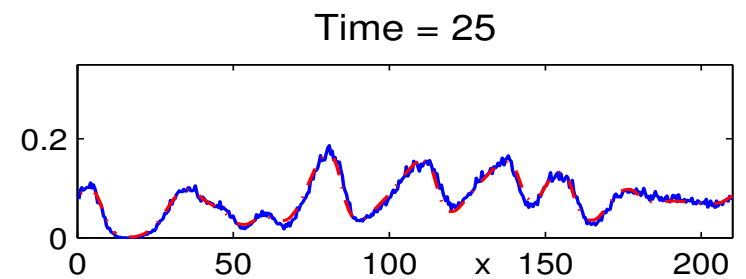
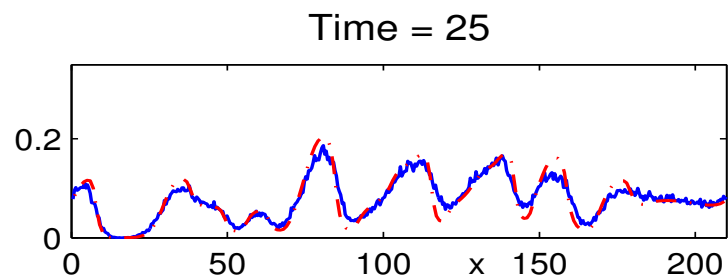
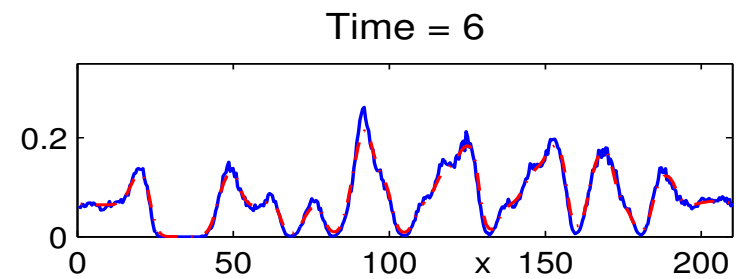
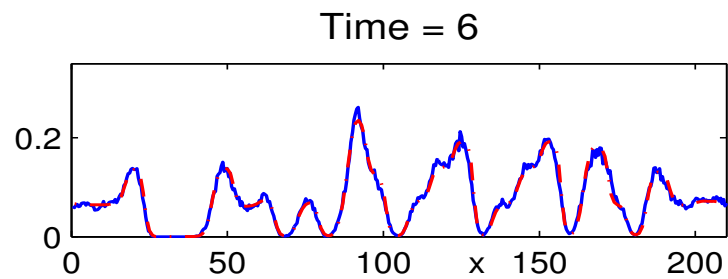
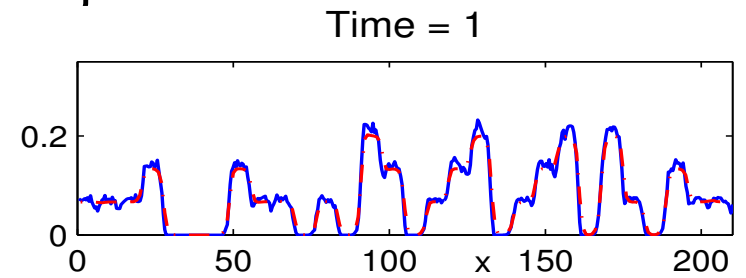
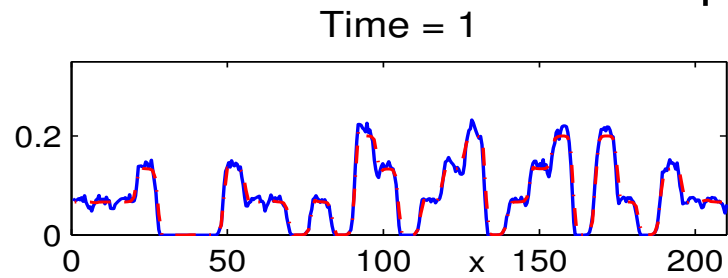


$a = 2$  (left-moving pedestrians)

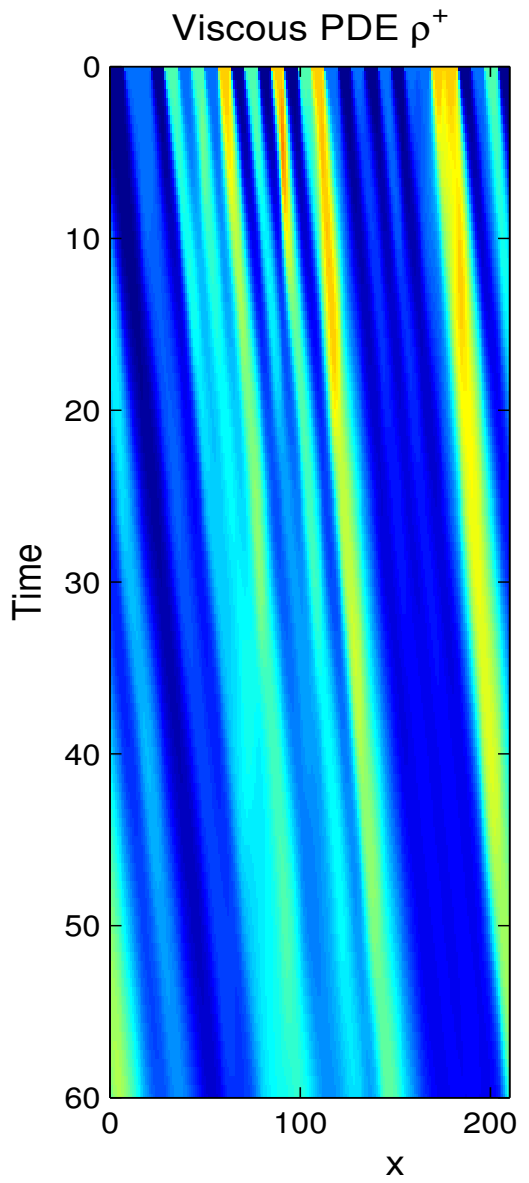
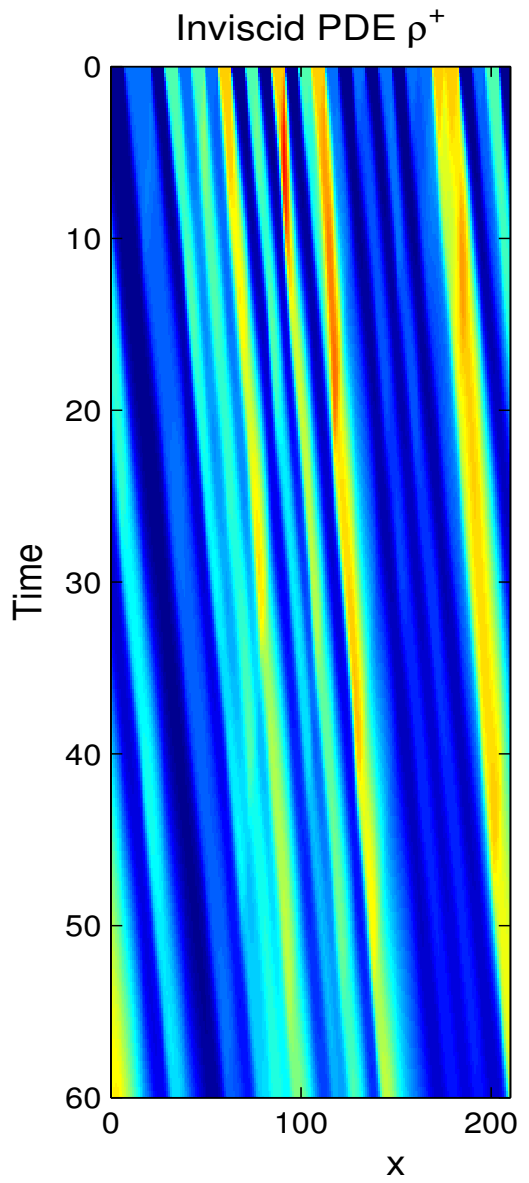
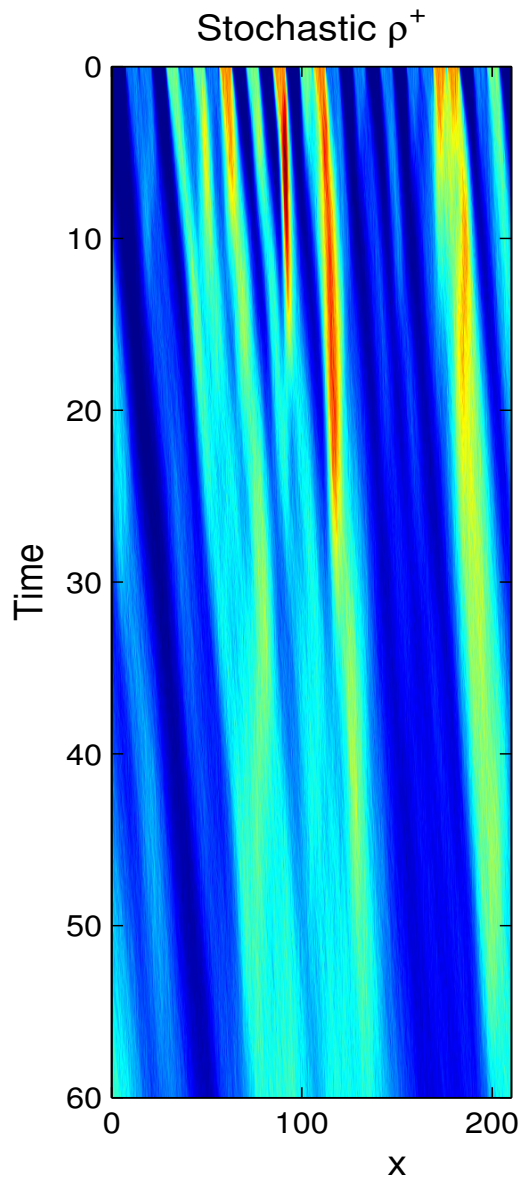


$a = 2$  (left-moving pedestrians)

### Snapshots of $\rho^-$

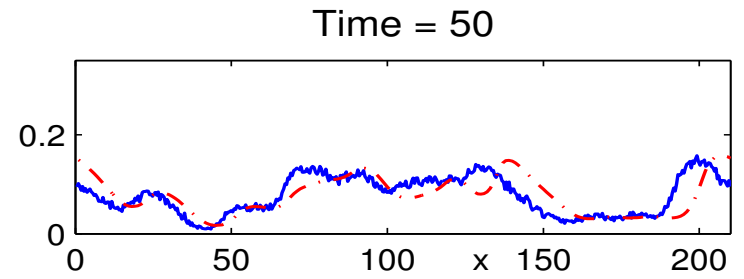
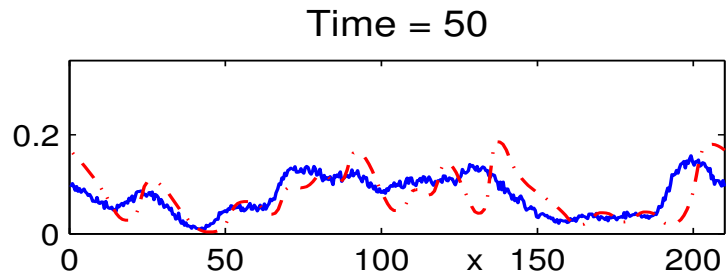
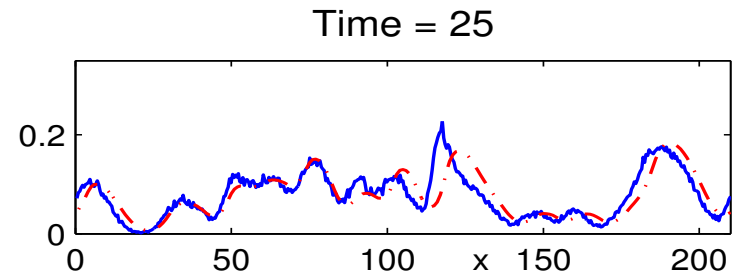
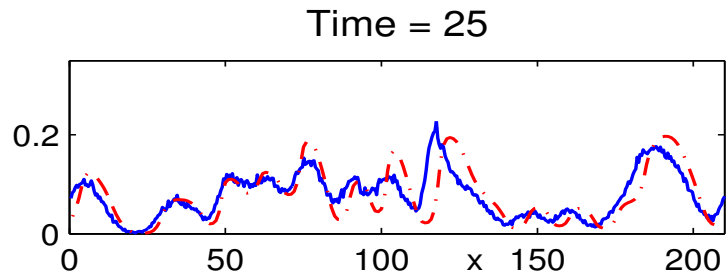
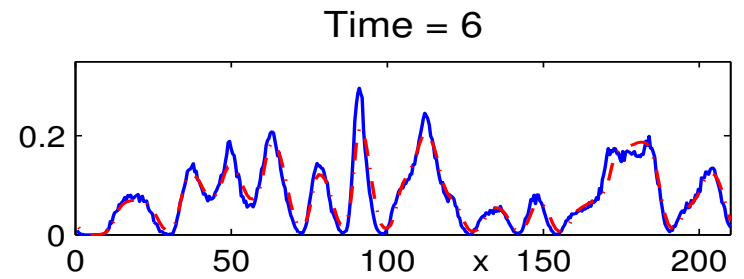
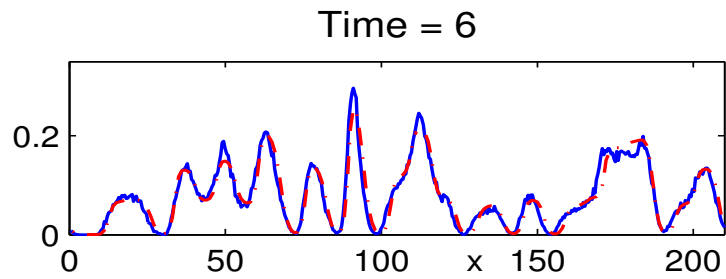
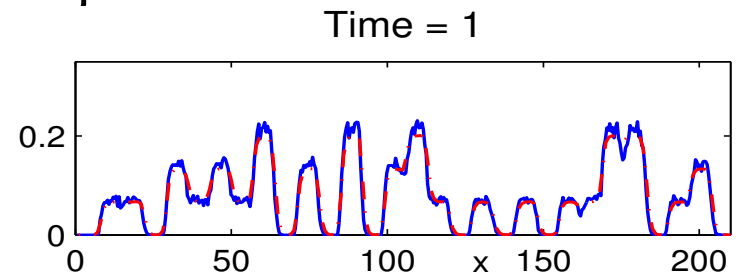
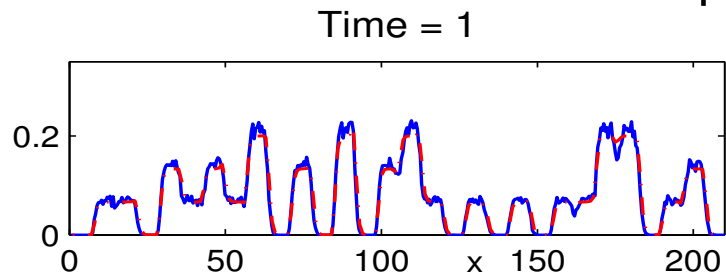


$a = 3$  (right-moving pedestrians)



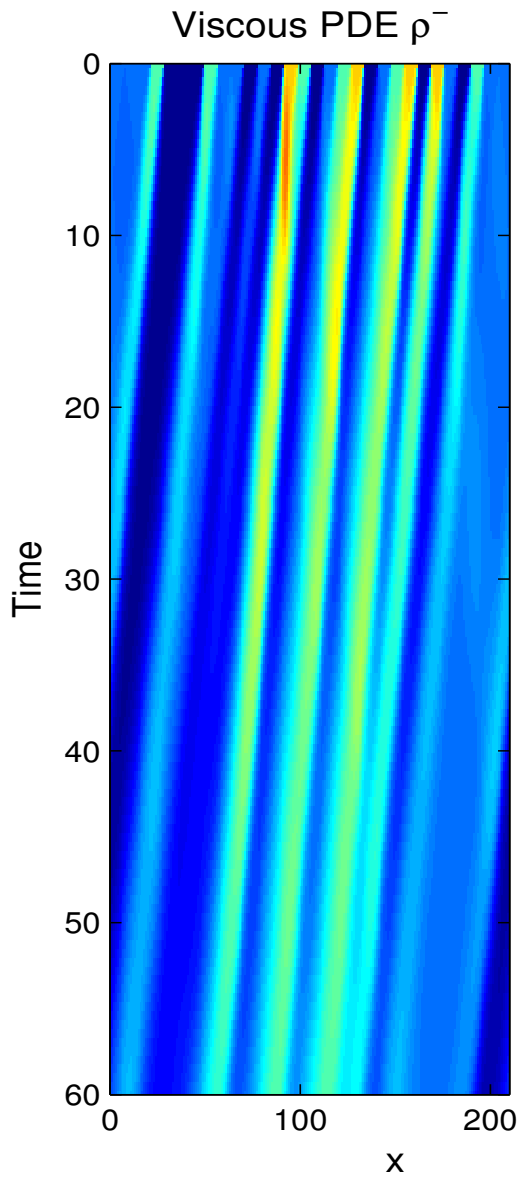
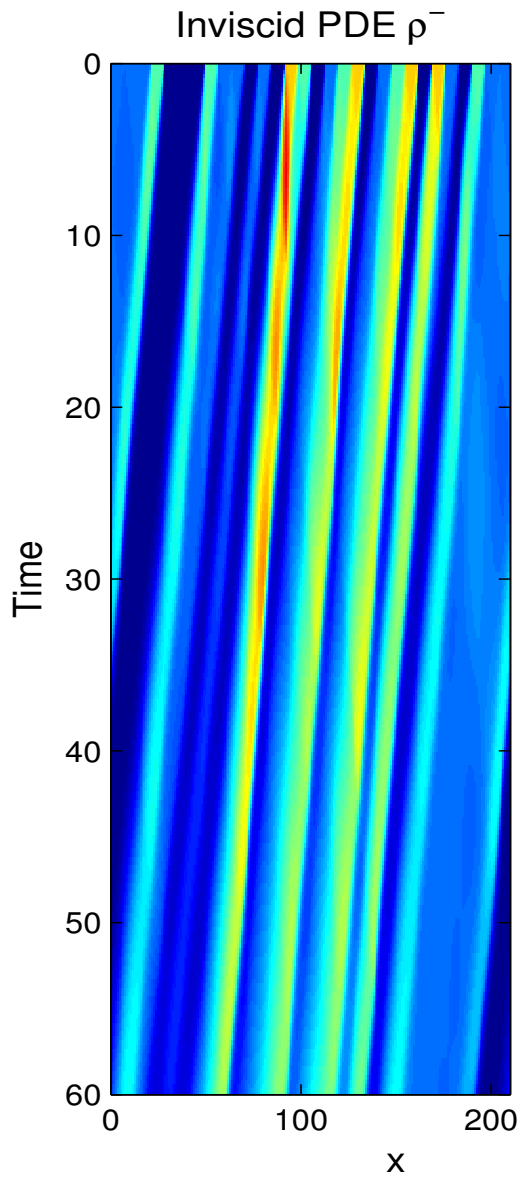
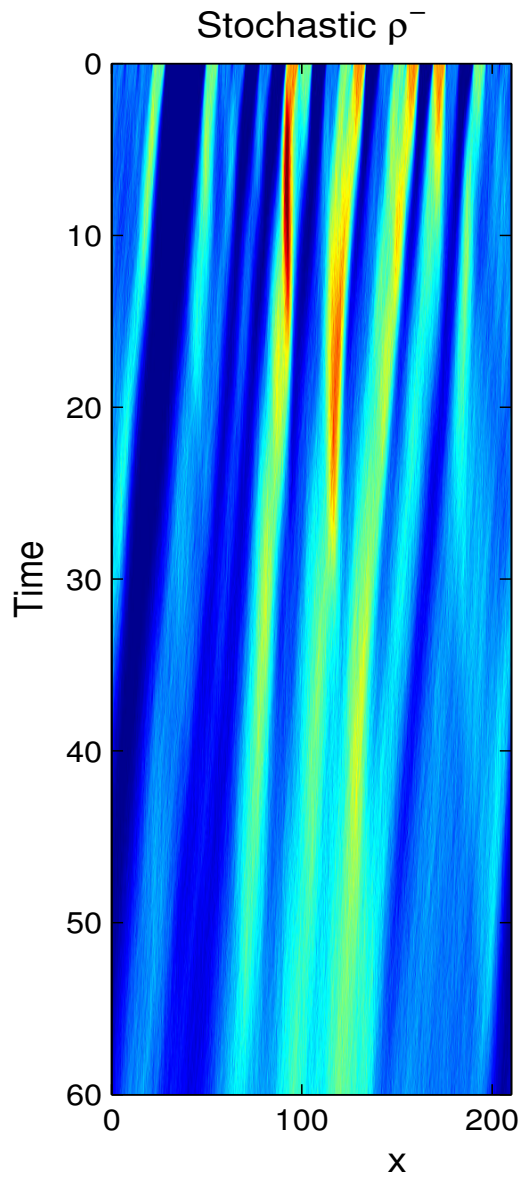
$a = 3$  (right-moving pedestrians)

### Snapshots of $\rho^+$



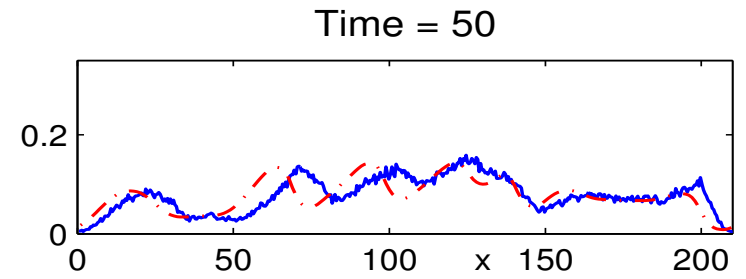
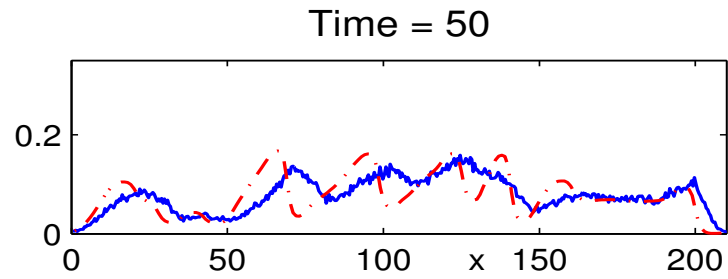
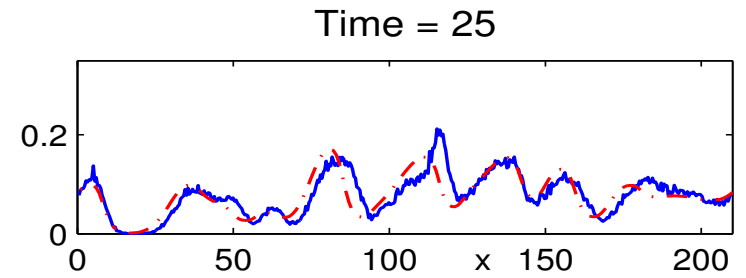
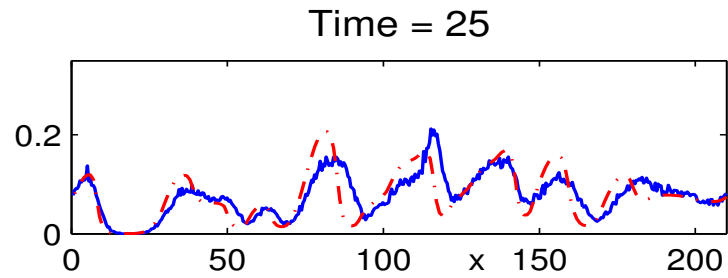
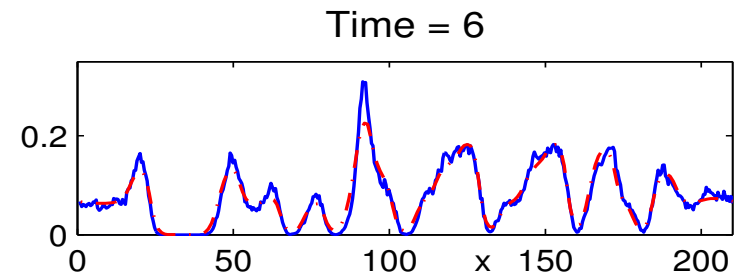
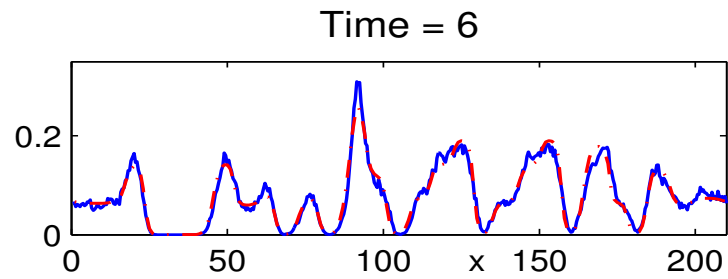
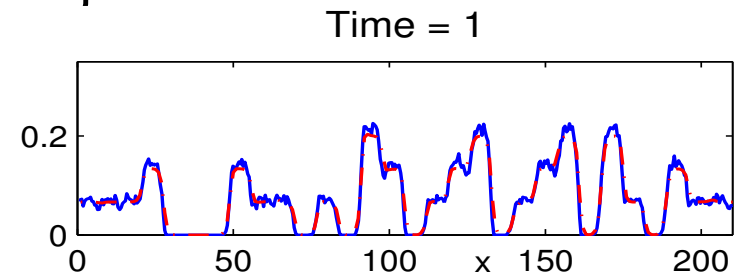
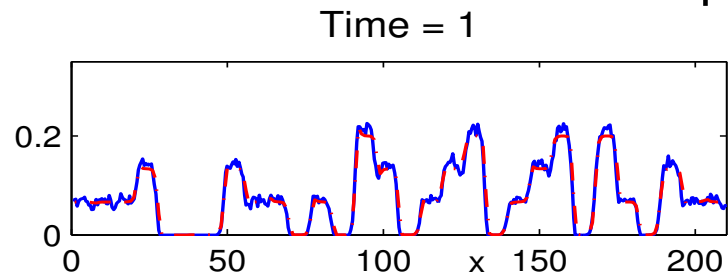


$a = 3$  (left-moving pedestrians)



$a = 3$  (left-moving pedestrians)

### Snapshots of $\rho^-$



## Example — Nonhyperbolic Regime

Pedestrian movement in a periodic domain with velocities

$$c_0 = 1m/s, \quad c_1 = c_2 = c_0/a, \quad c_3 = c_0/(2a), \quad a = 2 \quad \text{or} \quad a = 3$$

- Microscopic model:

- the number of cells is  $N = 900$

- the cell size is  $420/900 \approx 0.4667m$

- the time step is  $\Delta t = 0.005$  and  $MC = 3000$

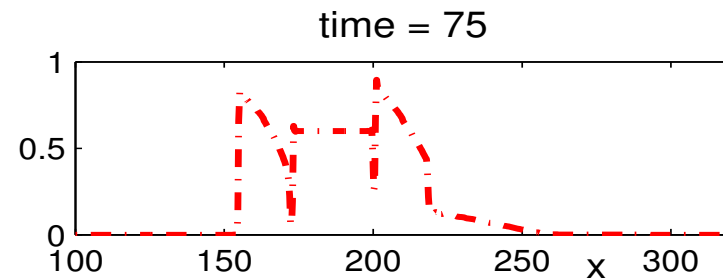
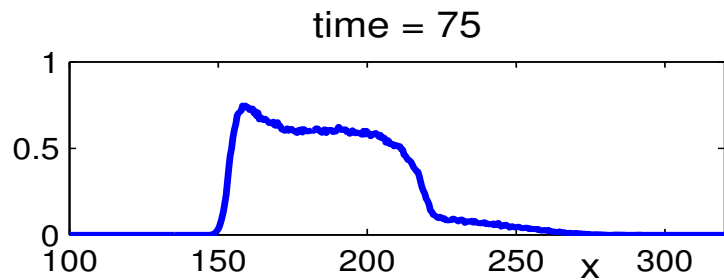
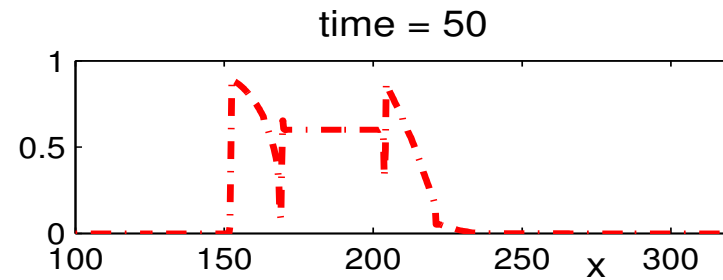
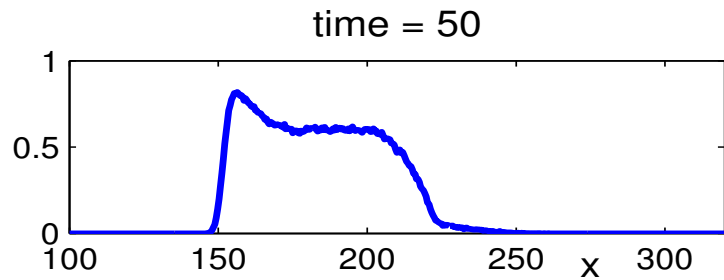
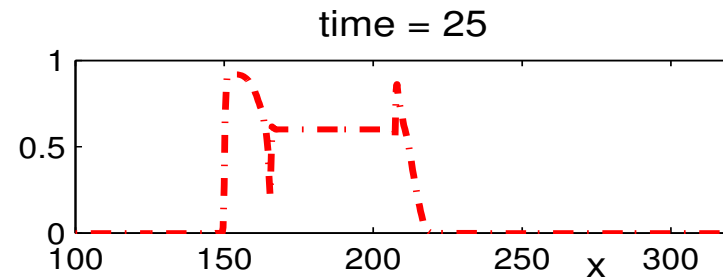
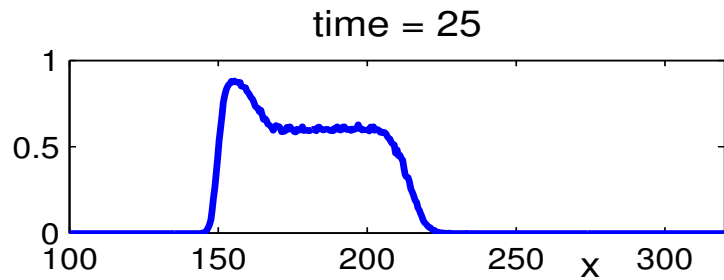
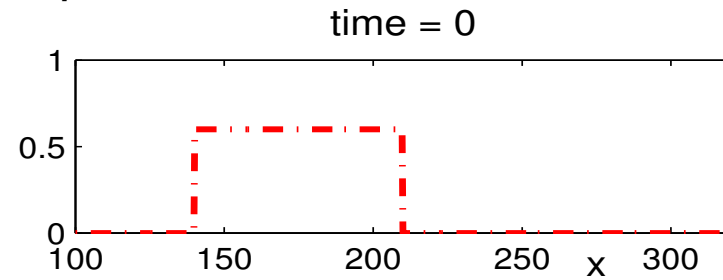
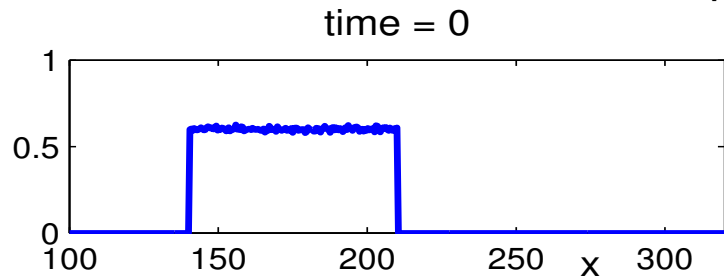
- Macroscopic model:

$$\rho^+(x, 0) = \begin{cases} 0.6, & 140 \leq x \leq 210 \\ 0, & \text{otherwise} \end{cases} \quad \rho^-(x, 0) = \begin{cases} 0.6, & 186.6 \leq x \leq 233.3 \\ 0, & \text{otherwise} \end{cases}$$

$$L = [0, 420], \quad \Delta x = 420/1280$$

$a = 2$  (right-moving pedestrians), CA vs. inviscid PDE

Snapshots of  $\rho^+$



$a = 2$  (right-moving pedestrians), CA vs. viscous PDE,  $\varepsilon = 0.5, 1.5$

Snapshots of  $\rho^+$

