



Fluid Models for Dense Crowd Tracking



Progress Report on a New Effort (Start: Fall 2009)

Oliver Lehmann and Gilead Tadmor, ECE & Math, Northeastern University, tadmor@coe.neu.edu

The Complexity Challenge

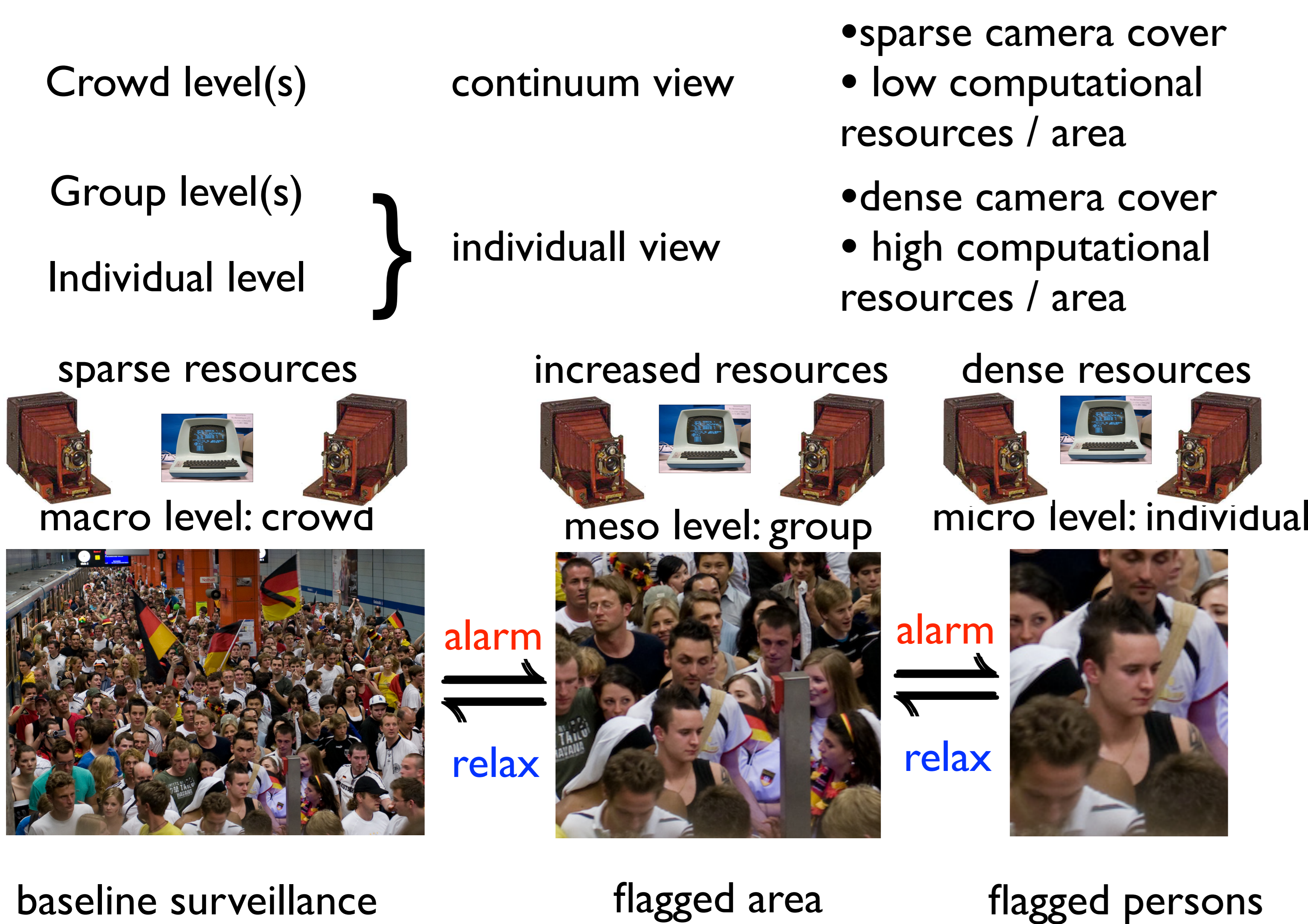
Complexity precludes individual tracking & behavior analysis in very large and dense crowd scenes

Examples of Driving Scenarios:

- Spectators entering or leaving a sport events
- Downtown sidewalk at noon
- Central underground station
- A busy airport terminal
- An open air market
- Large political rallies

Big Picture Objective: Dynamic Resource Allocation

- **Develop low resolution tracking for dense crowds**
- **Integrate large and low density methods**



State of the Art in Crowd Flow Models

- Off-line simulation-based prediction and design optimization of safe public environment
- Off-line crowd behavior analysis from video data
- **Real-time crowd behavior analysis:** still “in diapers”
- Animation, motion synthesis

Our Choice of a Modeling Paradigm

Option I: Granular flow / statistical mechanics ✗

- Advantages:
 - Well developed over 25 years
 - Natural representation of social behaviors
- Disadvantages:
 - Combinatorial complexity exclude real time utility

Option II: Compressible fluid flow / fluid mechanics ✓

- Advantages:
 - Far simpler model
 - Galerkin approximations facilitate real time estimation
- Disadvantages:
 - A challenge to capture social phenomena with few parameters

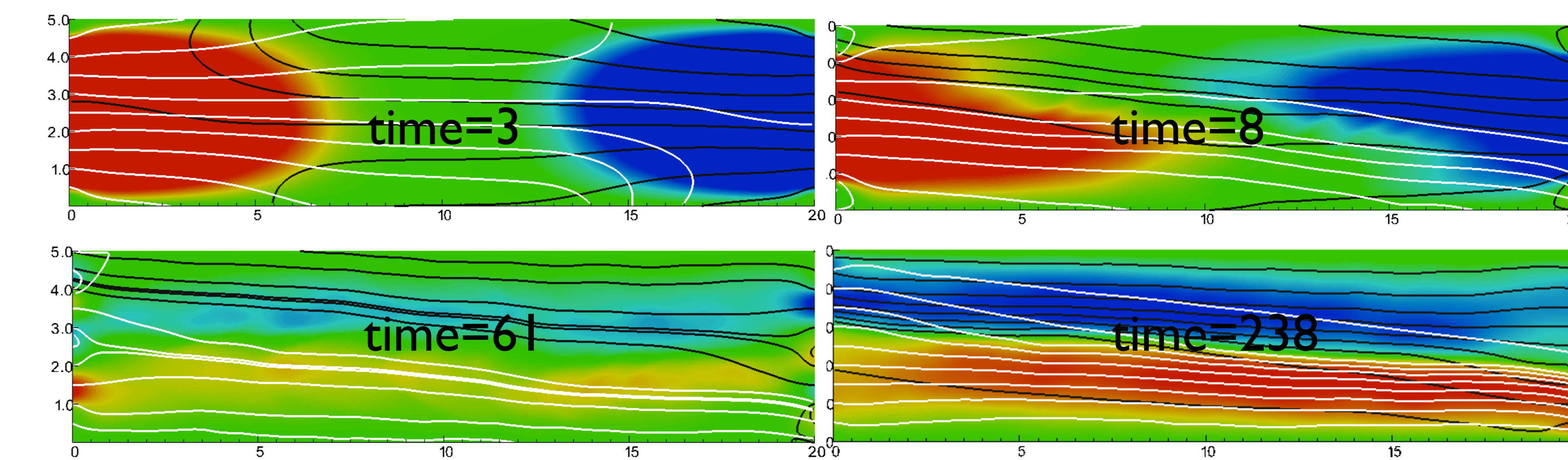
Achievements To Date

Developed, refined, coded & validated a parameterized Navier-Stokes inspired model structure

- Multiple populations with
 - $\rho_i(\mathbf{x})$: distributed density of population # i
 - $\mathbf{u}_i(\mathbf{x})$ distributed velocity of population # i
 - $\mathbf{u}_{d,i}(\mathbf{x})$ distributed desired velocity of population # i
 - $\rho(\mathbf{x}) = \sum \rho_i(\mathbf{x})$: combined density
 - $\mathbf{u}(\mathbf{x}) = \rho(\mathbf{x}) \setminus \sum \rho_i(\mathbf{x}) \mathbf{u}_i(\mathbf{x})$: locally averaged velocity
 - $\tau_i(\mathbf{u}_i)$ = time constant of tracking desired path

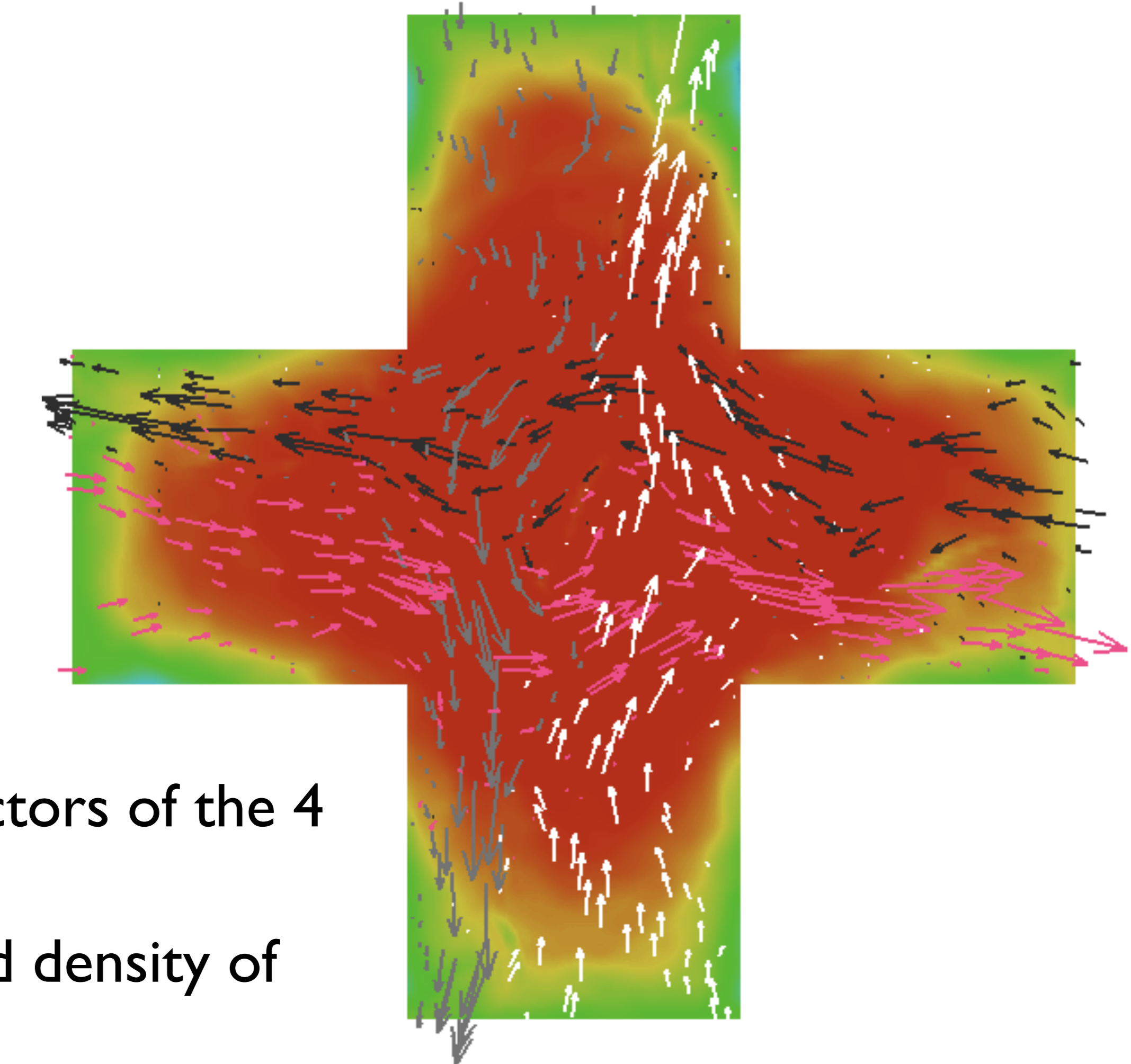
Validation: Reproducing Generic Social Patterns

Example: Spontaneous lane formations by pedestrian walking in opposite directions, in a corridor



- pedestrians enter, using the entry's full width
- color code: total mass flux. Red: left-to-right population dominates. Blue: Right-to-left population dominates.
- White curves: Streamlines of left-to-right population. Black curves: Same for right-to-left population

Example: A 4 populations / crossing corridors variant

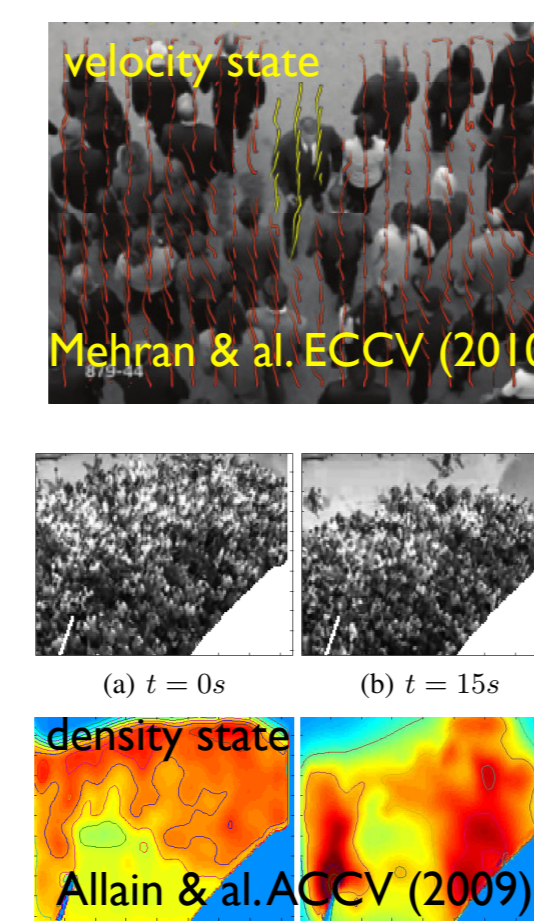


- Arrows: momenta vectors of the 4 populations
- Color code: combined density of all pedestrians

Background: Physically Inspired Flow Models

Guiding Principle: Simplicity @ the limit to ∞

- Shift from individuals to mass flow
- State components:
 - the distributed velocity field
 - crowd / sub-population densities
- Physically inspired models:
 - statistical mechanics: granular flow models
 - fluid mechanics: continuum models
- Behavioral pattern \Leftrightarrow Values of model parameters
- Tracking \Leftrightarrow Real time model parameter estimation
- Event detection \Leftrightarrow Model invalidation, abrupt parameter changes



$$\partial_t (\rho_i \mathbf{u}_i) = \underbrace{-\nabla \cdot (\rho_i \mathbf{u}_i) \otimes \mathbf{u}_i}_{\text{Convection term}} + \underbrace{\frac{\rho_i}{\rho} (\mu \Delta \mathbf{u} - \nabla p)}_{\text{viscosity \& pressure forces}}$$

$$+ \underbrace{\frac{\rho_i}{\tau_i} (\mathbf{u}_{d,i} - \mathbf{u}_i)}_{\text{driving objective \& ground friction in population \# i}} + \underbrace{\rho_i \eta_i}_{\text{stochastic acceleration in population \# i}}$$

Momentum Eq

$$p = \frac{1}{1 - \frac{\rho}{\rho_{sat}}}$$

$$\partial_t \rho + \nabla \rho \cdot \mathbf{u} + \rho \nabla \cdot \mathbf{u} = 0$$

repelling potential ("pressure") **mass conservation**

Present & Near Future Effort

- Real time model parameter estimation from benchmark video data sets
- Model refinement to match a large number of populations and complex geometries. Driving scenario: Pedestrians heading to different platforms and exits in an underground station
- Large influx / density variations
- Stochastic dominance over targeted motion (an open market scenario)