# Why are we studying crowds?

### Real Crowds

- Design, dimensions of public building
  - Normal usage
  - Emergency evacuation





# Why are we studying crowds?

## Virtual Crowds

- Video games
- Movies, ads
- Participative democracy



Le Seigneur des Anneaux



Crowd simulation software

Avatar

## Data

- https://ped.fz-juelich.de/db
  - Collection of data from experiments involving the Forschungszentrum Jülich Includes large scale experiment of stadium evacuation
  - Links to other experimental data obtained all over the world

# Experiment in a stadium



Project HERMES, cf [Burghardt, Seyfried, Klingsch, Transp. research Part C 37, 268 (2013)]

# Pedestrian experiments: tracking

## From Videos

- Easier if pedestrians wear a special hat
- Use of some tracking software
  - Example: Open Software PeTrack
    (https://www.fz-juelich.de/en/ias/ias-7/services/software/petrack)





From [Boltes et al, Neurocomput. (2013)]

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Exp. of [Nicolas & al, Sci. Rep. (2019)]



# Pedestrian experiments: tracking





## From Depth-Field movies

- Preserves more anonymity
- Kinect (Easy, narrow field of view, works only inside)
- Double cameras (Synchronization needed)

# High precision motion capture (VICON)

- Markers detected by infrared cameras
- Used for virtual production, in virtual reality, etc







# High precision motion capture (VICON)

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From the PEDIGREE Project



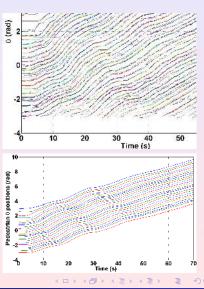




# Reconstruction of trajectories

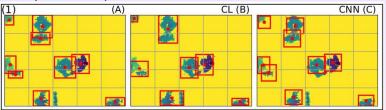


- From raw data to 3D markers' trajectories
- From markers to pedestrians
- Interpolating for missing data



# Deep learning

#### Example with depth field measurements

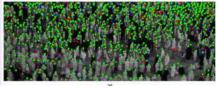


From [Corbetta et al, AVSS 2017]

- (A) = synthetic data (ground truth)
- (B) = CL = Clustering Approach
- (C) = CNN = Deep Convolution Neural Network

# Deep learning

## Intermediate and high densities





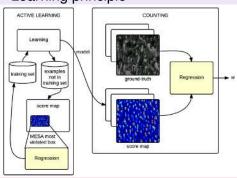
From [J. Vandoni, PhD Thesis, SATIE, Univ. Paris Saclay (2019)]

- (a) Makkah Dataset
- (b) Regent's Park Dataset
  - Red blobs = ground-truth heads
  - Green blobs = True positive
  - Blue blobs = False positive

# Deep learning

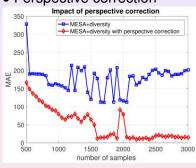
## Learning to estimate density without individual tracking

• Learning principle



From [J. Vandoni, PhD Thesis, Univ. Paris Saclay (2019)]

### • Perspective correction

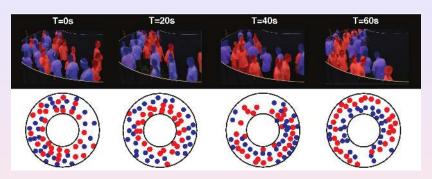


MAE = Mean Absolute Error (compared to ground truth)

# Ring



# Ring



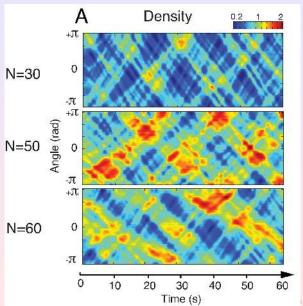
[M. Moussaïd, E. Guillot, M. Moreau, J. Fehrenbach, O. Chabiron, S. Lemercier, J. Pettré, C. Appert-Rolland, P. Degond and G. Theraulaz, *Traffic Instabilities in Self-organized Pedestrian Crowds*, PLoS Computational Biology (2012)]

Can we model pedestrian flows as fluids?



# Ring: Experimental density fields





## global density

0.59,

 $0.98 \text{ ped/} m^2$ 

1.18

# Macroscopic models for pedestrians

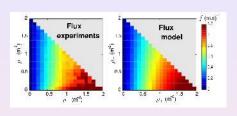


- $\blacktriangleright$  2 densities  $\rho_+$  and  $\rho_-$
- ightharpoonup 2 velocities  $u_+$  and  $u_-$

Possible to develop 1st or 2nd order macroscopic models, but with double number of equations.

## 1st order macroscopic model for pedestrians

## [S. Motsch et al, Math. Biosci. Eng. 15 (2018) 1271]

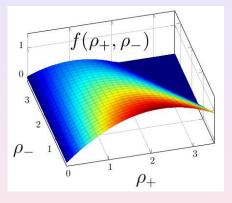


$$f(\rho_+, \rho_-) = a\rho_+ (1 - b\rho_+ - c\rho_-)$$

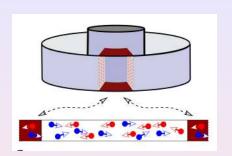
Sample set	a	b	С	$R^2$
50% - 50%	1.218	0.273	0.181	0.944
75% - 25%	1.216	0.087	0.203	0.972
100% - 0%	1.269	0.077	0	0.982

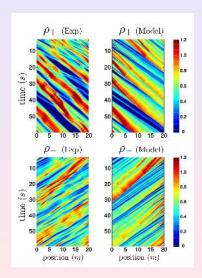


- b ∼ friction with co-moving pedestrians
- lacktriangle c  $\sim$  friction with counter-moving pedestrians



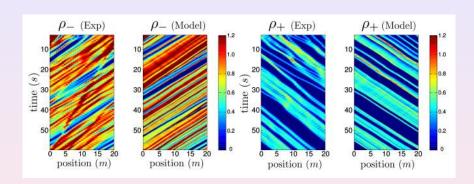
# Comparison Model/Experiment





Transport modeling

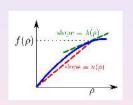
## Comparison Model/Experiment

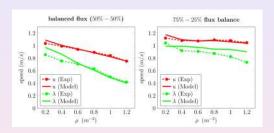


75% – 25%

From [Motsch et al, Math. Biosci. Eng. 15 (2018) 1271]

# Comparison Model/Experiment



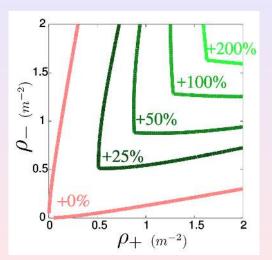


Velocity 
$$u(\rho) = f(\rho)/\rho$$
  
Cluster velocity  $\lambda(\rho) = f'(\rho)$ 

From [Motsch et al, Math. Biosci. Eng. 15 (2018) 1271]

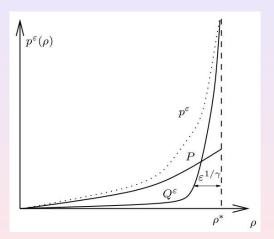
# Model prediction

## Relative gain using the segregation strategy



# Pressure divergence

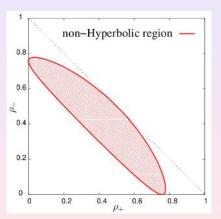
## Pressure term diverging at maximum density



From [Appert-Rolland, Degond, Motsch (2011)]

## Two-Ways Aw-Rascle model

Contrarily to the One-Way Aw-Rascle model, the Two-Ways Aw-Rascle model is not always hyperbolic.

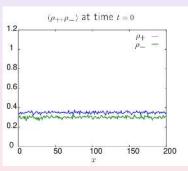


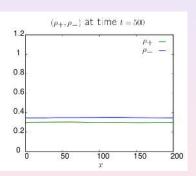
From [Appert-Rolland, Degond, Motsch (2011)]

Example for one particular choice of  $p(\rho_+, \rho_-)$ .

# Numerical simulations of the Two-ways Aw-Rascle model

## Initial condition in the hyperbolic region:

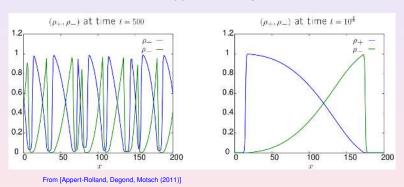




From [Appert-Rolland, Degond, Motsch (2011)]

# Numerical simulations of the Two-ways Aw-Rascle model

## Initial condition outside of the hyperbolic region:



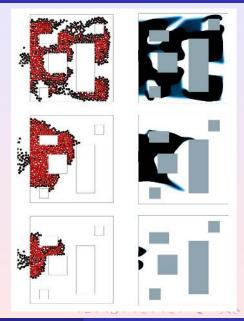
First many clusters; after longer time, only one cluster remains.

# Pbl Macroscopic models

Properties of pedestrians:

- Incompressible at high density
- Undeformable

PhD thesis Aude Roudneff-Chupin (LMO, Orsay, 2011)



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